1	Storage, patterns and influencing factors for soil organic carbon in coastal wetlands of							
2	China							
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4	Running title: Controls on SOC stocks in coastal wetlands							
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Abstract: Soil organic carbon (SOC) in coastal wetlands, also known as 'blue C', is an essential 30 component of the global C cycles. To gain a detailed insight into blue C storage and controlling 31 factors, we studied 142 sites across ca. 5000 km of coastal wetlands, covering temperate, 32 subtropical and tropical climates in China. The wetlands represented 6 vegetation types 33 (Phragmites australis, mixed of P. australis and Suaeda, single Suaeda, Spartina alterniflora, 34 mangrove (Kandelia obovata and Avicennia marina), tidal flat) and 3 vegetation types invaded 35 by S. alterniflora (P. australis, K. obovata, A. marina). Our results revealed large spatial 36 heterogeneity in SOC density of the top 1-meter ranging 40–200 Mg C ha⁻¹, with higher values 37 in mid-latitude regions (25-30° N) compared to those in both low- (20° N) and high- latitude 38 (38-40° N) regions. Vegetation type influenced SOC density, with P. australis and S. 39 alterniflora having the largest SOC density, followed by mangrove, mixed P. australis and 40 Suaeda, single Suaeda and tidal flat. SOC density increased by 6.25 Mg ha⁻¹ following S. 41 alterniflora invasion into P. australis community, but decreased by 28.56 and 8.17 Mg ha⁻¹ 42 following invasion into K. obovata and A. marina communities. Based on field measurements 43 and published literature, we calculated a total inventory of 57×10^6 Mg C in the top 1-meter soil 44 45 across China's coastal wetlands. Edaphic variables controlled SOC content, with soil chemical properties explaining the largest variance in SOC content. Climate did not control SOC content, 46 but had a strong interactive effect with edaphic variables. Plant biomass and quality traits were 47 a minor contributor in regulating SOC content, highlighting the importance of quantity and 48 quality of OC inputs and the balance between production and degradation within the coastal 49 wetlands. These findings provide new insights into blue C stabilization mechanisms and 50 sequestration capacity in coastal wetlands. 51

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Keywords: Blue carbon, salt marsh, mangrove, vegetation type, plant invasion, coastal
wetlands, climate change

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59 **1 Introduction**

Global wetlands, represent 6–8% of the world's land surface area, and store approximately 60 20-30% of the terrestrial soil carbon (C) pool (estimated to contain 2500 Pg C) (Lal, 2008; 61 Nahlik and Fennessy, 2016). Specifically, coastal wetlands which occupy only <0.3% of the 62 ocean surface contribute approximately 47% of the total organic carbon (OC) buried in marine 63 sediments, with an estimated OC accumulation rate of 0.07-0.22 Pg C year⁻¹ globally 64 (Hopkinson et al., 2012; Duarte et al., 2013). Coastal wetlands are characterized by high 65 primary productivity, deposition rate, C burial rate and low methane (CH₄) emissions, and thus 66 play an important role in land-ocean ecosystem structure and function (Serrano et al., 2019). 67 Therefore, enhancing the capacity of wetlands to sequester OC is an important component of 68 the global effort to mitigate CO₂ entering the atmosphere contributing to climate change. 69

Although wetlands function as important C sinks, coastal wetlands are less studied than 70 those of uplands (i.e., cropland, grassland, and forest) due to the challenges posed by 71 hydrological conditions (i.e., Yang et al., 2007; Xu et al., 2018). Thus, it is important to study 72 C stocks at different spatial scales to allow the assessment of C sequestration and fixation. 73 74 Given the complexity in the biogeochemistry and ecology of coastal wetlands, there is high variability in the quantity and rate of OC buried in underlying soils, depending on geographic 75 settings, vegetation composition, exotic plant invasion and wetland type (e.g., saltmarsh, 76 estuary) across the coastal wetlands (Atwood et al., 2017; Hayes et al., 2017; Osland et al., 77 2018; Rovai et al., 2018; Xia et al., 2021b). However, most studies on coastal wetland C stocks 78 and sequestration mechanisms have been focused on sites characterized by high C density 79 (Whitaker et al., 2015), a single vegetation type (Gao et al., 2019) or single wetland type (Yan 80 et al., 2008), and sites strongly disturbed by human activities (Kirwan and Megonigal, 2013), 81 or wetlands with Spartina alterniflora invasion (Gao et al., 2020). On the other hand, studies 82 focused on a national scale are based on comprehensive meta-analysis using a dataset compiled 83 from published literature rather than field measurements (Xiao et al., 2019). In addition, how 84 different are the SOC density in coastal wetlands with inland wetlands, is not fully clear. Such 85 comparison would strengthen researchers' cognition of the contribution of coastal wetland C 86 storage. Although these existing works are key to understanding the role of coastal wetland C 87

cycles, it is still difficult to accurately quantify the C stocks and processes controlling these 88 stocks due to the paucity in field data to validate models (i.e., soil physico-chemical properties, 89 climate, and vegetation biomass). In addition, data variability among sites likely resulted from 90 differences in methods/protocols and the limited number of sampling locations (Meng et al., 91 2019; Hinson et al., 2017). The deficiency and regional bias of studies from China's coastal 92 wetlands have limited the reliable estimation of their capability as regional and global C sinks 93 (e.g., Jiao et al., 2018; Liu et al., 2014; Fu et al., 2021). Therefore, it is necessary to measure 94 95 these variabilities in climate, vegetation and soil conditions to quantify soil organic carbon (SOC) stocks and their uncertainties in coastal wetlands across local and global scales. 96

To better contribute to achieving carbon neutralization, understanding how the soil C pools 97 and their specific stability in coastal wetlands respond to environmental conditions are 98 important for climate change mitigation and sustainable wetland management (Davidson and 99 Janssens, 2006; Lal, 2004; Han et al., 2020). The predominant factors driving SOC dynamics 100 include climate, plant biological traits, and soil properties (Dungait et al., 2012; Lehmann and 101 Kleber, 2015; Luo et al., 2017). Firstly, climatic variables such as precipitation and temperature 102 103 are usually considered to be critical (Carvalhais et al., 2014) because of their direct effect on organic C inputs via plant CO₂ assimilation, and output via microbial aerobic and anaerobic 104 respiration. However, the effects of precipitation and temperature on net OC accumulation are 105 highly variable (e.g., Chen et al., 2013; Wang et al., 2020; Luo et al., 2020) with different 106 responses in topsoil and subsoil (Hicks Pries et al., 2017; Melillo et al., 2017; Zhou et al., 2018), 107 including increase, decrease and no change. For example, coastal peat swamp systems are 108 usually C sinks with higher primary productivity and lower CO₂ release because long-term 109 water supersaturation and low temperature in the systems are not favorable for microbial 110 111 activities and subsequent SOC mineralization (Bernal and Mitsch 2012; Friborg et al., 2003).

112 Secondly, plant C input is the determinant factor of soil C stocks although higher biomass 113 production may not necessarily result in higher litter C input (Gao et al., 2019). In addition to 114 biomass or litter quantity, litter quality is another important factor regulating organic C stability 115 and preservation (Osland et al., 2018; Bahadori et al., 2021). The vegetation diversity, including 116 native grasses *Phragmites australis, Suaeda, Acorus calamus*, invasive grasses *Spartina* *alterniflora*, and mangrove *Kandelia obovata*, *Avicennia marina*, *Ceriops tagal*, *Bruguiera sexangular*, can also regulate the quality of C inputs into wetland soils (Yang et al., 2019; Xia
et al., 2021a; Fu et al., 2021). The influence of vegetation communities and in particular changes
in vegetation composition (i.e., native vs. invasive species or grasses vs. mangrove) on SOC
remains relatively unstudied in these wetland systems.

Thirdly, although climatic and biological factors (e.g., plant C inputs) could regulate the 122 magnitude or rate of apparent SOC shifting from one status to another, the final SOC stock 123 capacity of soil is generally controlled by the physicochemical properties of soil (Luo and 124 Raphael, 2020). Increasing evidence has shown that soil geochemistry and physical structure 125 provides a physicochemical barrier to microbial accessibility of SOC (Sun et al., 2019, Duan et 126 al., 2020). The physicochemical environment also regulates the supply of water, oxygen, 127 nutrients, and other resources, which are necessary for microbial communities to utilize SOC, 128 as well as for plant C assimilation and deposition as detritus or rhizodeposits (Luo and Raphael, 129 2020). The stabilization of organic carbon in soil has been shown to vary significantly because 130 of physical stabilization mechanism, chemical stabilization mechanism, and biochemical 131 132 protection mechanism (Feng et al., 2013; Sui et al., 2021). However, large field datasets showing spatial heterogeneity of edaphic variables in coastal wetlands remain scarce, which 133 perpetuates the extensive uncertainties about the patterns of SOC stocks and the main 134 controlling factors. It is thus critical to understand the extent to which soil properties and 135 processes regulate SOC stabilization and stocks in different field scenes. 136

The climatic, biotic, and edaphic factors often interact with each other and collectively 137 regulate SOC dynamics through different processes and mechanisms (Zhang et al., 2002; Yang 138 et al., 2019). However, most previous studies have focused on a single independent factor, with 139 140 few quantitatively analyzing the relative effects on SOC stocks based on large scale field data. Studies focusing on the effects of a single factor on SOC dynamics may lead to uncertainties in 141 the outcomes (Meng et al., 2019; Tangen et al., 2020). In addition to topsoil layers (e.g., 0-30 142 cm), deeper soil layers (below 30 cm) could store more organic C than the topsoil layers (Meng 143 et al., 2019). This motivated us to extend our soil sampling to 100 cm to better understand the 144 deep soil C stocks. Such information can provide new insights into mechanisms underpinning 145

SOC dynamics, which are essential for predicting SOC stocks as well as understanding
feedback loops from global environmental change (Belyea et al., 2004).

148 In the present study, we attempted to sample wetlands across broad regions that include almost all coastal wetlands of China, covering temperate, subtropical and tropical climate zones. 149 Our objectives were to: (1) systematically quantify the SOC content and density in China's 150 coastal wetlands and their interrelationship with the location of wetlands, vegetation type, 151 invasion species, and climatic zone; (2) estimate the total inventory of SOC sequestration and 152 regional variability across China's coastal wetlands and compare SOC density with inland 153 wetlands; (3) explore climatic, biological, and edaphic drivers of the distribution and dynamics 154 of SOC contents and stocks. We hypothesized that (1) SOC contents and densities were largely 155 affected by vegetation type, invasion species, and climatic factors; (2) SOC densities greatly 156 varied among different coastal wetlands and were lower than that of inland wetlands; and (3) 157 SOC dynamics were interactively controlled by climate, plant biological traits and edaphic 158 variables, and the effects of edaphic variables override climate and plant inputs on SOC stocks. 159 This study provides crucial information for understanding the contribution of coastal wetlands 160 161 in China to the global C cycles, and will allow for the optimization of wetland management strategies and policy decisions at the national scale. 162

163

164 2 Materials and methods

165 2.1 Study areas

We selected 142 representative wetland sampling sites along China's coast line from 166 Liaoning to Hainan Province (108°-122°E, 20°-40°N) covering temperate, subtropical and 167 tropical climate zones (Figure 1). The sampled wetlands include Liao River Delta (LRD), 168 Duliujian River (DLJR), Nandagang wetland (NDG), Yellow River Estuary (YRE), Linhong 169 River Estuary (LRE), Sheyang River Estuary (SRE), Doulong Harbor (DH), Dafeng Wetland 170 (DW), Qidong Wetland (QW), Wanggang wetland (WGW), Yancheng Wetland (YCW), 171 Chongming Island (CI), Hengsha Island (HI), Yueqing Bay (YB), Minjiang River Estuary 172 (MRE), Jiulong River Estuary (JRE), Zhangjiang River Estuary (ZRE), Zhanjing (ZJ), Beihai 173 (BH), Dongzhai Bay (DB), Qinglan Harbor (QH), Xinying Harbor (XH), Sanya Bay (SYB), 174

175 Sibi Bay (SBB), Xinyang Bay (XB), Tielu Harbor (TH), Danzhou (DZ) and Lingao (LG) (Table

176 1). The mean annual precipitation (MAP) and mean annual temperature (MAT) ranged from

177 551.6 to 1871 mm and from 8.40 to 23.4°C, respectively. The dominant native species at these

178 sites include Phragmites australis, Suaeda glauca, Suaeda salsa, Spartina alterniflora,

179 Kandelia obovata, Avicennia marina, Aeluropus littoralis, Sonneratia apetala, Rhizophora

180 *stylosa*, *Aegiceras corniculatum* (Table 1).

182 2.2 Experimental design, vegetation survey and sample collection

Field surveys and soil samples were conducted during the plant's growing seasons from 183 June to August in 2015–2019. To provide the required spatial heterogeneity, we selected 184 representative wetlands in each coastal city or province with several sampling sites in each 185 186 independent wetland (Table 1). Based on the plant biogeography and dominant vegetation, we categorized all the sampling sites into six vegetation types, including P. australis, Suaeda, P. 187 australis + Suaeda, S. alterniflora, mangrove, and tidal flat. Three invasive vegetation types 188 (occurred at sites CI, HI, YB, MRE, JRE, ZRE, ZJ and BH): P. australis invaded by S. 189 alterniflora, K. obovata invaded by S. alterniflora, A. marina invaded by S. alterniflora. Using 190 historical records on the native species composition at each site, we have established that 191 invasive S. alterniflora completely outcompeted native P. australis. For the native mangrove 192 sites, the invasion had begun along the margin and sparsely populated zones within the 193 mangrove community with the invasion spreading. The invasion by S. alterniflora occurred 194 from 7 to 15 years ago, while the native species had been growing at these sites for more than 195 30 years (Wang et al., 2019). Specifically, the sites with native species were relatively close to 196 their corresponding invaded sites, and we considered SOC contents and soil properties are 197 198 similar before S. alterniflora invasion in paired sites. The vegetation biomasses are shown in Table S1 (our field survey), Table S2 (Yang et al., 2016) and Table S3 (Wang et al., 2019). 199

Each independent soil was homogenized from three subsamples (within a distance of 20– 50 m), and we dispersedly collected several samples from each independent wetland. Samples (from sites LRD, DLJR, NDG and YRE) were collected with an auger at the depths of 0–10, 10–20, 20–30, 30–40, 40–60, 60–80, and 80–100 cm, including a total of 28 sites with 196

¹⁸¹

samples, while samples (CI, HI, YB, MRE, JRE, ZRE, ZJ and BH) were collected at the depths 204 of 0-10, 10-20, 20-30, and 30-40 cm, including a total of 54 sites with 216 samples. The SOC 205 contents and densities at 60 sites in Jiangsu Province (LRE, SRE, DH, DW, QW, WGW, YCW) 206 and Hainan Province (DB, QH, XH, QH, SYB, SBB, XB, TH, DZ, LG) were obtained from 207 published literature. Together, a total of 142 sites were ultimately obtained. Field samples were 208 divided into two parts. One part was freeze-dried for measuring microbial biomass carbon 209 (MBC), dissolved organic carbon (DOC) and lignin phenols. The other part was air-dried at 210 room temperature in the shade. After removal of the visible stones and root residues, the air-211 dried soils were gently ground into fine powder, and sieved through a 2 mm and 0.15 mm 212 stainless steel screen to measure soil physico-chemical properties. 213

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215 2.3 Measurements of soil physico-chemical parameters

216 2.3.1 Analysis of soil physical parameters

Soil bulk density (BD, g cm⁻³) was measured using the cutting ring method (100 cm⁻³). Soil water content (SWC, %) was determined by drying soil samples at 105°C to constant weight. Soil grain size, defined as clay (< 2 μ m), silt (2–20 μ m) and sand (>20 μ m), were measured by a Mastersizer (3000, Malvern Instruments, Malvern, UK) (Lu, 2000).

221 2.3.2 Analysis of soil chemical parameters

Soil pH and electrical conductance (EC, μ S cm⁻¹) were determined using a pH meter and 222 DDS-307 conductivity analyzer (at a soil to water ratio (w:v) of 1:5; Leici company, Shanghai) 223 (Lu, 2000). DOC (mg kg⁻¹) was extracted with deionized water (w:v = 1:5 ratio), and MBC (mg 224 kg⁻¹) was extracted using the chloroform-fumigation method (Vance et al., 1987), and then 225 measured using a TOC-V_{CPH} analyzer (Shimadzu, Japan). Labile organic carbon (LOC, g kg⁻¹) 226 was extracted with 2.5 M H₂SO₄ and measured by the colorimetric method using a UV-Vis 227 spectrophotometer (UV-2600, Daojin company, Japan; Rovira and Ramón Vallejo, 2007). For 228 total phosphorus (TP, g kg⁻¹) analyses, about 0.03 g soil was melted with lithium metaborate at 229 950°C for 0.5 h, then dissolved by 4% (v/v) HNO₃, and measured by the colorimetric method 230 using a UV-Vis spectrophotometer (Ru et al., 2018). 231

For SOC determination, 0.50 g soil was acidified with 20 mL HCl (1.0 M) for 24 h to

remove carbonates, and then washed 3–4 times with distilled water until neutral condition. The
total OC and total nitrogen (TN) of the samples were measured using an Elementar Vario EL
III (Elementar Analysensysteme, GmnH, Germany).

236 2.3.3 Analysis of plant-derived lignin phenols

For the determination of plant-derived lignin phenols in soil, 1.00 g soil, 1.00 g copper 237 oxide (CuO), 0.10 g ammonium iron (II) sulfate [Fe(NH4)2(SO4)2.6H2O] and 15 mL of nitrogen 238 (N₂)-purged NaOH (2 mmol L⁻¹) were combined in Teflon-lined bombs. The bombs were 239 flushed with N₂ for about 10 min and heated to 150°C for 2.5 h in an oven. The lignin oxidation 240 products (LOPs) were derivatized with N, O-bis-(trimethylsilyl) trifluoroacetamide (BSTFA) 241 and pyridine at 70°C for 3 h to yield trimethylsilyl (TMS) derivatives, and were quantified using 242 internal standards (i.e., trans-Cinnamic acid) on an Agilent 7890B-7010B TQ GC-MS system 243 (Agilent, USA), with separation of derivatized lignin phenols using a DB-5MS column (30 m 244 \times 0.25 m \times 0.25 µm) (Xia et al., 2021a). Vanillyl, syringyl and cinnamyl (VSC) phenols were 245 summed to represent lignin in soils, and lignin content (Λ_8) was normalized to the OC content 246 to reflect its relative abundance. 247

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249 2.4 Statistical calculation and analysis

Data were checked for homogeneity of variance and normality before comprehensive 250 analysis. If not, we then performed logarithmic transformation on the data. To utilize and take 251 advantage of all SOC measurements, we used a generalized boosted linear model to obtain the 252 missing data of individual soil properties for a few samples based on the "mice" package by R253 software (www.r-project.org). Data were compared the statistical significance (p < 0.05) using 254 a one-way analysis of variance (ANOVA). The heatmap of Pearson correlation coefficients was 255 conducted by "ggplot2" package within R software. Regression prediction analysis was 256 conducted on the "randomForest" package based on a classification tree to explore the relative 257 influence on SOC contents and density. Data were graphed using Origin 10.0. Variance 258 partitioning analysis (VPA) was conducted using R software with the 'vegan' package to 259 quantify the explanations of categories of different factors for SOC content and their 260 interactions. The stepwise multiple regression (SMR) model was conducted by SPSS 21.0 to 261

assess the variances in SOC content and density explained by different environmental factors.

In this study, we sampled a total of 88 soil profiles, 28 profiles with depths of 0–100 cm 263 (LRD, DLJR, NDG and YRE) and 60 profiles with depths of 0-40 cm (CI, HI, YB, MRE, JRE, 264 ZRE, ZJ and BH). For the 28 profiles with depths of 0-100 cm, we randomly selected 14 265 profiles as group 1 from the given 28 profiles to build a linear regression equation between the 266 measured SOC density in the 0-40 cm layer and measured SOC density in the 40-60 cm layer 267 (p < 0.001; Figure S1A). According to the modeled equation obtained from group 1, we used 268 269 the measured SOC density in the 0-40 cm layer (group 2) to calculate the predicted SOC density in the 40-60 cm layer of group 2. The differences between the measured density in the 40-60 270 cm layer (group 2) and the predicted SOC density in the same layer (group 2) were tested using 271 a paired-sample *t*-test (p < 0.001; Figure S1B) (Li et al., 2019). Therefore, we consider that the 272 method of the prediction model is reasonable and effective. 273

Given that, we then established a linear regression equation between the measured SOC 274 density in the 0-40 cm layer and measured SOC density in the 40-60 cm layer based on 28 275 profiles (group 1 + group 2; Figure S1C). The measured SOC density in the 0-40 cm layer 276 277 (including sites CI, HI, YB, MRE, JRE, ZRE, ZJ and BH) was used to predict the unknown SOC density in the 40-60 cm layer (CI, HI, YB, MRE, JRE, ZRE, ZJ and BH) based on the 278 prediction model (y = 0.35x + 3.35; $R^2 = 0.73$, p < 0.001). Similarly, we used the measured 279 SOC density in the 0-60 cm and 0-80 cm layers to calculate the predicted SOC density in the 280 60–80 cm (Figure S2) and 80–100 cm (Figure S3) layers, respectively (p < 0.001). Thus, the 28 281 soil profiles were used to predict the SOC density in the 40–100 cm layer of 60 soil profiles. 282 For the obtained SOC contents from published literature, we firstly used the formula (Figure 2) 283 to get soil bulk density and then to calculate the SOC density. Similarly, we used the known 284 285 SOC density to predict the unknown SOC density with corresponding soil layers based on the prediction method. 286

Calculations of SOC density at national scales in coastal wetlands are obtained primarily in line with four patterns: independent wetlands, vegetation types, invasion types, and climate zones (temperate, subtropical, tropical). We calculated SOC density (SOC_D) for each soil layer as follows (Xiao et al., 2019):

291
$$SOC_D = \frac{SOC \times BD \times D}{100} \times 10$$
(1)

where SOC_D is SOC density (Mg C ha⁻¹), SOC represents SOC content (g kg⁻¹), BD is bulk density (g cm⁻³), and D is soil thickness (cm).

294 The current SOC sequestration stock was calculated with the following equation:

$$SOCS = Area \times SOC_D$$
(2)

where SOCS, Area (Table 1) and SOC_D are SOC sequestration, soil area and SOC density, respectively.

298 The total inventory of SOC was calculated as follows:

299
$$TI_{soc} = SOCS_{p1} + SOCS_{p2} + \dots \dots SOCS_{pn}$$

300 where TI_{SOC} represents a total inventory of SOC in the top 1-meter across China's coastal 301 wetlands, $SOCS_{pn}$ represents the current SOC sequestration stock in each coastal province of 302 China.

303

304 3 Results

305 3.1 Relationship between bulk density and soil organic carbon

We established a database with 412 measurements of SOC content and paired bulk density. SOC contents ranged from 0.63 to 36.7 g kg⁻¹, and bulk density was in the range of 0.45–1.87 g cm⁻³, demonstrating large variability of organic C and physical properties in the coastal wetlands. There was an empirical relationship between measured SOC content and bulk density in the form of $y = -0.47 \ln(x) + 2.24$ ($R^2 = 0.72$; p < 0.001) (Figure 2).

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312 3.2 Vertical and geographic SOC content distributions

The large standard deviations for vegetation types reflect the wide distribution of vegetation and diversity of wetlands in coastal regions. SOC contents overall decreased with soil depth within the 0–40 cm layer (Figure 3). The wetlands had large spatial differences in SOC contents, with the highest SOC content in the Minjiang River Estuary (23.0 ± 3.94 g kg⁻¹), and the lowest SOC content in the Yellow River Estuary (3.93 ± 1.94 g kg⁻¹). Different sampling sites within the same wetland also displayed some differences in SOC contents (Figure 3A). When grouped by vegetation types, *S. alterniflora* (16.6 ± 4.86 g kg⁻¹) had the highest SOC content, and tidal flat $(4.03 \pm 1.76 \text{ g kg}^{-1})$ had the lowest SOC content (Figure 3B). When grouped by invasion types, the average SOC content in the native *P. australis* (Δ SOC = -0.51 g kg⁻¹; p > 0.05), *K. obovata* (Δ SOC = -2.76 g kg⁻¹; p < 0.01) and *A. marina* (Δ SOC = -0.87 g kg⁻¹; p < 0.01) communities decreased sharply following *S. alterniflora* invasion (Figure 3C). The measured SOC content was highest in subtropical zones (17.4–20.2 g kg⁻¹), which are 38.5–68.9% and 96.2–155% higher than those in tropical (10.3–14.6 g kg⁻¹) and temperate zones (6.85–10.3 g kg⁻¹), respectively (Figure 3D).

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328 3.3 Comparisons in SOC storage between wetlands or vegetation types

SOC density varied broadly within soil profiles depending on the independent wetland, vegetation type, invasion type, and climatic zone. Different spatial patterns of SOC density for independent wetlands were observed between 20°N and 40°N, with relatively higher SOC densities in mid-latitude regions compared to high- and low- latitude regions. The Yellow River Estuary had the lowest SOC densities (28.5–51.7 Mg C ha⁻¹), whereas the Hengsha Island had the highest SOC density (176–202 Mg C ha⁻¹) (Figure 4A).

When grouped by vegetation type, soils with *P. australis* community had the highest OC 335 density (127 Mg C ha⁻¹), being about two times higher than tidal flat (59.5 Mg C ha⁻¹), followed 336 by S. alterniflora > mangrove > P. australis + Suaeda > Suaeda > tidal flat (Figure 4B). The 337 average SOC storage in the native P. australis community slightly increased following S. 338 alterniflora invasion, but in the mangrove community, it markedly decreased compared to the 339 corresponding S. alterniflora community, especially in the K. obovata community (Figure 4C). 340 We found that wetland SOC density was much higher in subtropical zones than those in the 341 temperate and tropical zones, while the SOC density was similar between the temperate and 342 343 tropical zones (Figure 4D).

Across the soil profiles, SOC density decreased with depth in the 0–40 cm layer, but displayed no significant decrease in the 40–100 cm layer. This suggests that SOC is more reactive or labile in topsoil than in subsoil/deeper soil, probably related to the root system distribution and rhizodeposits in soil profile. In the past, the IPCC (2003) recommended a soil depth of 30 cm for the assessment of SOC density in response to global climate change. However, we found that only 40% of the organic C was stored in the top 30 cm, and the majority of the soil C stocks was partitioned in the 30–100 cm layer regardless of the location of wetland, types of vegetation, vegetation invasion, and climate-zone (Figure 4), suggesting that the deeper OC should be also considered in OC stock assessment.

353

354 3.4 Correlations between environmental variables and SOC content under different scenarios

For independent wetlands, the Pearson correlations showed that SOC contents were largely 355 356 different and were significantly correlated to various edaphic variables (Figure 5A), suggesting a large spatial heterogeneity of soil properties for the location of each coastal wetland. For the 357 three climatic zones, the significant differences between SOC content and edaphic variables 358 were BD, clay and MBC (Figure 5B). Both TN and BD were the common factors affecting 359 SOC content for soils with different vegetation, while the other soil parameters were a variant 360 for each vegetation type (Figure 5C). We found that the correlations between SOC content and 361 edaphic variables were strongly altered following S. alterniflora invasion of native species (P. 362 australis, K. obovata, A. marina). For example, the particle size composition (clay, silt and sand) 363 364 was significantly correlated to SOC content for soils with S. alterniflora vegetation, but not for *P. australis* vegetation (Figure 5D). The correlations between climatic factors (MAT and MAP) 365 and SOC density was best described using a cubic function, and the relationships in 0-40 cm 366 layer were more prominent than in 40–100 layer (Figure 6). 367

368

369 3.5 Environmental controls on SOC content and density

We conducted variance partitioning analysis (VPA) using two categories (soil properties and climate), soil properties solely explained 29% of the variance, and climate only explained 3% of the variance for SOC content, while soil properties and climate had the largest interactive effects (42%; Figure 7A). Then, the results of the VPA using three categories (soil chemical properties, physical properties and climate) showed that soil chemical properties were the most important variable explaining SOC content (59% of the variation), while soil physical properties (15%) only accounted for a small proportion (Figure 7B).

377 In determining the relative importance of soil factors on SOC content and density, random

378 forest analysis was carried out and the results demonstrated edaphic variables were in the order

379 $TN > BD > TP > (Ad/Al)_V > MBC > pH > SWC > EC > DOC > clay > silt > sand > S/V > COC > clay > silt > sand > S/V > S/V > COC > clay > silt > sand > S/V > S/V > COC > clay > silt > sand > S/V > S/$

 $(Ad/Al)_{s}$ (p < 0.05; Figure 8A), while for SOC density followed the order of TN > EC > SWC >

381 MBC > TP > DOC > pH > sand > silt > sand > (Ad/Al)_V (p < 0.05; Figure 8B). However, Λ_8

- (reflecting plant C inputs into soil) was not a significant factor influencing SOC content anddensity.
- Stepwise multiple regression analysis was used to assess the combined effects of soil 384 physical, chemical properties, biological traits (i.e., plant-derived lignin phenols) and climatic 385 factors on SOC content. The extracted factors explained more than 90% of the variation in the 386 SOC content. For the extracted parameters based on these analyses of different classification 387 basis, soil chemical properties were the dominant factor controlling SOC content, followed by 388 soil physical properties. Biological traits were a minor contributor to regulating SOC content, 389 and MAP and MAT only exerted effects under specific regions with a large latitude span (Table 390 391 2).
- 392

393 4 Discussion

394 4.1 SOC inventories in China's coastal wetlands

Our dataset here represents the most comprehensive assessment of measured organic C 395 density and sequestration inventory of coastal wetlands in China. Our study utilized 100 cm as 396 an ideal sampling depth to identify the SOC density and sequestration stock (Howard et al., 397 2017), thereby capturing both the more dynamic organic C in the surface soil and the more 398 stable C stocks in the subsoil. Previous studies have demonstrated that SOC density varied 399 widely depending on wetland types (Xiao et al., 2019) and locations (Nahlik and Fennessy, 400 401 2016), with higher C stocks under lower temperature and anaerobic conditions (Lee et al., 2018). As such, we have carried out a more detailed division of coastal wetlands, including the location 402 of independent wetland, vegetation composition, invasion type and climatic zone, to more 403 accurately quantify the C stocks in coastal wetlands of China. 404

We found a wide range of SOC densities, ranging from about 40 to 200 Mg C ha⁻¹. VPA results likely showed that the diversity of precipitation and temperature in different climatic

regions (temperate, subtropical and tropical) contributed to these substantial variations on long 407 time-scales (Figure 7; Osland et al., 2018). The results showed that the Yellow River Estuary 408 had the lowest SOC content and density across the national scale. This system had the highest 409 sand content reinforcing the notion that grain size plays an important role 410 in aggregate stabilization and C sequestration capacity in global coastal wetlands (Yu et al., 411 2021). The differing contribution of autochthonous vs. allochthonous C inputs is also likely to 412 vary spatially because of differences in local hydrological conditions, wetland management and 413 vegetation in the catchment (Saintilan et al., 2013). Autochthonous inputs mainly include plant 414 aboveground litter, root residues and their secretions, primary and secondary products of 415 phytoplankton and benthos; while allochthonous inputs mainly include particulate organic 416 carbon (POC) and dissolved organic carbon (DOC) carried by the processes of tidal inundation, 417 surface runoff and groundwater (Saintilan et al., 2013; Xia et al., 2021a). This highlights the 418 importance of both broad and small spatial-scale data in understanding differences in C stocks 419 under natural scenarios and estimating regional wetland C stocks. The soil C densities of coastal 420 wetlands in China were basically close to the values (93.7 Mg ha⁻¹) reported by Xiao et al. 421 422 (2019), but lower than those reported for coastal wetlands in USA (300 Mg ha⁻¹) (Nahlik and Fennessy, 2016) are higher than ours. 423

424

425 4.2 Factors affecting SOC distribution patterns

426 4.2.1 Climatic influences

Temperature and moisture control net primary productivity which influences inputs 427 (detritus and rhizodeposits) of OC, as well as SOC decomposition (Jobbágy and Jackson, 2010). 428 Hilasvuori et al. (2013) have reported that temperature is the primary factor affecting the 429 accumulation and decomposition of SOC in wetlands of Central Finland. Increased temperature 430 stimulates the loss of organic C pools, especially in high latitudes (Clair et al., 2002; Inglett et 431 al., 2012), thus stressing the risk of SOC loss under a warming climate (Bond-Lamberty and 432 Thomson, 2010). SOC density in tropical zones (i.e., Zhanjiang and Beihai) was generally lower 433 than that in subtropical areas (Figure 4D), most likely due to higher rates of SOC decomposition 434 in the former. The difference of SOC density of each layer in subsoil (40-100 cm) is much 435

lower than that in topsoil (0-40 cm), we assumed that SOC density in the subsoil is relatively 436 more stable compared to topsoil across the three climate zones. However, Li et al. (2020) 437 demonstrated that SOC in subsoil in forest ecosystems is likely to be more vulnerable than in 438 topsoil under rising temperatures, and this phenomenon was primarily controlled by climate 439 and soil C quality. We consider that changing climate will have different effects on SOC stocks 440 based on whether the soils are from upland and wetland. The availability of water (i.e., high 441 water content or shallow water table) in soils can be strongly affected by hydrological dynamics 442 and soil porosity, which may largely restrict oxygen availability for microbes to utilize SOC 443 and soil thermal regimes (Luo et al., 2020). As such, the response of the soil C pool to 444 temperature in wetlands is also influenced by abiotic processes of soil itself, including water 445 availability, nutrient input, and oxygen supply (Olefeldt and Roulet, 2012). The VPA showed 446 soil properties and climate had the largest interactive effects (42%; Figure 7A). This further 447 revealed that other physical and chemical factors, such as temperature, could also affect the 448 mineralization of OC pools by microorganisms (Villa and Bernal, 2018). 449

High precipitation and temperature are usually coupled with high plant productivity (Beer 450 451 et al., 2010), thus influencing plant C inputs into soil, which interacts with inundation periods and depths. In addition, climate can have a direct effect on soil texture, chemical properties and 452 mineralogy, which are closely linked with SOC turnover (Luo et al., 2017). A recent study 453 focusing on the main driving factors controlling C cycling suggested that soil properties (e.g., 454 soil clay content and C:N ratio) rather than climate control the vertical variations of SOC, 455 microbial biomass carbon, and microbial metabolic quotient (Sun et al., 2020). In other words, 456 the sensitivity of SOC dynamics to climate variability may be buffered by changes in primary 457 productivity and soil properties. Our results demonstrate the importance of the interactions 458 between climate, soil, and the amounts and quality of C input (Figure 7; Figure S4). Because 459 of the complex processes involved in forming organic C under different climate conditions, 460 further studies on the trade-off or net effect between primary productivity and soil C 461 decomposition are justified (Bradford et al., 2016). However, to fully consider the influence of 462 all factors on SOC dynamics, previous studies often incorporate climate factors with plant 463 biomass as well as soil properties into one model, which will overestimate the impact of climate 464

on SOC dynamics. This clearly needs further attention and warrants in-depth studies (i.e.,
geological and tidal information) in the future.

467

468 4.2.2 Vegetation composition and exotic plant invasion

The quantity and quality of C inputs are strongly affected by vegetation type, which is 469 predominantly controlled by climatic conditions and interacts with soil conditions (Beer et al., 470 2010; Luo et al., 2017). Different plant communities can influence OC sequestration rates 471 within each vegetation type (Mitsch et al., 2013; Villa and Mitsch, 2015). Researchers often 472 take it for granted that SOC stocks are high under vegetation communities with high net primary 473 productivity (NPP). While it is established that NPP influences C inputs into soil, its effect on 474 SOC stock, particularly in deeper soil layers, is minimal (Figure 4). In other words, it may 475 largely depend on how much biomass ends up in soils, transformation pathways to SOM, 476 priming effects and transportation to deeper soil layers. Thus, NPP may not be a useful indicator 477 of C inputs, especially in deeper soil layers (Xiao et al., 2019). Subsoils may be subject to 478 greater environmental controls than topsoil, including water logging and anoxic conditions. 479 480 These environmental conditions may result in more complex SOC stabilization processes and a divergent behavior in the decomposer community (Keiluweit et al., 2017), lessening the 481 influence of NPP on SOC stock in deeper soil layers. Therefore, the difference in SOC stocks 482 in subsoil among vegetation types was less than that in topsoil (Figure 4). The relatively minor 483 role of plant C inputs (Λ_8 ; p > 0.05) in determining the spatial distribution of SOC content and 484 density (Figure 8), however, does not signify less important factor for local C management. 485 Under similar climatic and edaphic conditions, as the mixed vegetation of P. australis and 486 Suaeda, Suaeda alone had lower NPP (Figure 9), thus lower inputs of SOC compared to other 487 488 vegetation types.

The influence of plant C inputs is straightforward as C influx to soil directly determines OC content. The quality of C inputs (e.g., lignin, C:N, and lignin: N) influences OC utilization by microorganisms and their utilization strategies (Bending et al., 2002; Cotrufo et al., 2013). This in turn controls the composition, preservation, and distribution of SOC pools and their decomposability as a cohort (Prescott, 2010; Raich and Tufekciogul, 2000). SOC content and

density in woody mangrove (K. obovata and A. marina) sharply decreased following the 494 herbaceous S. alterniflora invasion (Figure 3C & 4C). Litters from above- and below-ground 495 components in mangrove forests tend to have greater recalcitrant C compounds (e.g., lignin, 496 tannins, cutin, suberin, and waxes) and are more difficult to decompose than those from invasive 497 S. alterniflora (Chanda et al., 2015). More importantly, soils or sediments tend to efficiently 498 sequester organic C in mangrove forests, which is attributed to the morphological structure of 499 mangroves and their widespread density and distribution of roots (Krauss et al., 2003). The 500 501 unique tree structure and complex aerial root systems (e.g., prop roots and pneumatophores) across mangrove species could result in greater biomasses than grasses, and these specific root 502 structures are more effective for trapping organic-rich sediments (Kristensen et al., 2008). For 503 two herbaceous plants, SOC density in native P. australis community slightly increased 504 following S. alterniflora invasion (Figure 4C). We mainly attributed this result to the following 505 three reasons: (i) the decomposition rate of S. alterniflora litter, particularly the belowground 506 root residues, was slower than that of P. australis litter due to the lower litter quality (i.e., higher 507 C:N ratio) of S. alterniflora (Liao et al., 2008; Duan et al., 2018); (ii) the biomass of S. 508 509 alterniflora was higher than that of P. australis (Wang et al., 2019); (iii) S. alterniflora grew closer to the coast and estuary in comparison to P. australis. 510

However, the SOC density in wetlands predominantly vegetated by mangroves (e.g., southern China from south-central Zhejiang to Hainan Province) is not the largest despite having the highest biomass and the percent of recalcitrant C components (Xia et al., 2021b). The net SOC density is dependent not only on the apparent NPP and litters (Figure 9), but largely on how much of the apparent NPP and litters eventually enters the soil, and how SOC is eventually preserved and stabilized in environments. In addition, plant/litterfall lateral export or import via tidal water also might be the key process to influencing SOC accumulation.

518

519 4.2.3 Soil properties

520 The stabilization mechanisms of SOC have been intensively discussed with respect to (i) 521 physical protection by aggregates associated with minerals, (ii) the nature of recalcitrant 522 compounds of SOC, and (iii) refractory biological components (such as microbial residual carbon), and cementation of physicochemical function and biological substances (Cui et al.,
2014; Throckmorton et al., 2015; Sarker et al., 2018). These SOC stabilization mechanisms
have been demonstrated in uplands to increase C sequestration (Fujisaki et al., 2018; Wang et
al., 2015; Poeplau et al., 2017; Luo et al., 2017; Sarker et al., 2018). Less attention, however,
has been paid to understanding the mechanisms of soil C stabilization in coastal wetlands at
large regional scale, and investigations on the relative influence of individual soil properties
remain scarce.

VPA showed that soil chemical properties were more important than physical properties 530 in controlling SOC content (Figure 7B). Also, stepwise multiple regression analysis showed 531 that the quality of C fractions and nutrients are the most important factors for SOC content 532 (Table S4 & S5). However, increasing evidence suggests that physical protection plays an 533 important role in the preservation of SOC and chemical make-up is less important in uplands 534 (Ekschmitt et al., 2008; Kleber et al., 2011; Guo et al., 2018). We attributed these differences in 535 preservation to (i) OC with recalcitrant compounds is decomposable in uplands at the time scale 536 from years to decades (Fontaine et al., 2007; Schmidt et al., 2011), and (ii) coastal wetlands are 537 538 in the intertidal zones or permanently flooded environments.

In detail, we found clay is an important parameter influencing SOC content in tidal flats, 539 and this is consistent with studies of Lehmann et al. (2007) and Dungait et al. (2012), showing 540 that clay content is an important factor in controlling SOC dynamics in salt marshes. For the K. 541 obovata community, SOC content significantly decreased following S. alterniflora invasion, 542 and C/V was the most important parameter controlling SOC content (Table S4). High contents 543 of plant-derived lignin phenols were found in mangrove soils (Supplementary original data). 544 Recalcitrant organic C was reported to be a major proportion of SOC in an estuarine ecosystem 545 546 in southern China (Lian et al., 2018). In the A. marina community, silt content is considered as a pivotal parameter, and this is also consistent with the study of Xiong et al. (2018) showing 547 that SOC content was primarily controlled by the proportion of finer soil particles in a mangrove 548 ecosystem. For S. alterniflora community, both (Ad/Al)_V and (Ad/Al)_S were important factors 549 controlling SOC content, and suggested SOC molecular composition exerted important roles in 550 protecting from decomposition. Sun et al. (2019) have reported that the simpler OC molecular 551

structure (higher ratio of alkyl C to O-alkyl C and lower aromaticity) of *S. alterniflora* soils
hindered the accumulation of SOC compared with mangrove. Therefore, intrinsic SOC
stabilization such as chemical protection possibly plays a significant role in controlling SOC
dynamics and turnover.

556 One study reported that organo-mineral association was the major mechanism of SOC 557 stabilization in salt marshes, recalcitrant C (as indicated by relatively high aromaticity and low 558 mineral OC) in mangrove soil contributes to SOC stabilization (Sun et al., 2019). Future studies 559 should strengthen the understanding of the role of mineral (i.e., iron, aluminum, calcium 560 compounds) association in SOC stabilization, and help quantifying the relative importance of 561 mineral association vs. chemical protection in SOC stabilization in coastal ecosystems at large-562 regional scales.

563

4.3 Comparisons of SOC density in coastal wetlands with inland wetlands

Based on the geographic locations, hydrological conditions and salinity, researchers often 565 categorized the wetlands as: river, coastal, lake and marsh (Figure 10; State Forestry 566 567 Administration, 2015). SOC density of these wetland types generally decreased in the order of marsh > lake > river > coastal. Marsh wetlands store the most C, and are mainly distributed in 568 the northeast China and Qing-Tibetan Plateau regions, most likely reflecting high litter 569 inputs/preservation, cool temperatures and relatively anaerobic conditions which minimize C 570 mineralization and favor C accumulation (Lin et al., 2013). Lakes also promote C accumulation 571 because they are mostly closed basins that stabilize C and transport little C from surrounding 572 environments (Euliss et al., 2014; Pennock et al., 2010). By contrast, river wetlands are 'neutral 573 pipes' for C accumulation that merely convey landward C to oceans, and mostly accept 574 particulate organic carbon and dissolved organic carbon from upstream (Cole et al., 2007; 575 Aufdenkampe et al., 2011). Compared to inland wetlands, coastal wetlands are adjacent to the 576 landward and ocean. Plant productivity is usually low under saline conditions and long-term 577 flooded environments with high water-level. Mild climatic conditions enhance the 578 decomposition of OC thus limiting the potential for C storage, because coastal wetlands would 579 be exposed to air at low tide and subjected to alternating wet-dry environment (Moinet et al., 580

2018), even though, they are generally believed to have the highest C store among terrestrial
ecosystem worldwide (Lu et al., 2017). Coastal wetlands are open ecosystems, and the different
roles of inland and coastal wetlands in C accumulation are interesting and need further study.
Wetland types (marsh, lake, coastal, and river) were examined to gain a better understanding of
their roles in contributing to SOC sequestration.

586

587 4.4 Implications for blue C management in coastal wetlands

588 Coastal wetlands are open ecosystems at the land and ocean interface, which are more prone to be influenced by anthropogenic activities, rising sea-level, hydrogeological conditions, 589 and river runoff (Rogers et al., 2014; Meng et al., 2017). The multifunctional and complex roles 590 of organic C in these systems remain largely unstudied, in particular relating to the storage, and 591 factors controlling the storage of blue C. Marsh wetlands are mainly distributed in inland 592 regions (Northeast China and Qing-Tibetan Plateau), and contain the largest C pool. However, 593 the coastal regions are at greater risk of being utilized for agricultural and industrial purposes, 594 thus resulting in accelerated decomposition and release of stored organic C (O'Connor et al., 595 596 2019). Therefore, salt marshes (such as Liao River Delta) are one of the primary focuses for wetland C managements in China. The salinity, redox state and nutrient content (particularly 597 TN and TP) of the wetlands can be affected by climate change and anthropogenic activities 598 including development, tillage, drainage. Treating these variables as a constant may result in 599 poor quantification of SOC stocks (Table S4). Eutrophication and pollution are the main 600 problems threatening ecological health of river and lake wetlands (Beaulieu et al., 2019). 601 Comprehensive protection measures are needed to strengthen supervision and administration 602 of environmental protection in river and lake basins, and to reduce the discharge of pollutants 603 604 from industrial and domestic wastewaters, thus maintaining the environmental status and C budgets of river and lake wetlands (Xiao et al., 2019). Additionally, land use change and 605 intensity (i.e., coastal wetland restoration, harvest of wetland macrophytes, and frequency of 606 tides) should be fully considered, which may significantly affect SOC stabilization processes 607 and thus SOC stocks in managed areas. If these processes are fully addressed, improved 608 predictions as well as increased blue C accumulation are possible, providing significant 609

610 opportunities to contribute to climate change mitigation.

611

612 4.5 Uncertainties in C sequestration and outlooks

While our study provides a vast and comprehensive dataset of SOC stocks and their 613 controlling factors in wetlands across climate gradient, vegetation composition and invasion 614 types, there are still limitations and uncertainties in the datasets and assessment. Firstly, most 615 of China's coastal wetland area data were taken from literature, and current data sources are 616 derived from the national soil census (Zheng et al., 2013), local investigation reports (Ding et 617 al., 2004), and remote sensing data (Ma et al., 2015). The above data used for calculating C 618 sequestration in wetlands are inconsistent in data acquisition. Secondly, although our sampling 619 sites in coastal wetlands were representative and decentralized, more sampling sites could be 620 added to further strengthen the statistical significance. Thirdly, there are diverse types with a 621 large amount of different dominant plants (Figure 10), even within the same wetland in the 622 whole coastal wetlands, the categorization was based on our investigations and subjectivity 623 (Table 1). Fourthly, depths of wetland soils varied significantly and there are several examples 624 625 from our data showing that sampling even to 100 cm may not truly represent the C storage potential of these systems. However, data for soil depths were unavailable in some sites. Due 626 to the lack of these reliable data, we used the existing soil C stocks (0-100 cm) to predict other 627 sites (40-100 cm). The model estimation methods might over- or under- estimate C stocks in 628 China's coastal wetlands. Although our results of C stocks here were obtained based on 629 multisource data, new calculations here are effective to reflect the whole picture of organic C 630 631 distributions at a national scale.

632

633 **5 Conclusions**

634 Coastal wetlands play an important role in C sequestration, and at the same time they are 635 sensitive to climate change and anthropogenic disturbances. Although high precipitation and 636 temperature conditions usually result in high plant productivity and thus higher C inputs into 637 soil, wetland SOC density are not simply correlated with MAT and MAP. Changes in SOC 638 density under vegetation communities were not completely consistent with NPP. This was

attributed to differences in SOC decomposition/preservation, litter inputs, and the quality of 639 SOC. We calculated a total inventory of 57×10^6 Mg C in the top 1-meter soil in coastal wetlands 640 of China. Edaphic variables solely override climate control of SOC content, chemical make-up 641 (i.e., the nature of recalcitrant compounds) plays more important role in the preservation of 642 SOC than physical protection. Soil chemical properties exert strong interactive effects with 643 climate. Plant biological traits were a minor contributor in regulating SOC content. Random 644 forest analysis results showed that the relative importance of predictor variables was in the order 645 of TN > BD > TP > (Ad/Al)v > MBC > pH > SWC > EC > DOC > Clay > Silt > and > S/V > Λ_8 646 for SOC contents, and with the order of TN > EC > SWC > MBC > TP > DOC > pH > Sand >647 Silt > Clay > (Ad/Al)v for SOC density. Knowledge and information gained have important 648 implications for understanding mechanisms in SOC stabilization and sequestration capability, 649 and will aid policy and decisions concerning vegetation cover and environmental management, 650 thus contributing to global efforts to mitigate climate change. 651

652

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659

660 **Conflict of interest**

661 All authors declare no conflict of interests.

662

663 Data availability statement

The data that support this study are available in the Supplementary materials, and the supplementary original data would be open access for researchers in figshare webpage <u>https://figshare.com/articles/dataset/Supplementary_original_data_for_Storage_patterns_and_</u> influencing factors for soil organic carbon in coastal wetlands of China /20180450.

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1020 Figure Captions

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FIGURE 1 Distributions of sampling locations along the temperate-subtropical-tropical climate zone in China's coastal wetlands, including 108 sampling sites from Liaoning Province (the north) to Hainan Province (the far south). Circles represent sites with our concurrent field and lab measurements, triangles denote sites or samples compiled with data from literature, and diamonds represent those with data extrapolated from similar wetlands.

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FIGURE 2 Relationships between soil organic carbon (SOC) content and bulk density (BD) in 1031 China's coastal wetlands. Data presented here included 412 paired SOC content and bulk 1032 density measurements for different layers in filed samples, including 0-100 cm layer in LRD, 1033 DLJR, NDG and YRE, and 0-40 cm layer in CI, HI, YB, MRE, JRE, ZRE, ZJ and BH. The 1034 1035 formula was modeled from the paired data points of all samples in the light blue region. The formulas of different colors were modeled from the paired data points of different soil layers. 1036 A logarithmic line model is used to estimate the bulk density for soils without the measured 1037 1038 data.

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FIGURE 3 Vertical distributions of SOC contents (mean \pm SD) in the top 0–40 cm profile based on sampling location (A), vegetation type (B), invasion type (C; *P. australis* invaded by *S. alterniflora*, *K. obovate* invaded by *S. alterniflora*, *A. marina* invaded by *S. alterniflora*) and climatic zone (D).

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FIGURE 4 Vertical distributions of SOC density (mean \pm SD) in the 0–100 cm profile based on sampling location (A), vegetation type (B), invasion type with invasive species S. alterniflora (C) and climatic zone (D). All the data analyzed were collected from our field survey and lab measurements. FIGURE 5 Pearson correlations between soil parameters and SOC content in independent wetlands (A), climate zones (B), vegetation types (C), and invasion types (D). Abbreviations: SOC, soil organic carbon; TN, total nitrogen; MBC, microbial biomass carbon; DOC, dissolved organic carbon; TP, total phosphorus; EC, electrical conductance; SWC, soil water content; BD, bulk density; S/V, the syringyl-to-vanillyl ratio; C/V, the cinnamyl-to-vanillyl ratio; (Ad/Al)v, the acid-to-aldehyde ratio of vanillyl unit; (Ad/Al)s, the acid-to-aldehyde ratio of syringyl unit; Λ_8 , organic carbon (OC)-normalized concentration of lignin phenols (V+S+C). FIGURE 6 Relationships between MAT (°C) and SOC density in the 0-40 cm (A) and 40-100 cm (B) profiles, and between MAP (mm) and SOC density in the 0-40 cm (C) and 40-100 cm (D) profiles. FIGURE 7 Variance partitioning analysis (VPA) for SOC contents in China's coastal wetlands. Soil chemical properties here include MBC, DOC, TN, TP, pH, and EC; Soil physical properties include SWC, BD, and particle size composition (clay, silt and sand); Climate include MAT and MAP. The numbers in the circles indicate the explained values, residuals represent unexplained value.

1079 FIGURE 8 The relative importance of predictor variables from the regression prediction analysis using random forest of the changes to SOC content (A) and density (B). The relative 1080 influence of three categories of variables are calculated as the sum of the relative importance 1081 of individual variables in each variable group (i.e., soil chemical properties, soil physical 1082 properties, and lignin reflecting litter input and composition). Soil chemical properties here 1083 include DOC, TP, pH, EC and MBC; Soil physical properties include BD, SWC, clay, silt and 1084 sand; and plant traits include lignin content Λ_8 , S/V, C/V, (Ad/Al)v, and (Ad/Al)s. Red columns 1085 1086 indicate that variables have significant differences (p < 0.05) in SOC contents.

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FIGURE 9 Vegetation composition and distributions in terms of tidal actions, and aboveground biomass with SOC density depending on each vegetation type. Green histograms represent the relative size of aboveground biomass of a given vegetation. Blue histograms represent the relative size of SOC density vegetated corresponding plants in the top 0–40 cm, and pink histograms are SOC densities in the 40–100 cm layer.

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FIGURE 10 Schematic showing the distributions of SOC density among different wetland types in the critical wetland zone. There are four wetland types (i.e., lake, river, coastal, and marsh) integrated with our field measurements and regional characteristics. The SOC density in top 1 m of lake, river and marsh wetlands are derived from Xiao et al. (2019).

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(D) P. australis S. alterniflora K. obovate S. alterniflora A. marina S. alterniflora (N=36) (N=36) (N=36) (N=36) (N=36) (N=36)

** ** ** ** ** TN ** * DOC ** ** * ** ** ** * MBC ** ** ** ** TP pН ** ** ** ** ** ** ** EC * ** ** ** ** SWC ** BD * ** ** ** Clay * ** ** ** Silt ** Sand ** ٨8 ** SN ** CN (Ad/Al)v (Ad/Al)s

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Province/ city	Area of coastal wetlands (10 ³ ha) ^a	Soil depth (cm)	SOC stock (Mg ha ⁻¹)	Representative wetlands	SOC sequestration (10 ⁶ Mg)	Main plants	The number of samples (sites × layers)	References
Liaoning	97.47	0–40	54.34±12.51	Liao River Delta (LRD)	5.30±1.22	P. australis, S. glauca, S. salsa,	49	This study
Diaoning		40–100	70.77±14.59		$6.90{\pm}1.42$	Tidal flat	(7×7)	Tills study
Tianiin	18.97	0–40	37.45±11.33	Duliuiian River (DLIR)	0.71 ± 0.21	P. australis, S. glauca, S. salsa,	84	This study
Tunjin		40–100	52.40±13.21	Dunigian River (DEsity)	0.99 ± 0.25	Tidal flat	(12 × 7)	This study
Hebei	18.65	0-40	63.01±9.00	Nandagang wetland (NDG)	$1.18{\pm}0.17$	P. australis, S. glauca, S. salsa,	21	This study
nebel		40–100	67.58±3.11	Tundugung wohund (TDG)	1.26 ± 0.06	Tidal flat	(3 × 7)	This study
~1 I		0–40	22.43±4.74		1.71 ± 0.36	P. australis, S. glauca, S. salsa,	42 (6 × 7)	This study
Shandong	76.34	40–100	21.77±3.67	Yellow River Estuary (YRE)	1.66±0.28	S. alterniflora, Tidal flat		
Jiangsu	52.88	0–100	47.61±21.97	Wanggang Wetland (WGW), Yancheng Wetland (YCW), Dafeng Wetland (DW)	2.52±1.16	P. australis, S. salsa, S. alterniflora, A. littoralis	22 sites	Gao et al., 2016; Liu et al., 2017; Zang, 2019; Yang, 2019; Yang et al., 2016; Fu et al., 2021
Shanghai	71.25	0-40	76.40±12.50	Chongming Island (CI),	5.44±0.89	P. australis,	84	This study
U		40–100	89.85±13.43	Hengsha Island (HI)	6.40±0.96	S. alterniflora	(12×7)	,
	29.51	0-40	38.49±7.93		1.14±0.23	K. obovate.	42	
Zhejiang		40–100	49.16±8.51	Yueqing Bay (YB)	1.45±0.25	S. alterniflora	(6 × 7)	This study
Fujian	34.23	0–40	57.51±9.46	Minjiang River Estuary (MRE), Jiulong River Estuary (JRE), Zhangjiang River Estuary (ZRE)	1.97±0.32	P. australis, S. alterniflora, K. obovate A. maring	168	This study
		40–100	69.57±10.16		2.38±0.35	A. Ooovaic, A. marina,	(27 ~ 7)	
	49.37	0-40	42.94±3.95		2.12±0.20	A. marina,	42	
Guangdong		40-100	53.94±4.24	Zhanjiang (ZJ)	2.66±0.21	S. alterniflora	(6 × 7)	This study
Guangxi	81.88	0-40	40.91±1.86	Beihai (BH)	3.35±0.15	A. marina,	42	This study

Table 1 Summary of SOC sequestration capacity in coastal wetlands of China synthesized from sampled soil cores and literature data

		40–100	51.76 ± 2.00		4.24±0.16	S. alterniflora	(6 × 7)	
Hainan	10.23	0–100	315.21±248.5 8	Dongzhai Bay (DB), Qinglan Harbor (QH), Xinying Harbor (XB), Sanya Bay (SYB), Sibi Bay (SBB), Xinyang Bay (XB), Tielu Harbor (TH), Danzhou (DZ), Lingao (LG)	3.22±2.54	S. apetala, K. obovate, R. stylosa, A. marina, A. corniculatum, B. sexangular, C. tagal, B. gymnorrhiza	38 sites	Gao et al., 2019; Wang et al., 2019; Huang et al., 2017; Xiong et al., 2018; Xin et al., 2014; Lin et al., 2015; Gao et al., 2018; Fu et al., 2021
Hong Kong ^b	0.11	0–100	110.90±5.30 ~ 160.3±21.00	_	0.09±0.01	_	_	Fu et al., 2021
Macao ^b	0.01	0–100	110.90±5.30 ~ 160.30±21.00	_	0.00±0.00	—	—	Fu et al., 2021
Taiwan ^c	20.35	0–100	135.80±25.90 ~ 192.00±23.80	—	0.13±0.02	—	—	Fu et al., 2021

^bSOC stocks Hong Kong and Macao was substituted by the data of Pearl River Estuary;

1267 SOC stocks in Taiwan was substituted by the data of Fujian, Guangdong, Guangxi and Hainan.

Classification basis		Regression equation		<i>p</i> value	Std. error of the estimate	Ν	Extracted parameters
	P. australis	Y=-21.53+7.85TN+0.02MAP+0.01MBC-0.02DOC- 0.15SWC+7.34TP+1.41pH-4.53(Ad/Al) _V	0.98	<0.001	1.31	96	TN, MAP, MBC, DOC, SWC, TP, pH, (Ad/Al) _V
	Suaeda	Y=0.07+7.97TN	0.98	< 0.001	0.62	16	TN
Vegetation types	P. australis, Suaeda	Y=-2.39+11.86TN	0.92	< 0.001	0.61	12	TN
	S. alterniflora	Y=30.20-47.59(Ad/Al) _V +11.13TN-0.68MAT-0.11Λ ₈ - 0.001EC+0.02DOC	0.91	< 0.001	1.64	108	$(Ad/Al)_V$, TN, MAT, Λ_8 , EC, DOC
	Mangrove	Y=24.55-15.22BD-0.001EC+0.02DOC+0.20SWC	0.96	< 0.001	1.08	72	BD, EC, DOC, SWC
	Tidal flat	Y=7.79+7.43TN+0.10Clay-0.70MAT-0.02DOC	0.83	<0.001	0.81	24	TN, Clay, MAT, DOC
	P. australis	Y=19.75+6.97TN-0.22SWC	0.91	< 0.001	1.26	36	TN, SWC
	S. alterniflora	Y=-14.68+11.41TN+1.69pH+0.21Sand+0.02DOC	0.98	< 0.001	0.73	36	TN, pH, Sand, DOC
Invasion	K. obovate	Y=2.97+9.07TN+3.52C/V	0.98	< 0.001	0.61	36	TN, C/V
types	S. alterniflora	$Y=49.14-150.56(Ad/Al)_V+27.25(Ad/Al)_S$	0.89	< 0.001	1.32	36	$(Ad/Al)_V$, $(Ad/Al)_S$
types	A. marina	Y=52.35+0.24SWC-2.21MAT+4.02TP+0.37pH-2.87BD	0.99	<0.001	0.23	36	SWC, MAT, TP, pH, BD
	S. alterniflora	Y=14.75-7.17BD+9.78TP	0.95	< 0.001	1.15	36	BD, TP
Climate zone	Temperate	Y=4.88+8.17TN+0.004MBC-0.07SWC-2.35BD	0.95	< 0.001	1.11	112	TN, BD, MBC, SWC, BD
	Subtropical	Y=5.57+8.78TN+5.70BD+0.005MBC+0.03DOC-0.001EC- 22.18(Ad/Al) _V	0.90	< 0.001	1.47	168	TN, BD, MBC, DOC, EC, (Ad/Al) _V
	Tropical	Y=3.71+4.23TN-4.03BD+0.07SWC+0.64pH	0.99	< 0.001	0.41	48	TN, BD, SWC, pH