

1 **Impacts of anthropogenic water regulation on global** 2 **riverine dissolved organic carbon transport**

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12 **Abstract.** Anthropogenic water regulation activities, including reservoir interception, surface water
13 withdrawal, and groundwater extraction, alter riverine hydrologic processes and affect dissolved organic
14 carbon (DOC) export from land to rivers and oceans. In this study, schemes describing soil DOC leaching,
15 riverine DOC transport, and anthropogenic water regulation were developed and incorporated into the
16 Community Land Model 5.0 (CLM 5.0) and the River Transport Model (RTM). Three simulations by the
17 developed model were conducted on a global scale from 1981 to 2013 to investigate the impacts of
18 anthropogenic water regulation on riverine DOC transport. The validation results showed that DOC
19 exports simulated by the developed model were in good agreement with global river observations. The
20 simulations showed that DOC transport in most rivers was mainly influenced by reservoir interception
21 and surface water withdrawal, especially in central North America and eastern China. Four major rivers,
22 including the Danube, Yangtze, Mississippi, and Ganges Rivers, have experienced reduced riverine DOC
23 flows due to intense water management, with the largest effect occurring in winter and early spring. In
24 the Danube and Yangtze River basins, the impact in 2013 was four to five times greater than in 1981,
25 with a retention efficiency of over 50 %. The Ob River basin was almost unaffected. The total impact of
26 anthropogenic water regulation reduced global annual riverine DOC exports to the ocean by
27 approximately $13.36 \pm 2.45 \text{ Tg C yr}^{-1}$, and this effect increased from 4.83 % to 6.20 % during 1981–
28 2013, particularly in the Pacific and Atlantic Oceans.

29 **1. Introduction**

30 Rivers are a pipe linking the two major carbon pools of terrestrial and ocean ecosystems and are one of
31 the key hubs of the global carbon cycle (Cole et al., 2007). According to the Fifth Assessment Report of
32 the Intergovernmental Panel on Climate Change (IPCC AR5), terrestrial ecosystems deliver about 1.7 Pg
33 C per year to rivers through surface and subsurface runoff and about 0.9 Pg C per year to oceans via
34 rivers. Approximately 0.21 Pg of this is dissolved organic carbon (DOC) (Ludwig et al., 1996), which is
35 equivalent to about 1 % of the global net primary productivity (NPP) of terrestrial ecosystems (Zhang,
36 2012). Riverine DOC is a rather highly reactive organic carbon, easily decomposed. It is a direct source
37 of carbon for microbial food webs in rivers and oceans, as well as a source of greenhouse gas emissions
38 from freshwater systems (Li et al., 2019; Tranvik & Jansson, 2002). It deeply affects the biogeochemical
39 cycles of rivers and offshore ecosystems. Therefore, it is important to clarify the transport characteristics
40 of riverine DOC for estimating global carbon budgets.

41 In recent years, anthropogenic water management activities, including reservoir interception, surface
42 water withdrawal, and groundwater extraction, have intensified the degree of interference with natural
43 processes on the surface of river basins, altered the hydrological and hydraulic processes of rivers, and
44 affected material circulation and transportation (Zhang, 2012). For example, extraction from
45 underground aquifers affects hydrological systems, leading to a reduction in subsurface runoff and
46 eventually to decreased soil carbon losses (Zeng et al., 2016). Whereas activities such as irrigation can
47 lead to increased surface runoff, resulting in increased soil carbon losses (Ren et al., 2016). Artificially
48 constructed large reservoirs or dams disrupt the carbon cycle balance of the river continuum in its natural
49 state (Maavara et al., 2017), resulting in retention of DOC and sediment, while lower river velocities and
50 higher material concentrations lead to increased microbial activity in the water body, thus changing the
51 nutrient state of the river ecosystem (Liu et al., 2022). However, the impact of these anthropogenic
52 disturbances on riverine carbon transport has been ignored in estimating the global carbon budget
53 (Regnier et al., 2013).

54 Based on field surveys involving global riverine DOC transport flux estimation, the United Nations
55 Environment Programme has constructed a world river discharge database, GEMS-GLORI, that lists 48
56 attributes of 555 major world rivers (Meybeck, 1982; Meybeck & Ragu, 2012). There are also regional
57 survey programs, such as the Pan-Arctic River Transport of Nutrients, Organic Matter, and Suspended

58 Sediments (PARTNERS, <https://arcticgreatrivers.org/>) and the United States Geological Survey (USGS)
59 Data Center (<https://waterdata.usgs.gov/nwis>), which provide riverine organic carbon flux data for parts
60 of large rivers. Field survey studies are directly limited by data availability and completeness and
61 therefore mostly focus on large rivers in developed regions, making it difficult to cover rivers in other
62 regions. Moreover, only annual averages are usually available, with no long-term time series variation.
63 Some researchers have started to explore the mechanisms of riverine carbon flux changes using empirical
64 statistical models, which combine observed data with driving factors including river basin characteristics
65 (Ludwig et al., 1996), soil carbon and nitrogen ratios (Aitkenhead & McDowell, 2000), land-cover types
66 (Harrison et al., 2005), and river discharge (Fabre et al., 2020). However, the empirical statistical method
67 does not consider complex ecological processes within the watershed and cannot describe material
68 changes in the river network in detail. To identify changes in carbon transport and its driving mechanisms
69 spatially and explicitly, numerous process-based numerical models are currently used for DOC transport
70 simulations. Futter et al. (2007) proposed the integrated catchments model for carbon (INCA-C), which
71 explicitly considers land use, hydrological processes, soil carbon biogeochemical cycles, and surface
72 water processes. Liao et al. (2019) developed a three-dimensional terrestrial ecosystem model (ECO3D)
73 considering the influence of lateral water flows. These models simulate regional riverine DOC dynamics
74 more accurately than earlier models, but their accuracy relies on complex parametric schemes of eco-
75 hydrological processes and extensive data surveys, so that it is difficult to extend these models to global-
76 scale simulations. Wu et al. (2014) integrated ecological driving factors and biogeochemical processes
77 to develop a TRIPLEX-DOC model that predicts DOC metabolism, sorption, desorption, and loss
78 processes in soils. Li et al. (2019) added a river hydrological process module to construct the TRIPLEX-
79 HYDRA model and applied it to simulate global riverine DOC fluxes. However, the model did not
80 consider the impact of human activities on riverine DOC transport. Tian et al. (2015) constructed the
81 dynamic land ecosystem model (DLEM), a fully distributed model that integrates vegetation dynamics
82 with processes such as water, carbon, nitrogen, and phosphorus cycling and the effects of human activities
83 and climate change to simulate DOC flux transport in eastern North American rivers. To better quantify
84 riverine carbon transport processes at watershed scale, Yao et al. (2021) coupled the scale-adaptive water
85 transport model (Li et al., 2013) to the DLEM model and applied the result to two mid-Atlantic
86 watersheds in the United States. Nevertheless, these models failed to consider the effects of

87 anthropogenic water regulation activities. Furthermore, constructing numerical simulation models is a
88 future development direction of riverine carbon flux estimation; at present, models are still not widely
89 used to simulate riverine carbon transport (Camino-Serrano et al., 2018).

90 In this study, we incorporated global soil and riverine DOC transport schemes considering
91 anthropogenic water regulation activities into Community Land Model 5.0 (CLM5.0) and conducted
92 numerical simulations at global scale (spatial resolution of about 1° for the land processes and 0.5° for
93 the river systems) during 1981–2013 to explore the impact of anthropogenic water regulation activities
94 on land-to-ocean riverine DOC transport.

95 **2. Model Development**

96 **2.1. Model Overview**

97 The model was developed based on CLM5.0, which is the land component of the CESM (Community
98 Earth System Model). CLM is widely used to simulate and study land surface ecohydrological processes,
99 surface energy exchange processes, and other biogeochemical processes. The latest version of CLM
100 updates most components of previous versions, explicitly represents land-use and land-cover change,
101 introduces a revised canopy interception parameterization, and significant improvements in soil layer
102 resolution, nitrogen cycle, and the snow model. Moreover, CLM5.0 includes two river routing methods:
103 the Model for Scale Adaptive River Transport (MOSART, Li et al., 2013) and the River Transport Model
104 (RTM). Because the scale of this study was global, the river transport model still uses linear scheme RTM.

105 However, CLM5.0 lacks an expression of the soil DOC leaching process and the DOC transport and
106 transformation process in rivers. Therefore, in this paper, schemes for DOC leaching in soils and DOC
107 transport in rivers will be proposed and incorporated into CLM5.0 to simulate riverine carbon transport.
108 To investigate the effect of anthropogenic water regulation activities on global riverine DOC transport,
109 this study used the scheme proposed by Zeng et al. (2016), and coupled it with DOC transport processes.
110 The model framework is shown in Fig. 1.

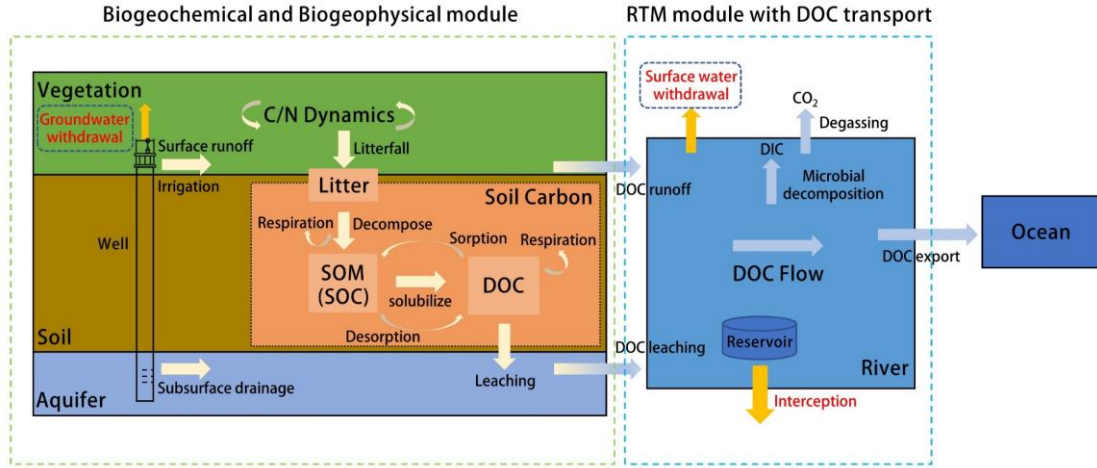


Figure 1. Schematic diagram of the land surface model with riverine dissolved organic carbon (DOC) transport and anthropogenic water regulation (C: carbon; N: nitrogen; SOM: soil organic matter; SOC: soil organic carbon; DIC: dissolved inorganic carbon).

111 2.2. Soil DOC loss to the river

112 Riverine DOC is mainly derived from organic carbon leaching processes in soil ecosystems in the
 113 watershed (Gommet et al., 2022; Li et al., 2019). In CLM5.0, only the leaching process of soil mineral
 114 nitrogen is included, and therefore a DOC production and loss process was introduced in this study. The
 115 soil biochemistry module in CLM5.0 was constructed based on the Century model (Parton et al., 1988),
 116 in which the decomposition of fresh litter into soil organic matter is defined as a transformation cascade
 117 between the coarse woody debris (CWD) pool, the litter pool, and the soil organic matter (SOM) pool.
 118 The NPP produced by plants eventually enters the soil in the form of litter to constitute the soil carbon
 119 pool, accompanied by an intervening loss through microbial heterotrophic respiration. Assuming that
 120 dissolved organic matter (DOM) production is part of the turnover of litter pools and soil organic matter
 121 pools and is proportional to soil water content, DOC production can be expressed as (Gerber et al., 2010):

$$122 \quad P_{DOC,u \rightarrow d} = f_{DOM} \theta CF_{u \rightarrow d}, \quad (1)$$

123 where $P_{DOC,u \rightarrow d}$ ($\text{g C m}^{-2} \text{s}^{-1}$) is the DOC flux from the decomposition process; f_{DOM} is the fraction
 124 that enters the soil DOM pool; θ ($\text{m}^3 \text{m}^{-3}$) is the soil water content; and $CF_{u \rightarrow d}$ ($\text{g C m}^{-2} \text{s}^{-1}$) is the
 125 carbon flux from upstream to downstream carbon pools in the decomposition cascade.

126 Soil organic carbon remaining after plant growth and soil respiration is subject to loss as a dissolved
 127 component leaching from the soil column. In this study, the DOC runoff is defined as the soil DOC in
 128 surface runoff, and the DOC leaching is defined as the subsurface losses of DOC in soil water. The
 129 general equation is written as:

130
$$DOC_{loss} = [DOC]Q_*k_{adsorb} - SR, \quad (2)$$

131 where DOC_{loss} ($\text{g C m}^{-2} \text{s}^{-1}$) denotes the soil DOC runoff or leaching, Q_* ($\text{kgH}_2\text{O m}^{-2} \text{s}^{-1}$) denotes the
 132 surface runoff or subsurface discharge, $[DOC]$ ($\text{g C kgH}_2\text{O}^{-1}$) is the DOC concentration in the soil water
 133 solution:

134
$$[DOC] = \frac{NS_{DOC}}{WS_{tot_soil}}, \quad (3)$$

135 where WS_{tot_soil} ($\text{kgH}_2\text{O m}^{-2}$) is the total mass of soil water content integrated over the soil column and
 136 NS_{DOC} (g C m^{-2}) is the DOC in the soil pool.

137 Soil DOC readily complexes with metal ions in the soil and forms soil agglomerates, which enable
 138 soil DOC to be adsorbed onto soil particles. The DOC adsorption coefficients can be estimated as (Li et
 139 al., 2019; Neff & Asner, 2001):

140
$$k_{adsorb} = \frac{X_i}{X_i + RE}, \quad (4)$$

141
$$RE = mX_i - b, \quad (5)$$

142 where X_i (mg g soil^{-1}) represents the initial DOC concentration, RE (mg g soil^{-1}) is the amount of DOC
 143 desorbed (negative value) or adsorbed (positive value), calculated by the simple initial mass (IM) linear
 144 isotherm, m (dimensionless coefficient) and b (mg g soil^{-1}) can be considered as measures of potential
 145 DOC sorption and desorption by soil.

146 The soil heterotrophic respiration flux of DOC, SR ($\text{g C m}^{-2} \text{s}^{-1}$), is estimated by an empirical function
 147 (Janssens and Pilegaard, 2003):

148
$$SR = R_{10}Q_{s10} \frac{T-10}{10}, \quad (6)$$

149 where T ($^{\circ}\text{C}$) is the soil temperature; R_{10} is the soil heterotrophic respiration flux at a soil temperature of
 150 10°C ; Q_{s10} is the soil respiration temperature sensitivity.

151 It is necessary to limit the total DOC runoff/leaching flux at each time step so that it does not exceed
 152 the total amount of DOC:

153
$$DOC_{loss} = \min(DOC_{loss}, \frac{NS_{DOC}}{\Delta t}). \quad (7)$$

154 **2.3. Riverine DOC transport**

155 Soil DOC enters the river network system along with surface and subsurface runoff, where it is lost due
 156 to processes such as microbial degradation. Therefore, based on the water transport framework, the large-

157 scale riverine DOC transport equation can be defined as:

$$158 \quad \frac{dS_{DOC}}{dt} = F_{DOC}^{in} - F_{DOC}^{out} + R_{DOC} + L_{DOC} - k_{doc} * Q_{10}^{\frac{rt-20}{10}} * S_{DOC}, \quad (8)$$

159 where S_{DOC} (kg C) is DOC storage within the current grid cell; R_{DOC} (kg C s⁻¹) and L_{DOC} (kg C s⁻¹)
160 represent soil DOC runoff and leaching; k_{doc} (s⁻¹) is the DOC decomposition rate in the river; Q_{10}
161 (=2.0) denotes the temperature coefficient; rt (°C) represents the river water temperature, which is
162 calculated by a large-scale river water temperature model (Liu et al., 2020; van Vliet et al., 2012; Yearsley,
163 2009); F_{DOC}^{in} (kg C s⁻¹) is the sum of inflows of riverine DOC from neighboring upstream grid cells;
164 and F_{DOC}^{out} (kg C s⁻¹) is the riverine DOC flux leaving the current grid cell, which is calculated as follows:

$$165 \quad F_{DOC}^{out} = \frac{vS_{DOC}}{d}, \quad (9)$$

166 where v (m s⁻¹) is the effective riverine flow velocity, which is estimated by a simplified Manning's
167 equation (Oleson et al., 2013); d is the Euclidean distance between two adjacent grid-cell centers.

168 **2.4. Anthropogenic water regulation**

169 Anthropogenic water regulation includes reservoir interception, surface water withdrawal, and
170 groundwater extraction and use. Because reservoir interception and surface water withdrawal are closely
171 related, they are together called surface water regulation. This study coupled the global reservoir
172 operation scheme (Hanasaki et al., 2006) with RTM using the method of Liu et al. (2019) to represent
173 the interception effect of reservoirs on runoff and solutes. The method assumed that the inflow from the
174 reservoir was the outflow from the current grid cell. Released flow from the reservoir was adjusted for
175 specific uses (flood control, irrigation, etc.), and surface withdrawals were deducted from the released
176 water (see Sect. S1 in the Supplement).

177 Surface water is extracted directly from natural rivers and reservoirs to meet human water demands
178 (Wang et al., 2020; Xie et al., 2020; Liu et al., 2019):

$$179 \quad S_{sw}' = S_{sw} - q_{sw}\Delta t, \quad (10)$$

180 where S_{sw}' (mm) is the surface water storage after extraction; S_{sw} (mm) is the original surface water
181 storage; q_{sw} (mm s⁻¹) is the rate of surface water intake; Δt denotes the model time step.

182 The groundwater extraction process can be expressed as (Zeng et al., 2016):

$$183 \quad S_{gw}' = S_{gw} - q_{gw}\Delta t, \quad (11)$$

184
$$h' = h - \frac{q_{gw}\Delta t}{s}, \quad (12)$$

185 where S_{gw} (mm) is the original unconfined aquifer water storage; q_{gw} (mm s⁻¹) is the rate of
 186 groundwater pumping; h (mm) represents the original groundwater table depth; s is the aquifer-
 187 specific yield; S_{gw}' (mm) and h' (mm) denote the aquifer water storage and the groundwater table depth
 188 after pumping.

189 Human water use can be divided into agricultural irrigation water and other industrial and domestic
 190 water, where irrigation water is considered as effective precipitation directly back to the soil surface and
 191 other water is directly added to the model surface runoff and evapotranspiration fluxes in a certain
 192 proportion (Zou et al., 2015). This process can be estimated by the following equations:

193
$$q_{top} = q_{top} + q_{irrig}, \quad (13)$$

194
$$q_{surf} = q_{surf} + 0.3q_{ind} + 0.3q_{dom}, \quad (14)$$

195
$$Q_{evap} = q_{evap} + 0.7q_{ind} + 0.7q_{dom}, \quad (15)$$

196 where q_{top} (mm s⁻¹) is the rate of net water flow entering the soil surface; q_{surf} and q_{evap} (mm s⁻¹) are
 197 surface runoff and evaporation; and q_{irrig} , q_{ind} , and q_{dom} (mm s⁻¹) denote irrigation, industrial, and
 198 domestic water respectively. The coefficients were set to 0.3 and 0.7 due to the limitation of data (Liu et
 199 al., 2019; Zou et al., 2014).

200 **2.5. DOC transfer induced by water withdrawal and use**

201 Anthropogenic water regulation activities also affect DOC transport processes between land and river. It
 202 was assumed here that (1) only the interception effect of reservoirs would be considered, ignoring the
 203 migration transformation process in reservoirs, and the loss rate in reservoirs would be equal to that in
 204 rivers; (2) because groundwater extraction usually occurs *in situ* and will pass through the filtering effect
 205 of the soil layer, the part of DOC that returned to soil with groundwater extraction was ignored; (3) the
 206 loss rate in the process of DOC returning to soil was equal to that in rivers.

207 The process of reservoir interception leading to retention of carbon in rivers can be expressed as:

208
$$F_{DOC,r} = \frac{v(con_r\Delta Q_r)}{d}, \quad (16)$$

209 where $F_{DOC,r}$ (kg C s⁻¹) denotes the DOC flux retained by the reservoir; con_r (kg C m⁻³) is the DOC
 210 concentration in the reservoir; ΔQ_r (m³) is the water volume change in the reservoir.

211 The DOC flux extracted from surface water is calculated based on the intake rate and the solute

212 concentration in the current grid cell and enters the soil DOC pool after irrigation. The reduction in soil
213 DOC leaching due to groundwater extraction is then calculated based on soil DOC concentration and
214 groundwater pumping rate.

215 **3. Data and Experimental Design**

216 **3.1. Data Sources**

217 The climate input forcing data set ($0.5^\circ \times 0.5^\circ$) used for the model proposed in this study was obtained
218 from CRU-NCEP Version 7 (Viovy, 2018), including air temperature, humidity, incoming solar radiation,
219 precipitation, surface pressures, and surface winds. The basic land-surface datasets required to drive the
220 model were set up using the default CLM 5.0 settings with a spatial resolution of $0.9^\circ \times 1.25^\circ$; more
221 details are available in the technical notes (Lawrence et al., 2018). The global monthly mean atmospheric
222 CO₂ concentration dataset came from the NOAA/Earth System Research Laboratory
223 (<https://www.esrl.noaa.gov/gmd/ccgg/trends/global.html>).

224 Reservoir information was obtained from the Global Reservoir and Dam Database (GRanD, Lehner et
225 al., 2011), containing information on 6,862 dams and their associated reservoirs worldwide, and
226 interpolated to a spatial resolution of $0.5^\circ \times 0.5^\circ$ (Fig.S2).

227 The human water use activity dataset was derived from the global long-term surface and groundwater
228 withdrawal dataset estimated by Liu et al. (2019). The dataset has a spatial resolution of $0.5^\circ \times 0.5^\circ$ and
229 contains agricultural, industrial, and domestic water demands from 1958 to 2017. It was derived based
230 on five datasets: the water use dataset from the Food and Agricultural Organization (FAO), a shape file
231 data of national boundaries, the Global Map of Irrigation Areas, version 5 (GAMIP5; Siebert et al., 2013),
232 the historical monthly soil moisture levels and saturated soil moisture levels (Zeng et al., 2017), and the
233 FAO water information system for 2010, which contained the agricultural, industrial, and municipal
234 water withdrawals.

235 **3.2. Observation Data**

236 Because there are few datasets of long time-series observations of DOC fluxes for large global rivers,
237 annual averages were used to validate the model simulations. The dataset was derived from the database
238 developed by Dai et al. (2012), which provides discharge and DOC flux observations for sites on the

239 world's major large rivers. These sites were globally distributed and were influenced by various climatic
 240 and human activities.

Table 1. Summary of the main datasets used in this study

Dataset	Resolution	Time period	Data Source
CRU-NCEP V7 forcing	0.5°/6 hr	1981-2013	Viovy (2018)
Surface water and groundwater withdrawal and use	0.5°	1958-2017	Liu et al. (2019)
Reservoir information	Site	Around 2011	Lehner et al. (2011)
River discharge	Site	Annual before 2009	Dai et al. (2012)
DOC export	Site	Annual before 2009	Dai et al. (2012)

241 3.3. Experimental Design

242 To investigate the effect of anthropogenic water regulation on DOC transport in rivers, three sets of
 243 simulations were designed using the developed model (Table 1). The first simulation (CTL) was a control
 244 experiment without considering any anthropogenic water regulation activities. The second simulation
 245 (EXPA) only considered surface water regulation, and the last simulation (EXPB) considered all
 246 anthropogenic water regulation. All simulations were run from 1981 to 2013 with a spatial resolution of
 247 $0.9^\circ \times 1.25^\circ$ for the land-surface module and $0.5^\circ \times 0.5^\circ$ for the RTM. The results were output on a
 248 monthly scale. Before the formal numerical simulations, the 1901–1920 atmospheric forcing data cycle
 249 was used to drive the model without any anthropogenic water regulation as the spin-up run to reach an
 250 equilibrium state.

Table 2. Experimental design

Name	Period	Surface regulation	Groundwater regulation
CTL	1981–2013	✗	✗
EXPA	1981–2013	✓	✗
EXPB	1981–2013	✓	✓

251 4. Results

252 4.1. Model Evaluation

253 Figure 2 shows the spatial distribution of multi-year average soil DOC losses, which are the sum of DOC
 254 surface runoff and subsurface leaching. The results show that the global distribution of soil DOC losses
 255 varied widely, especially in Russia and Southeast Asia, western Africa, and tropical South America,

256 where the losses exceeded $1.8 \times 10^4 \text{ kg C km}^{-2} \text{ yr}^{-1}$, whereas low runoff arid regions such as northwestern
 257 China, India, and North Africa had the smallest soil DOC losses. The tropics and the temperate regions
 258 of the Northern Hemisphere were the regions with the highest soil DOC losses, which is generally
 259 consistent with previous studies (Harrison et al., 2005).

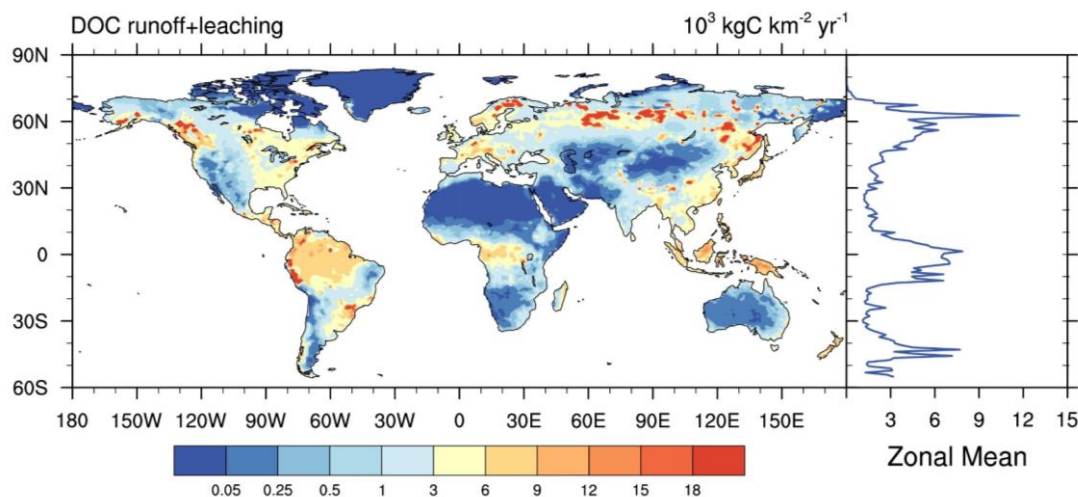


Figure 2. Spatial distribution and zonal mean of multi-year average soil DOC losses from 1981 to 2013.

260 The multi-year average river discharges and DOC export fluxes simulated by the developed model
 261 were then compared with observed data. Because the model resolution was $0.5^\circ \times 0.5^\circ$, only 106 rivers
 262 with watershed areas larger than $2,500 \text{ km}^2$ were selected. The simulated river discharges were slightly
 263 underestimated (Fig. 3c), but fit well with observations (Fig. 3a) and provided a solid basis for subsequent
 264 simulation of river carbon exports. In addition, the simulated riverine DOC export fluxes tended to be
 265 overestimated in temperate regions and underestimated in the tropics (Fig. 3d), but were close to the 1:1
 266 line compared to the observed DOC fluxes, with R^2 reaching 0.61 and significantly correlated (Fig. 3b).
 267 Moreover, the total global river DOC export fluxes simulated by the proposed model were compared
 268 with the results of previous studies. We estimated that the global terrestrial ecosystem delivers about
 269 199.78 ± 36.63 (± 1 standard deviation) Tg of DOC per year to the ocean via rivers, which was in the
 270 middle of the values derived from previous studies (Table 2). Therefore, it could be believed that the
 271 model has reasonable accuracy and can be applied to global-scale riverine DOC export simulation studies.

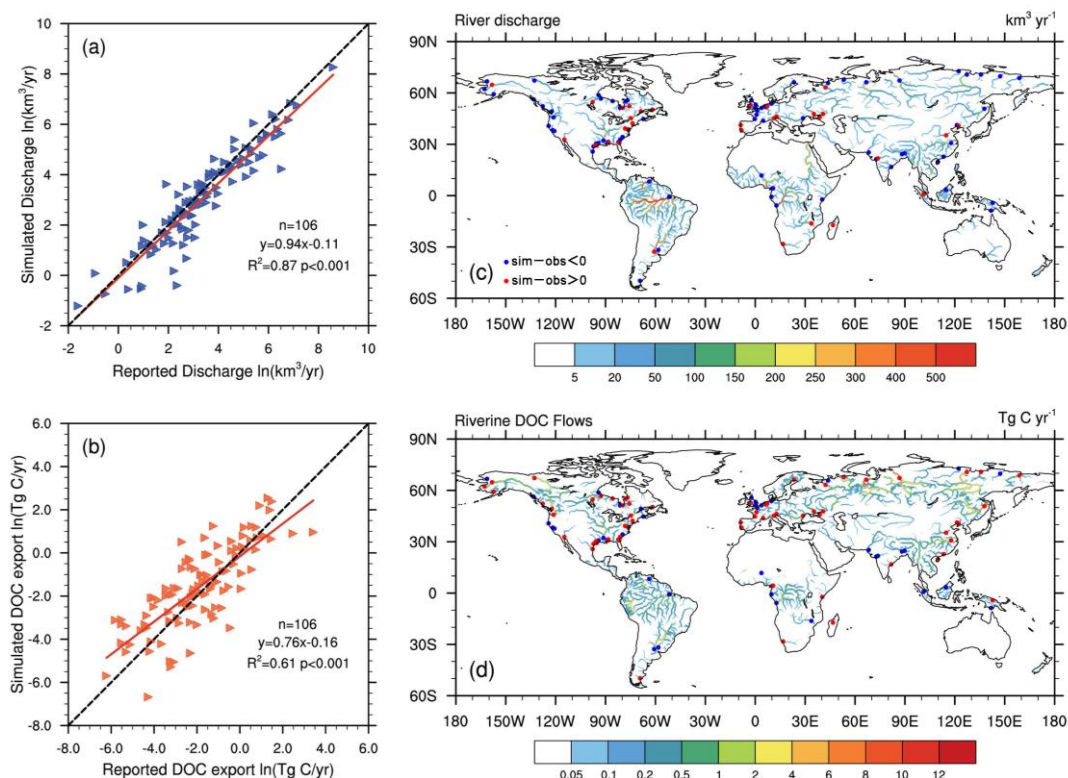


Figure 3. Simulated and reported annual (a) river discharge and (b) riverine DOC export flux for 106 global rivers. Spatial distributions of (c) annual discharge and (d) annual riverine DOC exports during 1981–2013. The dots in the map correspond to the locations of the 106 river sites, where blue dots indicate sites that are simulated underestimates and red dots indicate sites that are simulated overestimates.

Table 3. Comparison of simulated global total riverine DOC export fluxes with previous studies

Method	DOC (Tg C yr ⁻¹)	Data Source
GEMS-GLORI	215	Meybeck (1982)
Empirical model	204	Smith & Hollibaugh (1993)
Empirical model	204.81	Ludwig et al. (1996)
Global C: N	361	Aitkenhead & McDowell (2000)
NEWS-DOC	170	Harrison et al. (2005)
Global-NEWS	170	Seitzinger et al. (2005)
Statistical estimation	246	Cai (2011)
Statistical estimation	232.22	Drake et al. (2018)
TRIPLEX-HYDRA	240	Li et al. (2019)
Empirical model	131.6	Fabre et al. (2020)
DISC-CARBON	132	van Hoek et al. (2021)
CLM5.0-RTM	199.78	This study

272 4.2. Effects of surface water regulation on riverine DOC transport

273 The difference between EXPA and CTL was used to obtain the effect of surface water regulation on land

274 surface hydrological variables. Surface water use has resulted in changes in latent and sensible heat fluxes
 275 in most global irrigation water-using regions (Fig. 4a, 4b), especially in arid or semi-arid regions such as
 276 northern China, India, and the central United States, where latent heat fluxes have increased and sensible
 277 heat fluxes have decreased. Soil and surface temperatures in these regions have also decreased due to the
 278 cooling effect of irrigation (Fig. 4c, 4d). Figure 4e shows that irrigation led to an overall increase in soil
 279 moisture, especially in northern India, Western Europe, and the midwestern United States. In addition,
 280 irrigation also led to an increase in total runoff (Fig. 4f).

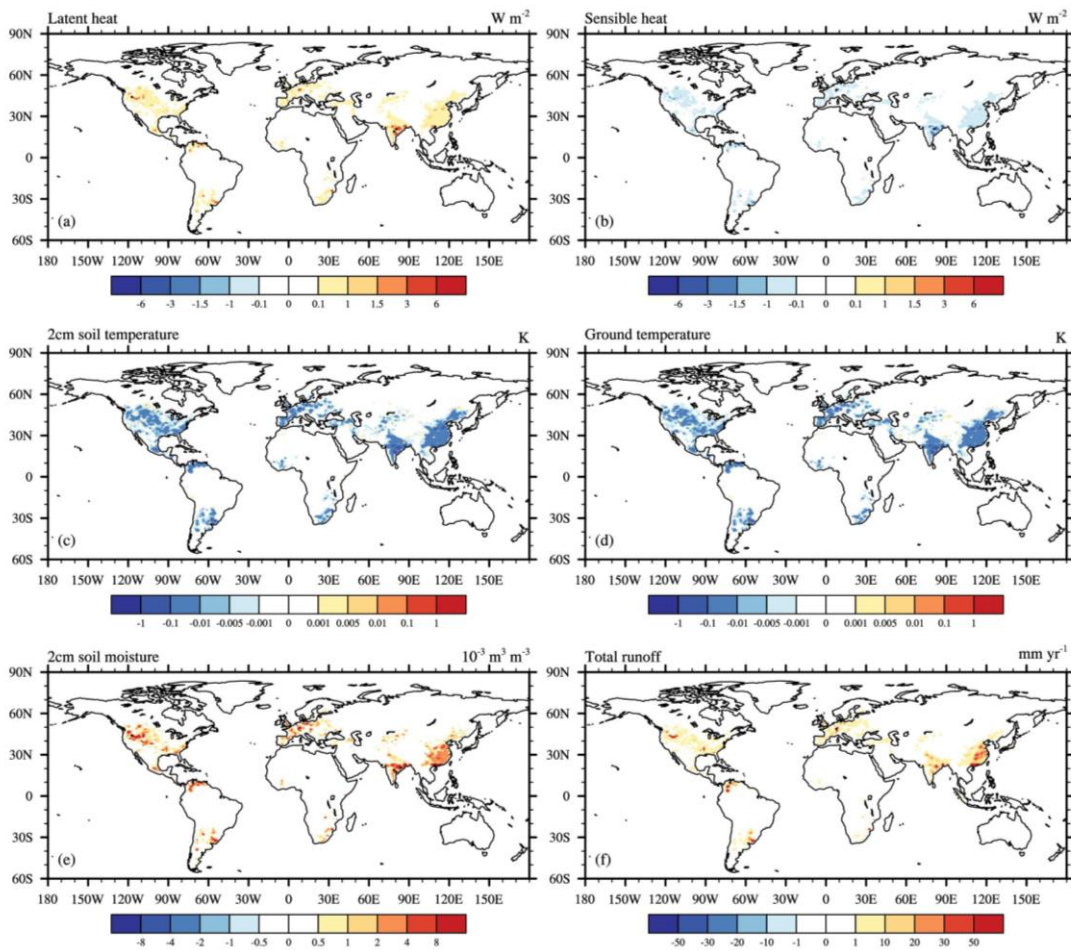


Figure 4. Spatial distribution of multi-year average differences in land surface hydrological variables between EXPA and CTL from 1981 to 2013: (a) latent heat flux, (b) sensible heat flux, (c) 2 cm soil temperature, (d) surface temperature, (e) 2 cm soil moisture, (f) total runoff. This figure demonstrates the effects of surface water regulation on land surface hydrological variables. The black dots are the regions that pass the significance *t*-test at the 95 % confidence level.

281 Figures 5a and 5b display the effects of surface water regulation on soil carbon losses. Specifically,
 282 the hotspots of significantly increased surface DOC runoff were in areas of high agricultural influence,

283 such as the central United States, northern India, and northern and eastern China, reaching up to 2,000
 284 $\text{kg C km}^{-2} \text{yr}^{-1}$, but the increase in subsurface leaching was relatively small. This may have been the case
 285 because surface water withdrawals from rivers and reservoirs were returned to the soil by irrigation,
 286 bringing back some DOC, directly increasing surface runoff, and also increasing subsurface runoff, and
 287 thus increasing soil DOC losses.

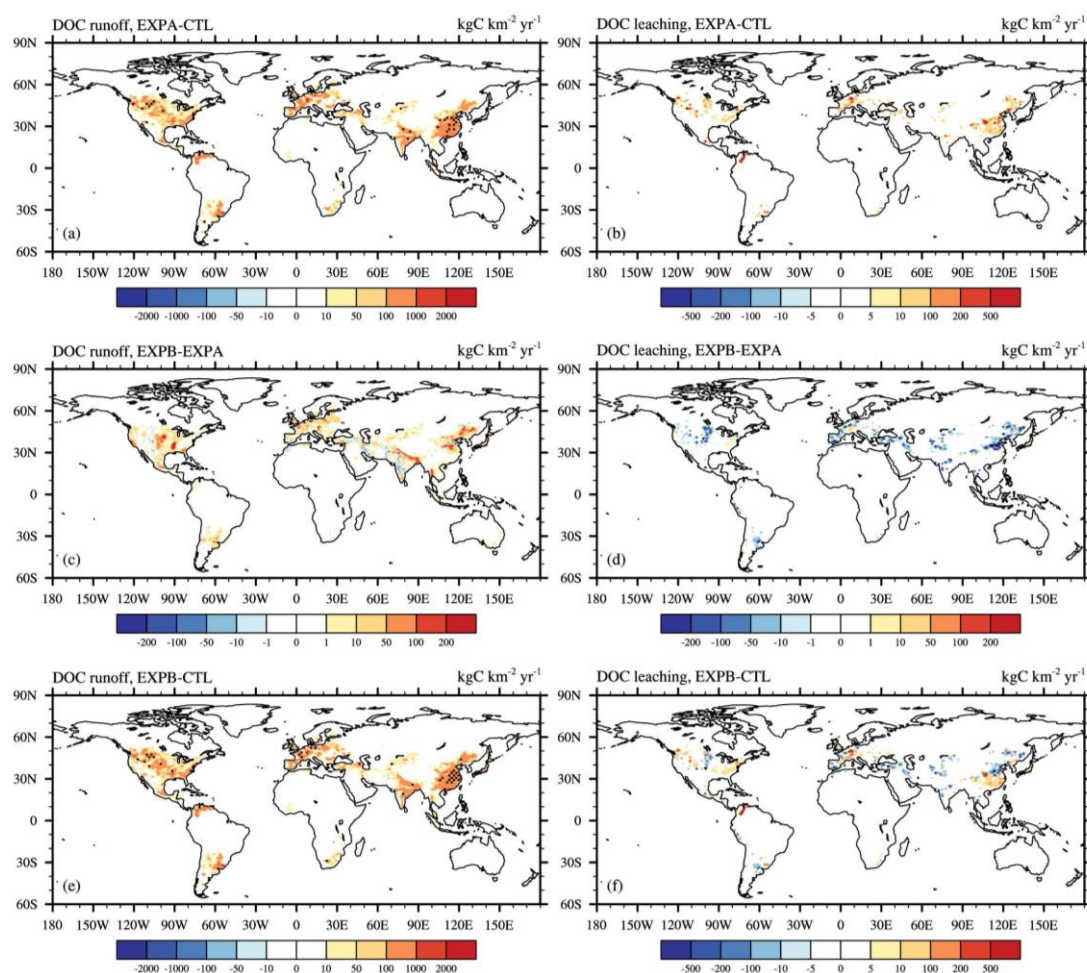


Figure 5. Spatial distribution of the multi-year average differences between different experiments from 1981 to 2013 in the (a) soil DOC runoff, EXPA-CTL, (b) soil DOC leaching, EXPA-CTL, (c) soil DOC runoff, EXPB-EXPA, (d) soil DOC leaching, EXPB-EXPA (e) soil DOC runoff, EXPB-CTL, (f) soil DOC leaching, EXPB-CTL. This figure demonstrates the effects of (a, b) surface water regulation, (c, d) groundwater regulation, and (e, f) anthropogenic water regulation on soil DOC losses. The black dots are the regions that pass the significance *t*-test at the 95 % confidence level.

288 From Fig. 6a and Fig. 6b, surface water regulation had a significant effect on river discharge and
 289 riverine DOC flow. The combined effects of reservoir interception and surface water withdrawal reduced
 290 the discharge and DOC export of most rivers globally, with significant reductions of more than 50 Gg C
 291 yr^{-1} in the Yangtze, Yellow, Mississippi, and Ganges Rivers and in some basins in Western Europe. Some

292 rivers in northern South America experienced increased riverine DOC export, but not significantly,
 293 probably because the increase in river flow caused by agricultural irrigation could have been greater than
 294 the decrease caused by surface water regulation.

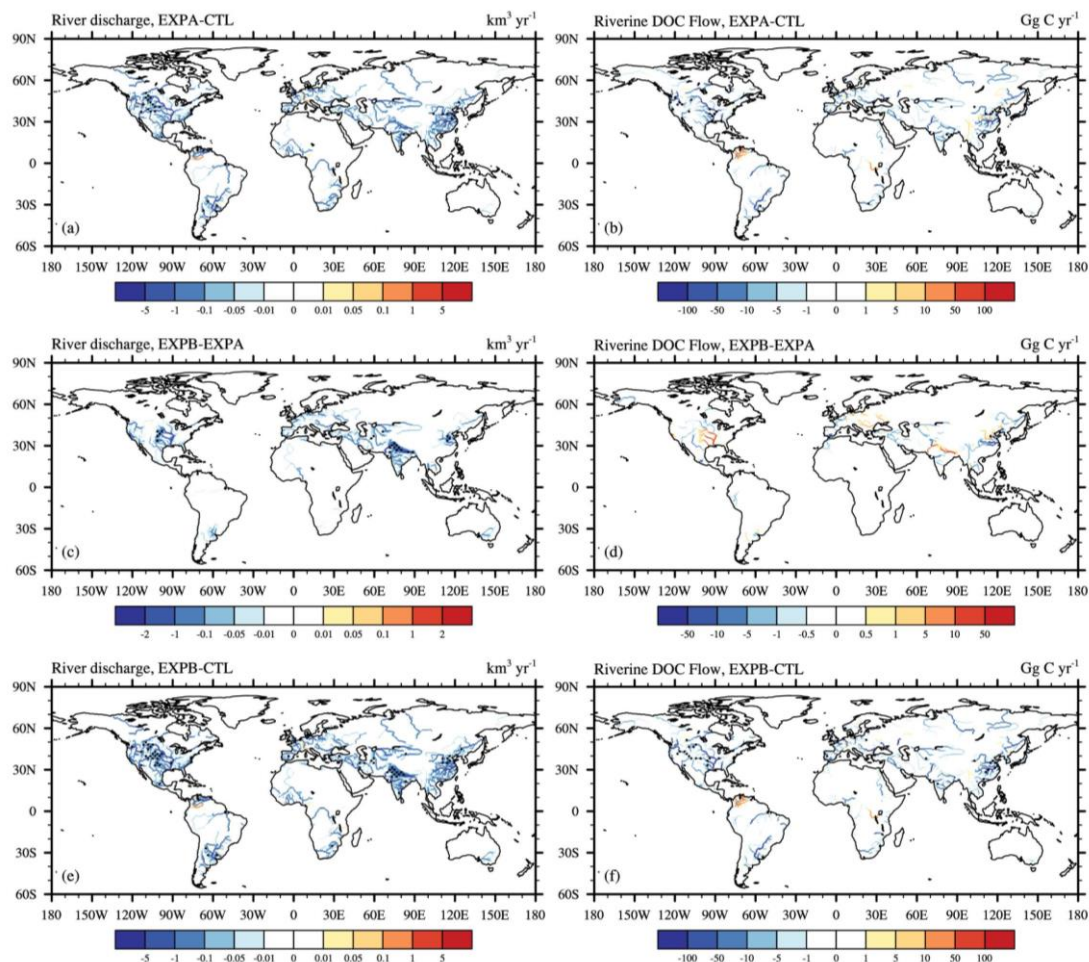


Figure 6. Spatial distribution of the multi-year average differences between different experiments from 1981 to 2013 in the (a) river discharge, EXPA-CTL, (b) riverine DOC flow, EXPA-CTL, (c) river discharge, EXPB-EXPA, (d) riverine DOC flow, EXPB-EXPA (e) river discharge, EXPB-CTL, (f) riverine DOC flow, EXPB-CTL. This figure demonstrates the effects of (a, b) surface water regulation, (c, d) groundwater regulation, and (e, f) anthropogenic water regulation on the river discharge and riverine DOC flow rate. The black dots are the regions that pass the significance *t*-test at the 95 % confidence level.

295 The blue line in Fig. 7 represents the time-series variation of surface water regulation on global riverine
 296 organic carbon to the ocean. Surface water regulation greatly reduced global riverine DOC transport to
 297 the ocean, from -11.1 Tg yr^{-1} in 1981 to -16.4 Tg yr^{-1} in 2013 (Fig. 7a), with a multi-year average
 298 retention efficiency of about 6 %. This may be related to the fact that the reservoir adjusting the river
 299 discharge and intercepting the riverine DOC. The regions most affected by surface water regulation were

300 the Pacific and Atlantic Oceans, and as surface water use in these regions became more frequent, the
 301 reduction in DOC delivery to the ocean was intensified each year. There was no significant change in the
 302 Arctic Ocean region, which may have been due to less anthropogenic disturbance in this area.

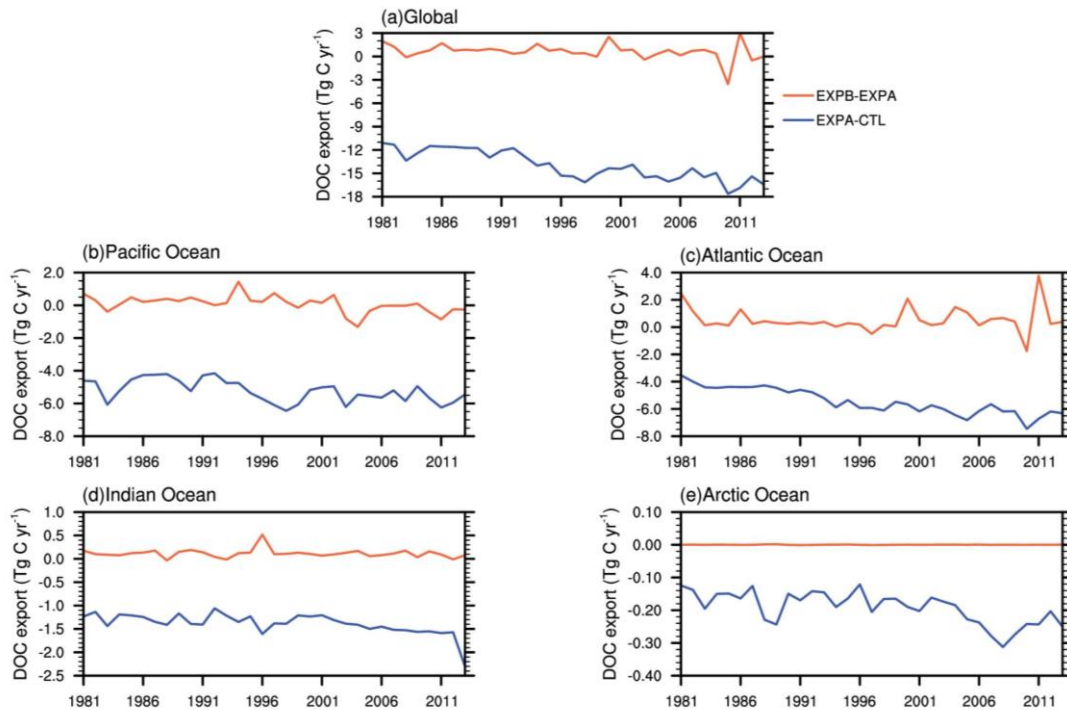


Figure 7. Time series of changes in DOC export to oceans due to surface water (blue line) and groundwater regulation (orange line) from 1981 to 2013: (a) global, (b) Pacific Ocean, (c) Atlantic Ocean, (d) Indian Ocean, (e) Arctic Ocean.

303 **4.3. Effects of groundwater regulation on riverine DOC transport**

304 The effects of groundwater regulation on land surface hydrological variables were obtained using the
 305 difference between EXPB and EXPA, as shown in Fig. 8. It can be seen that groundwater extraction
 306 increased latent heat fluxes, decreased sensible heat fluxes, decreased soil and surface temperatures, and
 307 increased soil moisture in most regions of the world. The most significant impacts were in northern China,
 308 northern India, Pakistan, and the central United States, where climate conditions are dry and groundwater
 309 extraction is frequent. Unlike surface water regulation, groundwater extraction has a negative impact on
 310 total runoff (Fig. 8f). Because groundwater is extracted from underground aquifers, whereas surface
 311 water is extracted from rivers and reservoirs, surface water use directly increases total land surface runoff.
 312 However, the impact of groundwater extraction on runoff depends on the groundwater pumping rate,
 313 infiltration rate, and soil evaporation capacity. The increase in latent heat flux leads to an increase in

314 surface evapotranspiration, which results in a decrease in runoff.

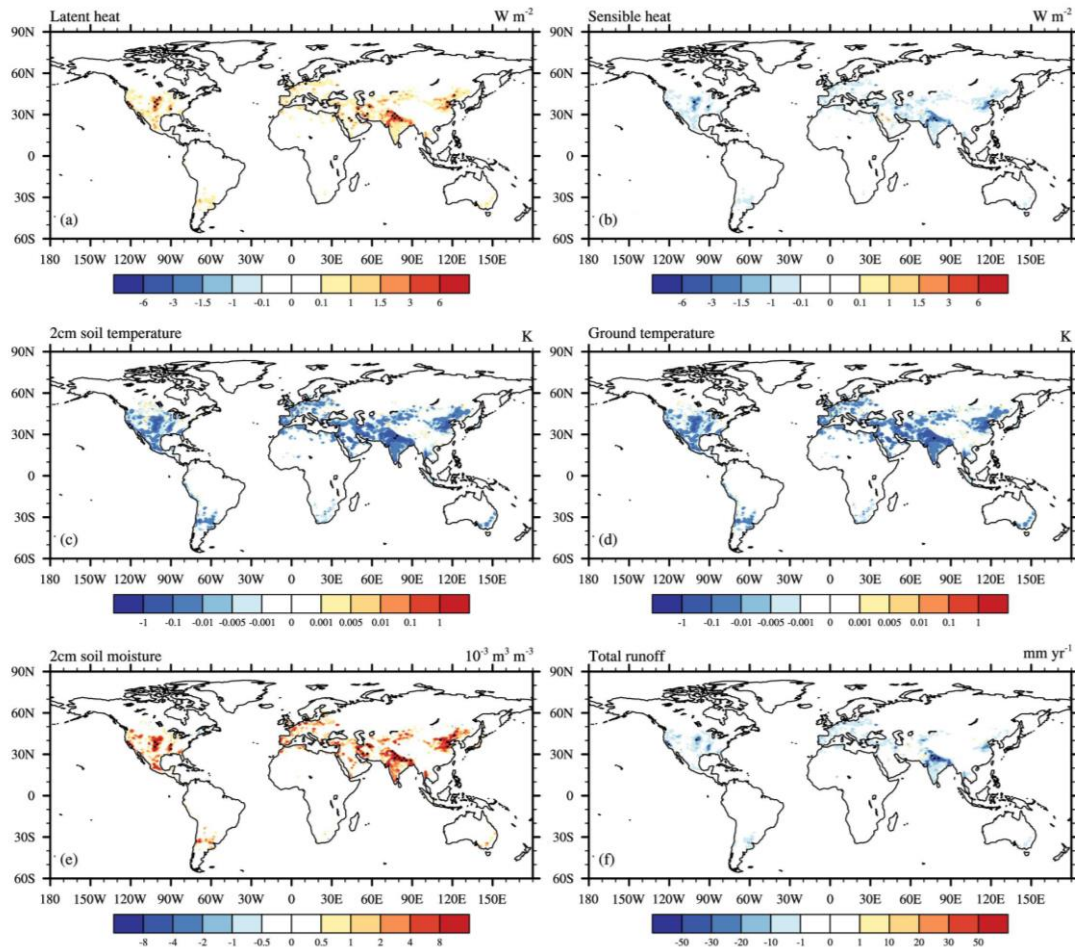


Figure 8. Spatial distribution of multi-year average differences in land surface hydrological variables between EXPB and EXPA from 1981 to 2013: (a) latent heat flux, (b) sensible heat flux, (c) 2 cm soil temperature, (d) surface temperature, (e) 2 cm soil moisture, (f) total runoff. This figure demonstrates the effects of groundwater regulation on land surface hydrological variables. The black dots are the regions that pass the significance *t*-test at the 95 % confidence level.

315 Figures 5c and 5d show the effect of groundwater regulation on soil carbon losses. On the one hand,
316 extracting water from underground aquifers led to a reduction in subsurface runoff and a consequent
317 reduction in DOC leaching, especially in northern China and the central United States, where DOC
318 leaching reductions reached $200 kg C yr^{-1}$. On the other hand, groundwater irrigation led to an increase
319 in surface runoff, which led to an increase in DOC runoff. The most affected areas are characterized by
320 well-developed agriculture.

321 Figures 6c and 6d show the spatial distribution of the effects of groundwater regulation on river
322 discharge and DOC export from 1981 to 2013. It can be seen that river discharge significantly decreased
323 in areas with high groundwater extraction rates, such as the central United States, Pakistan, Afghanistan,

324 and northern China, resulting in a decrease in riverine DOC export. The largest decrease occurred in the
325 Yangtze River Basin in China, reaching 50 Gg C yr⁻¹; most other rivers were around 10 Gg C yr⁻¹. In
326 addition, although river discharge was reduced in some river sections, soil DOC loss was higher, and
327 DOC export fluxes were still increasing, especially in the lower Yellow River, Mississippi River, and
328 Ganges River basins. This was due to the predominance of agricultural irrigation water in these regions.

329 The amount of carbon flux variation influenced by groundwater regulation was relatively small
330 compared to that influenced by surface water regulation, but there was some interannual fluctuation, with
331 the greatest impact during 2009–2012 (Fig. 7). The intermittent increase and decrease of the variation
332 indicate that river carbon transport fluxes did not decrease directly with increases of groundwater
333 pumping rate, but were also related to the complex carbon and nitrogen cycling processes in terrestrial
334 ecosystems. In addition, irrigation after groundwater extraction from an underground aquifer did not
335 consider directly sending DOC back to the soil carbon pool, and therefore the carbon flux changes were
336 smaller. Because groundwater regulation activities are mostly concentrated in the northern temperate
337 zone, the Pacific and Atlantic regions were the most obviously affected, whereas the remaining regions
338 did not change much.

339 **4.4. Effects of anthropogenic water regulation on riverine DOC transport**

340 This section discusses the combined effects of anthropogenic water regulation on soil and riverine carbon
341 transport using the EXPB minus CTL results. The effects of anthropogenic water regulation on total
342 runoff both increased and decreased globally (Fig. 9f). The western United States, Venezuela, and
343 northern China showed an increase in runoff due to the high intensity of irrigation water use in agriculture.
344 In contrast, regions such as northern India and the central United States showed a decrease in runoff due
345 to frequent groundwater extraction. Overall, human water regulation activities led to an increase in latent
346 heat fluxes and soil moisture and a decrease in sensible heat fluxes and in soil and ground temperatures.

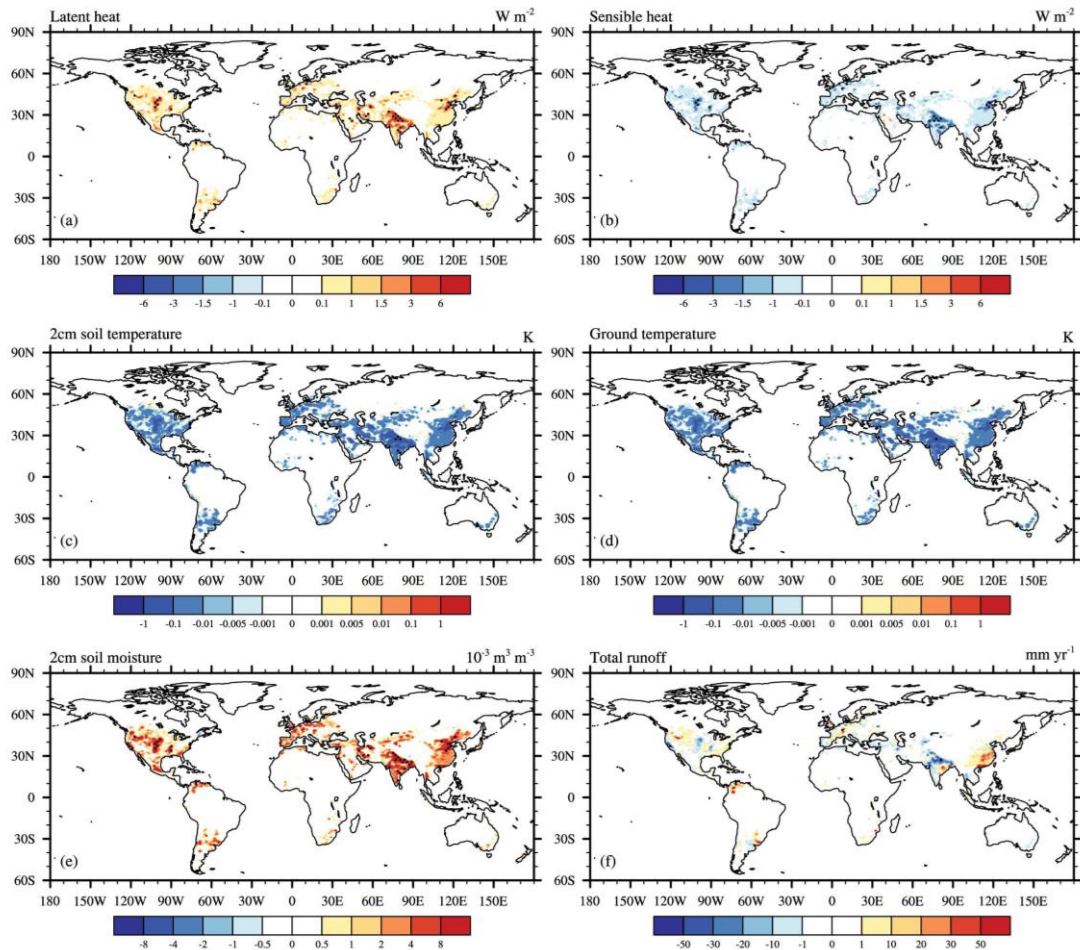


Figure 9. Spatial distribution of multi-year average differences in land surface hydrological variables between EXPB and CTL from 1981 to 2013: (a) latent heat flux, (b) sensible heat flux, (c) 2 cm soil temperature, (d) surface temperature, (e) 2 cm soil moisture, (f) total runoff. This figure demonstrates the effects of anthropogenic water regulation on land surface hydrological variables. The black dots are the regions that pass the significance *t*-test at the 95 % confidence level.

347 Figure 5e shows that soil DOC runoff increased, especially in northern China and the midwestern
 348 United States. DOC leaching decreased in some river sections (Fig. 5f), but not significantly. Although
 349 soil DOC runoff showed an overall increase, DOC export fluxes decreased in most rivers globally due to
 350 water regulation (Fig. 6f). On the one hand, human water use activities led to a decrease in river discharge
 351 (Fig. 6e), and on the other hand, reservoirs have intercepted part of riverine DOC, which led to an increase
 352 in microbial activity, resulting in a decrease in river carbon flux. In contrast, in the Mississippi and
 353 Ganges River basins, although groundwater regulation increased their DOC export fluxes (Fig. 6d), they
 354 still showed a decrease under the negative feedback effect of surface water regulation, indicating that
 355 most rivers globally are mainly influenced by reservoir interception and surface water withdrawal.

356 Five typical rivers were selected to exhibit how anthropogenic water regulation affects monthly and
357 annual average DOC flows in rivers. The selected rivers were the Mississippi River in the United States,
358 the Danube River in Europe, the Ob River in Russia, the Yangtze River in China, and the Ganges River
359 in India. Figure 10 displays the seasonal and interannual variation of DOC flow rates in the five rivers as
360 calculated by the three sets of simulations respectively. Anthropogenic water regulation had a significant
361 impact on the Mississippi, Danube, Yangtze, and Ganges Rivers, which decreased significantly in winter
362 and early spring, whereas the Ob River was almost unaffected. This was the case because of weak water
363 management activities in the Ob River, whereas the other subtropical and temperate rivers had intense
364 water management activities and significant seasonal variation in runoff. In addition, only the Mississippi,
365 Yangtze, and Ganges rivers were affected by minor groundwater regulation, usually occurring during dry
366 periods, whereas in most seasons, the rivers were affected only by surface water regulation (including
367 reservoir interception). The annual results showed a significantly strengthening trend of riverine DOC
368 reduction due to the influence of anthropogenic water regulation, especially in the Danube and Yangtze
369 Rivers, where the retention percentage in 2013 was four to five times higher than in 1981, up to more
370 than 50 %, indicating a clear intensification of human water management activities. The influence on the
371 Mississippi and Ganges Rivers increased slightly and stabilized at about 30–40 %, whereas the influence
372 on the Ob River was almost 0.

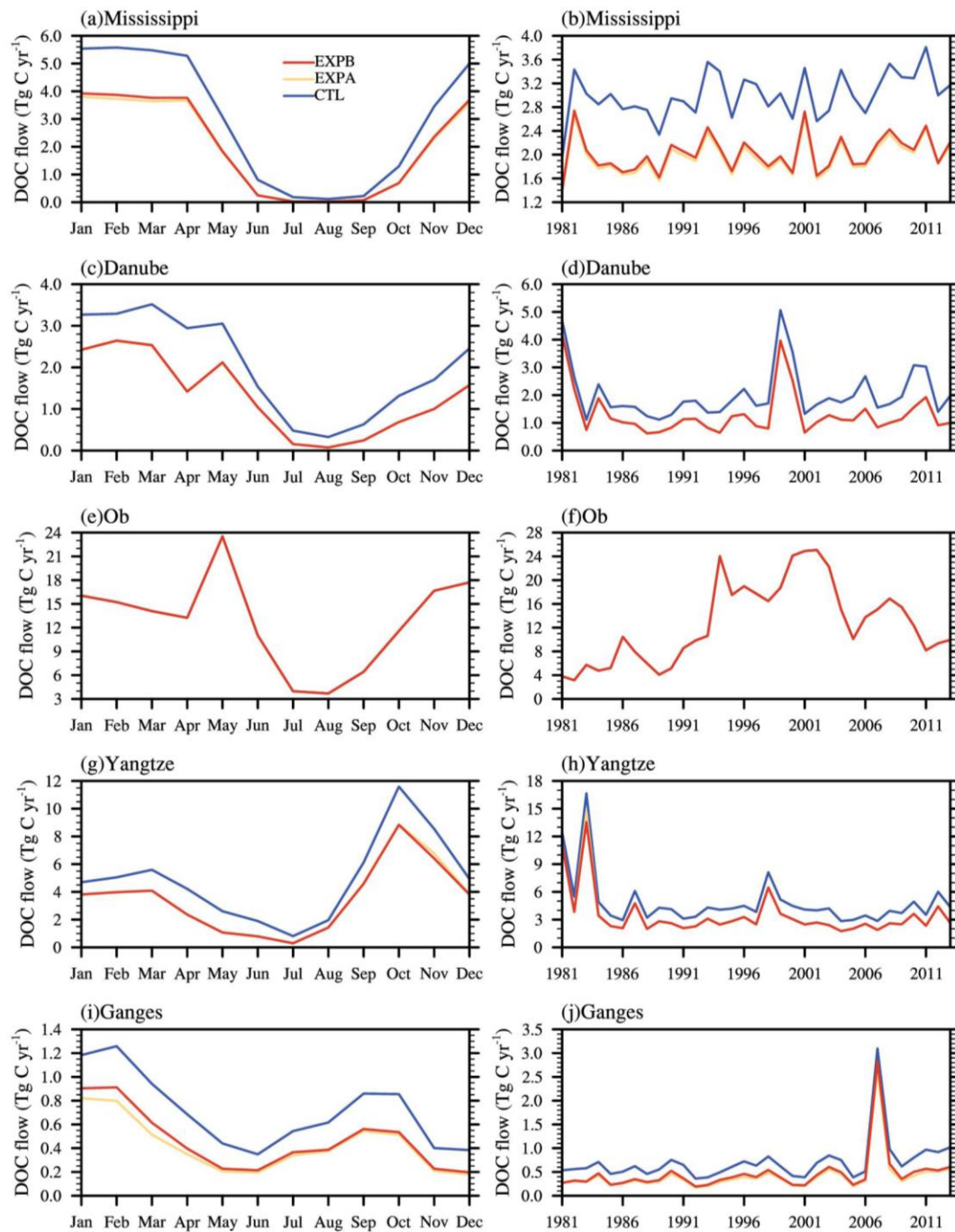


Figure 10. Time series of (a, c, e, g, i) monthly and (b, d, f, h, j) annual average riverine DOC flow rates for the five typical rivers simulated by CTL (blue line), EXPA (yellow line), and EXPB (red line): (a, b) Mississippi River (32.25° N, 91.25° W), (c, d) Danube River (45.25° N, 28.75° E), (e, f) Ob River (66.25° N, 66.75° E), (g, h) Yangtze River (30.75° N, 117.75° E), (i, j) Ganges River (24.25° N, 88.25° E).

373 Riverine DOC export fluxes have obvious spatial heterogeneity. Six zones were defined according to
 374 the latitudes where the river mouths are located, and the effects of the presence or absence of
 375 anthropogenic water regulation on DOC export fluxes are shown in Fig. 11. The hotspot regions of

376 riverine DOC export are concentrated in the tropics (23.5° S–23.5° N) and the mid and high latitudes of
 377 the Northern Hemisphere (40–90° N). The DOC export fluxes of rivers between 40° N and 66° N
 378 accounted for 35.32 % of total global export flux. Due to anthropogenic water regulation, the global DOC
 379 export flux was reduced by $13.36 \pm 2.45 \text{ Tg C yr}^{-1}$ compared to the case with no human regulation, with
 380 the greatest impact concentrated in the subtropical and temperate regions of the Northern Hemisphere
 381 (23.5–66° N) because this is the region with the highest intensity of human water use activity.

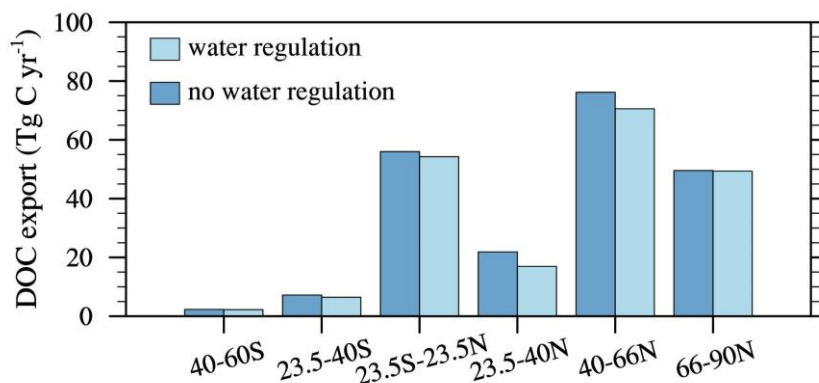


Figure 11. Bar chart of latitudinal band distribution of multi-year average DOC export fluxes from 1981 to 2013. Dark blue indicates no water regulation, and light blue indicates anthropogenic water regulation.

382 Overall, anthropogenic water regulation reduced global riverine carbon fluxes, and the reduction in
 383 DOC fluxes also intensified over time, from $-9.13 \text{ Tg C yr}^{-1}$ to $-16.45 \text{ Tg C yr}^{-1}$ (Fig. 12), and the
 384 reduction percentage also increased from 4.83 % to 6.20 %. Rivers in the Pacific and Atlantic regions
 385 were more affected by water regulation, and the interannual changes were more consistent with the global
 386 picture. The flux of rivers into the Indian Ocean, which was reduced by water regulation, was about 1.27
 387 $\pm 0.23 \text{ Tg C yr}^{-1}$, which was small compared to the global flux, and the flux into the Arctic Ocean was
 388 almost negligible due to the scarcity of human activities.

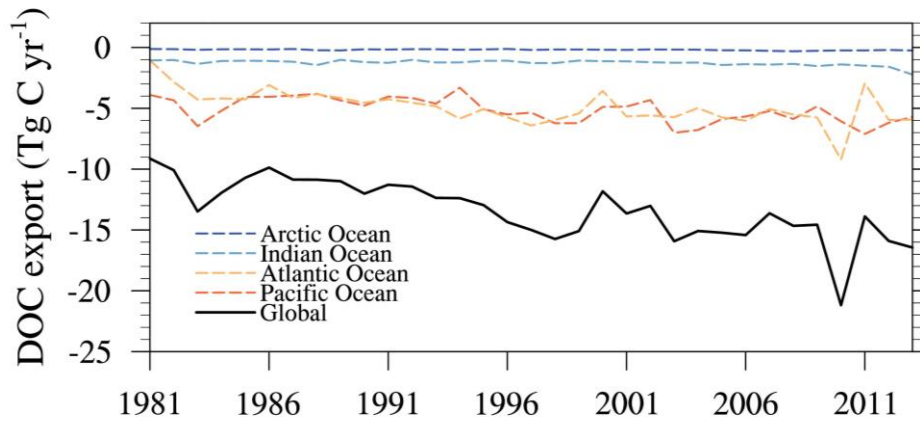


Figure 12. Interannual variability in the impact of anthropogenic water regulation on riverine DOC delivery from rivers to the ocean.

389 **5. Conclusions**

390 This study has developed schemes that consider soil and riverine DOC dynamics and anthropogenic
 391 water regulation activities and has incorporated them into the land surface model CLM5.0. The simulated
 392 river discharges and riverine DOC export fluxes were in good agreement with observations obtained for
 393 106 major world rivers. Surface water and groundwater use datasets were used as inputs to the model,
 394 and three sets of numerical simulations were conducted from 1981 to 2013 on a global scale to investigate
 395 the effects of anthropogenic water regulation on riverine DOC transport.

396 The main conclusions of this study are as follows. First, anthropogenic water regulation activities
 397 increased soil losses in most arid and semi-arid regions of the world, although groundwater extraction
 398 reduced subsurface runoff and decreased DOC leaching; however, this decrease was less than the increase
 399 in DOC runoff due to irrigation. Second, the DOC export fluxes of the Yangtze, Yellow, Mississippi, and
 400 Ganges River basins were significantly reduced by reservoir regulation and surface water withdrawal.
 401 However, DOC export fluxes in these areas showed an increase under groundwater regulation, but the
 402 increase was small, indicating that DOC transport in most rivers globally is mainly influenced by
 403 reservoir interception and surface water regulation. Third, further analysis showed that subtropical and
 404 temperate rivers with intensive water management regimes were more affected and that DOC flows
 405 decreased substantially in winter and early spring. The retention percentage has been increasing year by
 406 year, up to over 50 %, indicating a clear intensification of human water management activities, especially
 407 along the Danube and Yangtze Rivers. In addition, the greatest impact of anthropogenic water regulation

408 activities was concentrated in the region from 23.5°N to 66°N because this zone contains the highest
409 intensity of human water use activities. Fourth, global riverine DOC flux transport to the ocean decreased
410 by an average of 13.36 ± 2.45 Tg C yr⁻¹ per year due to anthropogenic water regulation activities, and
411 the decrease in DOC flux became more pronounced with time, from -9.13 Tg C yr⁻¹ (4.83 %) in 1981 to
412 -16.45 Tg C yr⁻¹ (6.20 %) in 2013, especially in the Pacific and Atlantic Ocean regions. Meanwhile, the
413 Arctic Ocean region was almost unaffected due to low anthropogenic disturbance. In general, this study
414 has developed an effective scheme to simulate DOC export from terrestrial to aquatic systems, which is
415 important for improving carbon budget estimation and integrated ecosystem management.

416 However, there are still some limitations and uncertainties in the developed model that need to be
417 addressed in the future. In this study, we evaluated global riverine DOC transport using observations
418 from a limited number of river sites in literature records, which may have induced a bias. Besides, the
419 simplification of the carbon dynamics of soils and rivers, the uniform parameters, and input data sets also
420 produce some uncertainties. To advance the current model, more observed data sets and more complex
421 schemes of carbon dynamics are needed. In addition, other human activities, such as fertilization,
422 wastewater discharge, and land use change, have a significant impact on riverine carbon transport
423 (Regnier et al., 2013), and should be considered in our future work.

424

425 **Code and Data Availability.** The observed river discharge and riverine DOC exports data can be available
426 through Dai et al. (2012). The source code of CLM 5.0 is available online
427 (<https://www.cesm.ucar.edu/models/clm>). The FORTRAN code of developed model in this study is
428 available upon request. Please contact Zhenghui Xie at zxie@lasg.iap.ac.cn. The drawing language is the
429 NCL language.

430

431 **Author contributions.** The scientific framing of this paper was developed by YY, ZX, BJ. The model
432 was initiated by YY and YW. The literature review was performed by HY, YT and SC. Analyses and
433 scientific post-processing were performed by LW and RL. All authors discussed the results and
434 contributed to the writing of the paper.

435

436 **Competing interests.** The contact author has declared that neither they nor their co-authors have any
437 competing interests.

438

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