Updated estimation of forest biomass carbon pools in China, 1977–2018

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Received: 17 January 2022 – Discussion started: 4 February 2022 Revised: 9 May 2022 – Accepted: 30 May 2022 – Published:

Abstract. China is one of the major forest countries in the world, and the accurate estimation of its forest biomass carbon (C) pool is critical for evaluating the country's C budget and ecosystem services of forests. Although several stud-

- 5 ies have estimated China's forest biomass using national forest inventory data, most of them were limited to the period of 2004–2008. In this study, we extended our estimation to the most recent period of 2014–2018. Using datasets of eight inventory periods from 1977 to 2018 and the continu-
- ¹⁰ ous biomass expansion factor method, we estimated that the total biomass C pool and average biomass C density in Chinese forests increased from $4717 \text{ Tg C} (1 \text{ Tg} = 10^{12} \text{ g})$ in the period of 1977-1981 to 7975 Tg C in the period of 2014-2018 and $38.2 \text{ Mg C ha}^{-1}$ to $45.8 \text{ Mg C ha}^{-1} (1 \text{ Mg} = 10^6 \text{ g})$,
- ¹⁵ respectively, with a net increase of 3258 Tg C and an annual sink of 88.0 Tg C yr⁻¹. Over the most recent 10 years (2009– 2018), the average national forest biomass C density and C sink were 44.6 Mg C ha⁻¹ and 154.8 Tg C yr⁻¹, respectively, much larger than those of 39.6 Mg C ha⁻¹ and 63.3 Tg C yr⁻¹
- ²⁰ in the period 1977–2008. These pronounced increases were largely attributed to afforestation practices, forest growth, and environmental changes. Our results have documented the importance of ecological restoration practices, provided an essential basis for assessing ecosystem services, and helped

25 to achieve China's C neutrality target.

1 Introduction

Terrestrial ecosystems' carbon (C) sinks, which absorb approximately 30 % of annual anthropogenic carbon dioxide (CO₂) emissions, are mostly contributed by forests (Pan et al., 2011; Terrer et al., 2021). Globally, forests cover ~ 30 % ³⁰ of the land area (Fahey et al., 2010; Guo et al., 2013) and sequester large amounts of C in woody biomass and soils (Pugh et al., 2019). Even small changes in the forest C pool could induce profound feedback on the climate system of the planet (He, 2012). Thus, forests play fundamental roles in regulating the global C balance and mitigating climate change (Pan et al., 2011; Harris et al., 2021).

China has a forest area of $\sim 0.2 \times 10^9$ harms, accounting for 5.51 % of the global total forest and ranking fifth among countries (FAO, 2021). The forests in China are distributed 40 over a wide range of climatic conditions and are diverse in forest types, which makes them unique for forest C cycle research (Fang et al., 2010). Meanwhile, China's forests, especially planted forests, are generally young, with low biomass C density, implying a great C sequestration potential in the 45 future (Xu et al., 2010; Zhao et al., 2019). In 2020, the Chinese government announced the goal to achieve C neutrality by 2060, for which sequestering more C through forest C sinks has been regarded as an essential part of China's action plans (Yu et al., 2021). Therefore, accurate estimates of 50 the forest biomass C pool and sink capacity are crucial for China's aim of reducing net greenhouse gas emissions (Xu et al., 2010) and reaching national C neutrality.

However, there are several limitations in the previous studies. On the one hand, most of the estimates of China's forest biomass C pool (and the corresponding C sink) were based on the national forest inventory data no later than the 7th Na-5 tional Forest Inventory (2004–2008) (e.g. Fang et al., 1996, 2001, 2007; Fang and Chen, 2001; Xu et al., 2010; Guo et al., 2013; Zhang et al., 2013; Li et al., 2015; Zhang et al., 2015). Nevertheless, China's large-area planted forests formed by extensive afforestation since the 1980s have just

- ¹⁰ grown into the rapid-growing middle-aged stage in the most recent decade (2009–2018) (Lu et al., 2018), thus clarifying how China's forest C pool has changed since the 2010s is of vital importance to accurately evaluate the C sink formed by large-scale afforestation. Meanwhile, a 4-decade-long es-
- ¹⁵ timate of the forest C pool can also provide necessary information for exploring the driving mechanism of China's forest C sink and validating the conclusions from various models. On the other hand, the estimated C pool and/or sinks of China's forest differed considerably among their dedicated
- ²⁰ periods. For example, the estimate of the forest C sink during 1994–1998 ranged from -2.9 to 108 Tg C yr⁻¹ (Fang et al., 2007; Zhang et al., 2013; Li et al., 2015; Zhao et al., 2019). This large discrepancy could be due to limitations in sample size and data representativeness, diversity of data
- ²⁵ sources, inconsistent biomass–volume fit relationships (Li et al., 2015; Tang et al., 2018), or in particular divergences in the methodologies. Therefore, it is necessary to adopt a unified and well-validated method to update the estimations of the forest biomass C pool at the national scale over the past

³⁰ 4 decades, especially the most recent decade (2009–2018), to fill gaps in our knowledge regarding China's forest C pool and its changes.

In this study, we used eight national forest inventories compiled during the period 1977–2018 and conducted a well-

- ³⁵ validated biomass expansion factor (BEF) method to update the estimate of China's forest biomass C pool. Three major aims of the study are to (1) objectively describe the long-term changes in China's forest biomass C pool and C sink at the national scale, especially focusing on the changes in the most
- ⁴⁰ recent decade (2009–2018); (2) clarify the contributions of different forest zonal types, age groups, and forest stand origins to China's forest biomass C pool and its changes; and (3) qualitatively evaluate the potential influencing factors of China's forest C sink by comparing the changes in biomass
- ⁴⁵ C stocks in the most recent decade with those in the previous 30 years (1977–2008).

2 Methods

2.1 Methods for estimation

Mean biomass density (MBD), remote sensing, and BEF are ⁵⁰ three common methods for estimating large-spatial-scale forest biomass C stocks (Guo et al., 2010; Zhang et al., 2013). The MBD method, defined as multiplying the mean biomass density value of each forest type by its area, is widely used to estimate the biomass C stocks at different spatial scales. However, because of investigators' intentional tendencies 55 to choose better-growing stands during forest censuses, the MBD method usually overestimates the C pool (Fang et al., 2006; Guo et al., 2010). The remote sensing method can provide large-scale information of vegetation, while there are several inherent methodological issues that must be ad-60 dressed, such as atmospheric and background noise, similar spectral characteristics of different vegetation, and saturation of signals in dense vegetation (Zhang et al., 2013). Different from the above two methods, the BEF method is based on the internal relationship between the biomass and timber 65 volume of forests (Fang et al., 2002). By establishing proper regression models between biomass and timber volume, the BEF method can incorporate the effects of forest age, forest stand density, and forest site quality on biomass density, thus achieving appropriate estimates of the forest biomass C 70 pool (Fang et al., 2002; Guo et al., 2010). Thus, the BEF method has obvious advantages in estimating forest biomass at regional and national scales (Fang et al., 1998; Fang and Wang, 2001; Guo et al., 2010; Teobaldelli et al., 2009).

Here, we used the continuous BEF method suggested by ⁷⁵ Fang et al. (2001) to estimate China's forest biomass C stocks during 1977–2018, which is a well-validated approach that enables upscaling estimates from field plots to a regional level. In this approach, the BEF is calculated by the firstorder derivative formula (Eq. 1) using the ground survey data ⁸⁰ of forest volume, and then BEF is used to calculate biomass (Eq. 2):

$$BEF = a + b/x,$$
(1)

$$y = BEF \cdot V, \tag{2}$$

where *x* and *V* are the stock volume density and stock volume of a forest type at a certain age in each province, and *a* and *b* are BEF function coefficients. BEF is the biomass expansion factor, and *y* is the biomass of a forest type at a certain age. The coefficients in Eq. (1) were retrieved from previous studies (Fang et al., 1998, 2002; Fang and Wang, 2001) (Table S1 in the Supplement). A constant C conversion factor of 0.5 was used to convert biomass into C (Fang et al., 2001).

2.2 Data sources

Eight national forest inventory datasets compiled by the ⁹⁵ Chinese Ministry of Forestry Administration from 1977 to 2018 were used in this study. Forests' dominant tree species, area, timber volume, forest age, and stand origins were reported for all provinces. According to Fang (2000), five zonal forest types including cold-temperate coniferous, temperate ¹⁰⁰ coniferous, temperate deciduous broad-leaved, temperate– subtropical mixed, and evergreen broad-leaved are dominant tree species. To quantify age-related tree growth, forests were further divided into five subgroups: young, middle-aged, premature, mature, and overmature. Forest area and C stocks were calculated for each province. Chongqing Municipality, which was separated from Sichuan Province in 1997, was 5 merged into Sichuan here. The detailed data of Taiwan, Hong

Kong, and Macau are missing from the inventory datasets; thus, the calculations did not account for these three regions.

2.3 Data correction

Since 1994, the canopy coverage criterion of forests in the $_{10}$ national forest inventory has been changed from > 0.3 to >

0.2. We unified the criterion by adopting the power functions (Eqs. 3 and 4) provided by Guo et al. (2013):

$$AREA_{0.2} = 1.290 \times AREA_{0.3}^{0.995} (R^2 = 0.996),$$
(3)

$$CARBON_{0.2} = 1.147 \times CARBON_{0.3}^{0.996} (R^2 = 0.996), \quad (4)$$

¹⁵ where AREA_{0.2} and AREA_{0.3} are the forest areas (10^4 ha) with canopy coverages of 0.2 and 0.3, respectively, and CARBON_{0.2} and CARBON_{0.3} are the biomass C pools (Tg C) with crown densities of 0.2 and 0.3, respectively.

3 Results

20 3.1 Forest biomass C pool and its changes

The total forest biomass C stock, average biomass C density, and biomass C sink during 1977–2008 were 5447 Tg C, 39.6 Mg C ha⁻¹, and 63.3 Tg C yr⁻¹, respectively (Table 1). The corresponding values were 7525 Tg C, 44.6 Mg C ha⁻¹, ²⁵ and 154.8 Tg C yr⁻¹, respectively, during 2009–2018, making an increasing C sink of 91.5 Tg C yr⁻¹ (Tables 1 and S3). Compared with the forest biomass C pool during 1977–1981, it increased by 3258 Tg C (69.1 %) during the 4 decades up to 2014–2018. The C density of forest biomass increased ³⁰ by 7.61 Mg C ha⁻¹ (19.9 %) during 1977–2018 (Table 1, Fig. 1). Meanwhile, the forest area increased by 41.0 % from 1.24 × 10⁸ ha during 1977–1981 to 1.74 × 10⁸ ha during 2014–2018 (Table 1). All these changes had led to a large C sink of 180.2 Tg C yr⁻¹ in 2014–2018 (Table 1, Fig. 1).

- ³⁵ In addition, the forest biomass C pool varied considerably across the different periods. It was found to have decreased by 2.9 Tg C yr^{-1} over 1994–1998, which was thought to have been due to the decrease in the area of natural forest from 1994–1998 (Tables 1 and 2, Fig. 1).
- ⁴⁰ The biomass C pools of planted forests and natural forests increased significantly during the study periods (Table 2). The biomass C pool of the planted forest increased from 250 Tg C in 1977–1981 to 1470 Tg C in 2014–2018. This indicated that biomass C sinks had an average in-
- ⁴⁵ crease of 33.0 Tg C yr⁻¹. The biomass C density of planted forests increased from 15.6 Mg C ha⁻¹ during 1977–1981 to 28.3 Mg C ha⁻¹ during 2014–2018. Compared with the previous 3 decades (1977–2008), the average biomass C

Table 1. Forest area, biomass C pool, C density, and C sinks from 1977 to 2018.

Period	Forest parameter						
	Area 10 ⁴ ha	C pool Tg C	C density Mg C ha ⁻¹	C sink Tg C yr ⁻¹			
1977–1981	12350	4717	38.2				
1984-1988	13 169	4885	37.1	23.9			
1989-1993	13971	5402	38.7	103.5			
1994–1998	13 241	5388	40.7	-2.9			
1999-2003	14279	5862	41.1	94.9			
2004-2008	15 559	6427	41.3	112.9			
Average 1977–2008	13762	5447	39.6	63.3			
2009–2013	16349	7074	43.3	129.4			
2014–2018	17 409	7975	45.8	180.2			
Average 2009–2018	16879	7525	44.6	154.8			
Overall change 1977–2018	5059	3258	7.61	88.0			

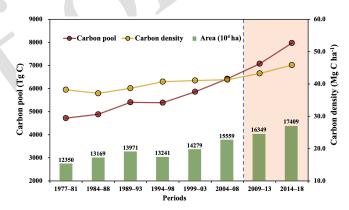


Figure 1. Changes in the area, biomass C pool, and C density of forests from 1977 to 2018. The pink box highlights the results of the recent decade.

density and C sink in the most recent decade (2014–2018) increased by 4.6 Mg C ha^{-1} and $10.0 \text{ Tg C yr}^{-1}$, respectively. For natural forests, the biomass C pool increased in most time steps during the study periods. Especially in the most recent 10 years, the biomass C sink of natural forests has grown rapidly, indicating $100.8 \text{ Tg C yr}^{-1}$ during 2009–2013 and $128.1 \text{ Tg C yr}^{-1}$ during 2014–2018 (Table 2). Compared with the previous 3 decades (1977–2008), the average biomass C density and C sink in the most recent decade (2014–2018) increased by 7.8 Mg Cha⁻¹ and 81.5 Tg C yr⁻¹, respectively. From 1977 to 2018, the increase in the biomass C pool of natural forests was 2037 Tg C, indicating C sinks of 55.1 Tg C yr⁻¹ on average.

Period	Planted forest				Natural forest			
	Area 10 ⁴ ha	C pool Tg C	C density Mg C ha ⁻¹	C sink Tg C yr ⁻¹	Area 10 ⁴ ha	C pool Tg C	C density Mg C ha ⁻¹	C sink Tg C yr ⁻¹
1977–1981	1595	250	15.6		10755	4468	41.5	
1984-1988	2347	418	17.8	24.1	10822	4467	41.3	-0.1
1989-1993	2675	526	19.7	21.6	11 296	4876	43.2	81.9
1994–1998	2914	642	22.0	23.3	10326	4746	46.0	-26.2
1999-2003	3229	836	25.9	38.7	11049	5026	45.5	56.2
2004–2008	4000	1067	26.7	46.2	11 559	5360	46.4	66.7
Average 1977–2008	2794	623	22.3	30.3	10968	4824	44.0	33.0
2009-2013	4665	1183	25.4	23.3	11685	5864	50.2	100.8
2014-2018	5193	1470	28.3	57.3	12212	6505	53.3	128.1
Average 2009–2018	4931	1327	26.9	40.3	11 948	6185	51.8	114.5
Overall change 1977–2018				33.0				55.1

Table 2. Area, biomass C pool, C density, and C sink of planted and natural forests.

3.2 Changes in biomass C pools in different zonal forest types

Compared with 1977–2008, temperate coniferous forest, temperate deciduous broad-leaved forest, and evergreen ⁵ broad-leaved forest all presented larger area during 2009–

2018. With the exception of the temperate deciduous broadleaved forest, the C sinks of four of the five forest types increased in the most recent 10 years (2009–2018) in comparison to 1977–2008 (Table 3). In particular, the biomass

- ¹⁰ C sink of evergreen broad-leaved forests during 2009–2018 was 84.6 Tg C yr⁻¹, which is much higher than the average C sink for the previous 30 years (22.8 Tg C yr⁻¹, 1977–2008). The C pool of the evergreen broad-leaved forest reached 2747 Tg C during 2014–2018 (Tables 3 and S2). Overall
- ¹⁵ changes in the biomass C pools from 1977 to 2018 indicated that, with the exception of cold-temperate coniferous forests, the biomass C pools of four out of the five forest types all increased. The largest increase took place in the evergreen broad-leaved forest (1463 Tg C), which had an average an-²⁰ nual C sink of 39.5 Tg C yr⁻¹ (Table S2). For more details
- about the area, C pool, C density, and C sinks between 1977 and 2018, please refer to Table S2.

3.3 Biomass C sequestration in different forest age groups

²⁵ In the original national forest inventory, only three age groups were recognized, namely young forests, middle-aged forests, and old forests. In the subsequent inventories (after 1984), forests were categorized into five different age groups as we mentioned above (Table S6). To facilitate the comparison of different periods, we grouped the premature forest, mature forest, and overmature forest into one forest age group – old forest. The young and middle-aged forests remained unchanged. The three classes are aggregated from the five classes.

The growth rate by area and C pool of all forest age groups ³⁵ over the most recent 10 years (2009–2018) were higher than those for the previous 30 years (1977–2008). In addition, the forest area and C pool for each age group reached the highest level recorded during 2014–2018. The area and C pool of old forests were both the largest, at 6.33×10^7 ha ⁴⁰ and 4387 Tg C, respectively (Fig. 2). The area of young, middle-aged, and old forests increased from 1977 to 2018 by 1.47×10^7 , 1.08×10^7 , and 2.51×10^7 ha, respectively (Fig. 2, Table S5). Meanwhile, the C pools increased by 657, 791, and 1810 Tg C, respectively (Fig. 2, Table S5). ⁴⁵ The biomass C densities of young, middle-aged, and old forests increased by 7.9, 7.5, and 1.9 Mg C ha⁻¹, respectively (Fig. 2, Table S5).

4 Discussion

4.1 China's forest C sink size

National forest inventories provide the most comprehensive statistical and temporal datasets for investigating forest change. Using data from eight forest inventories including the most recent forest inventory period of 2014–2018, this study provided an updated estimate of China's forest ⁵⁵ biomass C pool and obtained how it has varied over the last 40 years (1977–2018). Our results showed that the lat-

Zonal forest types	Periods	Area 10 ⁴ ha	C pool Tg C	C density Mg C ha ⁻¹	C sink Tg C yr ⁻¹
Cold-temperate coniferous forest	1977–2008	2080	1461	70.2	-8.2
	2009–2018	1844	1321	71.6	11.3
Temperate coniferous forest	1977–2008	1125	381	33.9	9.9
	2009–2018	1740	652	37.5	13.7
Temperate deciduous broad-leaved forest	1977–2008	3614	1430	39.6	26.2
	2009–2018	3714	1730	46.6	19.5
Temperate-subtropical mixed forest	1977–2008	3989	975	24.4	12.6
	2009–2018	3834	1308	34.1	25.7
Evergreen broad-leaved forest	1977–2008	2953	1200	40.7	22.8
	2009–2018	5748	2515	43.7	84.6

Table 3. Area, biomass C pool, C density, and average C sink of different zonal forest types in the most recent decade compared to the previous 30 years^{*}.

* The average C sink for the previous 30 years was calculated by dividing the difference between the 2004–2008 and 1977–1981 C pools by 27. The average C sink over the most recent decade was calculated by dividing the difference between the 2014–2018 and 2004–2008 C pools by 10.

est forest biomass C pool (7975 Tg C, 2014–2018) was much larger than 4717 Tg C from 1977–1981, with an increase of 3258 Tg C (69.1%), and accordingly, the C sink averaged 88.0 Tg C yr⁻¹. In particular, the biomass C pool increased ⁵ by 154.8 Tg C yr⁻¹ during the period of 2009–2018, which was significantly higher than the average C sequestration rate of 63.3 Tg C yr⁻¹ in the previous 3 decades (1977–2008). In summary, China's forests acted as a significant C sink in biomass during 1977–2018, while the C sink was further en-¹⁰ hanced in the most recent decade.

In terms of the order of magnitude, our estimates of biomass C stocks are comparable with those from previous studies (e.g. Fang et al., 2007; Zhang et al., 2013; Li et al., 2015; Zhang et al., 2021; Zhao et al., 2019, 2021), while ¹⁵ there are some discrepancies between the size of the C pool

- in this study and those reported by others. In particular, most studies provided smaller estimates than ours in the periods prior to 1994 (Table S7). This may be due to methodological differences in the conversion of canopy coverage criteria.
- ²⁰ Some studies did not consider the shift from the old criterion (> 0.3 canopy coverage) to the new criterion (> 0.2 canopy coverage), which led to shrinking forest areas and thus significant underestimation of the C pool in the earlier periods (e.g. Li et al., 2015; Zhang et al., 2021). Others used a linear
- ²⁵ model proposed by Fang et al. (2007) to correct the inventory data (e.g. Zhang et al., 2013; Zhao et al., 2019, 2021), while the linear conversion equations would underestimate forest areas and biomass C stocks in the provinces with large amounts of forest (Fig. S1). Obviously, these underestimates
- ³⁰ would inevitably induce previous studies to report higher C sinks than those of this study. Moreover, the divergences in BEF model parameters, root-to-shoot ratio, and C conversion factor would also contribute to the discrepancies among the

estimates of C stocks, thus inducing the different C sequestration rates (e.g. Zhao et al., 2019, 2021).

Despite the differences in the estimates of biomass C stocks and their changes, most previous studies and this work all agreed that the average biomass C sink of China's forests since the 1980s is approximately 100 Tg C yr^{-1} . In addition, several studies based on field investigations pointed out that 40 both the dead organic matter and soils in China's forest also functioned as C sinks over the past 4 decades, with average values of 6.7 and 57.3 Tg C yr⁻¹, respectively (Zhu et al., 2017; Yang et al., 2014; Fang et al., 2018). Together with the biomass C sink estimated in this study, it CEI suggested that 45 China's forest ecosystems would sequester approximately $100-200 \text{ Tg C yr}^{-1}$ in total over the last 4 decades. Recently, some studies based on "top-down" atmospheric inversions reported that the net land C sink of China could reach 0.8-1.1 Pg C yr⁻¹ during the 2010s and suggested that this large 50 sink was mainly attributed to forest ecosystems (Wang et al., 2020, 2022). However, according to the size of the forest C sink we mentioned above, that large C sink could not be supported by this and other studies based on field investigations including national forest inventories. 55

4.2 Potential influencing factors of biomass C sinks

As we mentioned above, China's forests presented an enhanced biomass C sink in most provinces during the recent decade (Table S3). Further analyses of this study showed that the enhancement of the biomass C sink widely occurred in ⁶⁰ China's forests during 2009–2018 (Table S3), although the enhancing extent would be different between forest stand origins or among forest zonal types/age groups (Tables 2, 3, S7). It has been demonstrated that forest area expansion and increased forest growth jointly contributed to the enhance- ⁶⁵

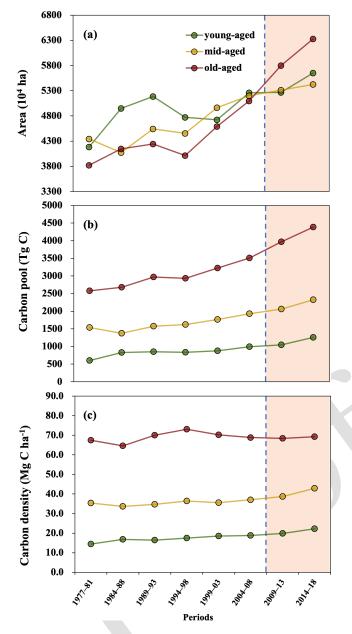


Figure 2. Changes in area (a), C pool (b), and C density (c) of forest for each age group from 1977 to 2018. The pink box highlights the latest results from the most recent 10 years.

ment of China's forest C sink, in which area expansion contributed more to planted forests while forest growth was a great contributor to natural forests (Fang et al., 2014a; Li et al., 2016). This conclusion could also explain the enhanced C

- ⁵ sink during the period of 2009–2018 observed in this study. Our results showed that between the 7th Forest Inventory (2004–2008) and the 9th Forest Inventory (2014–2018), the area of planted forest increased by 30 % (from 0.40×10^8 to 0.52×10^8 ha), while the C sink increased by only 24 %
- ¹⁰ (from 46.2 to 57.3 Tg C yr⁻¹) during the same period, indicating that area expansion is the main factor driving the en-

hanced C sink of planted forests (Table 2). In contrast, the area of natural forest increased by only approximately 6% (from 1.15×10^8 to 1.22×10^8 ha), but the corresponding C sink nearly doubled (from 66.7 to 128.1 Tg C yr⁻¹) (Table 2), suggesting that the enhanced C sink of natural forests is mainly contributed by forest growth rather than area expansion.

The area expansion of China's forest could be largely due to national ecological projects. Since the late 1970s, ²⁰ China has launched six key national ecological projects to restore degraded ecosystems and to protect the country's environment (Lu et al., 2018). Within the framework of these projects, large-scale afforestation and reforestation have been conducted, which contributed 71.2 % of the total forest area ²⁵ expansion in China during 1977–2018 (Table 2), resulting in China having the largest planted forest area in the world (Guo et al., 2013; Lu et al., 2018). Obviously, such area expansion, which was mainly led by ecological projects, would promote an increase in the biomass C sink in China's forests, ³⁰ especially in planted forests.

The increased growth in China's forest could be due to several reasons. On the one hand, environmental changes (e.g. elevated CO₂, climatic change, nitrogen deposition) may be an important reason for the increase in China's forest growth 35 over the most recent decade (Piao et al., 2009a, b; Tian et al., 2011; Pan et al., 2011, 2013; Fang et al., 2014b; Liu et al., 2021). We noticed that differing from what happened in the old forest, in which the C density remained roughly steady, the C densities in the young and middle-aged forests 40 showed a significant increase from 1977 to 2018, especially during 2009-2018 (Fig. 2). A previous study pointed out that such an increase in the C density of the forest at an age stage often means accelerated tree growth induced by environmental changes (Fang et al., 2014b). Meanwhile, af- 45 ter excluding the potential impacts from tree species and forest age, our regression analysis showed that the average C density presented significantly positive relationships with atmospheric CO₂ concentration, mean annual temperature (MAT), and nitrogen deposition (Fig. S2), which could 50 also reflect the promoting effect of elevated CO₂, rising temperature, and increased nitrogen deposition on forest growth. Over the past 4 decades, China's forests have experienced significant increases in CO2 concentration (Global Monitoring Laboratory, https://www.gml.noaa.gov/, last ac-55 cess: 1533), mean annual temperature (MATOD2; China Meteorological Data Service Center, https://data.cma.cn/data/ index.html, last access: 134), and nitrogen deposition (Eyring et al., 2013), which have increased from 342 ppm, 10.5 °C, and 0.8 g N m^{-2} during the first 5 years of the 1980s to $_{60}$ 404 ppm, 11.6 °C, and 1.4 g N m⁻² during 2014–2018, respectively. In particular, the annual mean growth rates of CO_2 and MAT in the most recent decade (2.29 ppm yr⁻¹) and $0.07 \,^{\circ}\text{C}\,\text{yr}^{-1}$, respectively) were greater than those in the previous 3 decades (1.68 ppm yr⁻¹ and $0.04 \circ C yr^{-1}$, re-65 spectively). All these changes would accelerate the growth of

forests in China and thus lead to the enhanced biomass C sink in the past 40 years, especially in the most recent decade.

On the other hand, the implementation of ecological projects could be another important reason for the increased

- ⁵ forest growth in China. These projects implemented a series of forest management practices, such as forest enclosure, tending, and reduction of timber harvesting, to promote the growth of forests and achieved remarkable effects (Xu et al., 2017; Fang et al., 2018; Lu et al., 2018). For example, due
- ¹⁰ to the illegal occupation of forestland and vast excessive logging, China's natural forest experienced a sharp decline in the area during the period of 1994–1998 (Table 2) (National Forestry Administration, 2000). Although the stock volume density and C density continued to rise, the severely shrink-
- ¹⁵ ing area induced reductions in forest stock volumes and C stocks (Table 2). Thus, the government launched the Natural Forest Protection project in 1998 (National Forestry Administration, 2000). As a result, the area and biomass C pool of natural forests increased greatly from 1.03×10^8 ha and
- ²⁰ 4746 Tg C during 1994–1998 to 1.22×10^8 ha and 6505 Tg C during 2014–2018, respectively (Table 2). Meanwhile, in the most recent decade, extensive young forests planted by these projects have gradually entered the middle-aged or premature stages, in which forests usually have rapid growth rates
- ²⁵ (Guo et al., 2013). Our results showed that during 1977– 2018, the areas of middle-aged and old forests expanded from 4.3×10^7 and 3.8×10^7 to 5.4×10^7 and 6.3×10^7 ha, respectively (Fig. 2, Table S5), suggesting that a large area of young forests entered a rapid growth stage with high C se-
- ³⁰ questration ability. The large C sinks brought by these forests could be an important source of power for China's forest C sinks after 2010 (Cai et al., 2021; Yu et al., 2021).

4.3 Effects of human intervention on forest C sink management

³⁵ In current forest C sink management, human intervention exists in many aspects, among which the most important is the cultivation of planted forest. Planted forests play a critical role in enhancing ecosystem services, absorbing atmospheric CO₂, and mitigating climate change (Fang et al., 2001; Li et ⁴⁰ al., 2018, 2019; Lu et al., 2018; Tang et al., 2018). In this study, we found that China's planted forest is an important C

sink, acting as the main source of forest C sink in nearly half of China's provinces (Table S4) and currently contributing more than 30% of the total forest biomass C sink across the ⁴⁵ whole country (Table 2). This result proved that afforestation

indeed promoted the growth of China's forest C sink.

However, it could be easy to fall into the misunderstanding of using this result to prove the success of China's afforestation and further overemphasize its importance in for-

⁵⁰ est C sink management. The national forest inventory only records the forest stands that are successfully established and last retained; thus the area of planted forest would be much less than the actual afforestation area. From 1977 to 2018, the cumulative afforestation area in China was approximately 1.9×10^8 ha, nearly 4 times the current planted forest 55 (National Forestry and Grasslands Administration, 2019a). This implied that most of the trees cultivated in afforestation projects have failed to survive to the current moment. This situation is particularly widespread in the vast arid and semi-arid regions of China. Compared with natural forests, 60 planted forests are more sensitive and vulnerable to drought stress because of their high transpiration rate, high plant density, and low biodiversity (Isbell et al., 2015; Martín-Benito et al., 2010; Zhong et al., 2021). Therefore, planted forests in arid and semi-arid regions such as northwest China generally present low survival rates due to frequent drought events (Cao, 2008; Cao et al., 2011; Wang et al., 2020; Zhang et al., 2022). Although people try to improve the survival rate of trees by selecting an appropriate afforestation method, choosing proper tree species, and cooperating with a series 70 of management measures, such as the application of fertilizer and irrigation, tending operations, etc. (Liu et al., 2016; Zhou et al., 2013), it is still a significant challenge to achieve successful afforestation in the arid and semi-arid regions of China (Wang et al., 2014; Yu et al., 2019; Zhong et al., 2021; 75 Zhou et al., 2013). Currently, most of the land available for afforestation in China is distributed in arid and semi-arid regions or other regions with poor site conditions (C. Zhang et al., 2018), which would be bound to greatly decrease in the success rate of afforestation. Thus, the realizability of en- 80 hancing forest C sinks through large-scale afforestation must be carefully evaluated in the future.

Compared with the implementation of large-scale afforestation, our results suggested that enhancing the regeneration of old-growth forests, especially those in natural 85 forests, would be of greater significance for the future management of forest C sinks. We noticed that from 1977 to 2018, the biomass C density of old forests generally remained stable (Table 2). This result suggested that the observed C sink of old forests would mainly derive from 90 their area expansion, i.e. the transformation from young and middle-aged forests to old forests, rather than forest growth. Due to the high mortality rate of old-growth trees, old forests, especially overmature forests, would even present a decline in C density, thus leading to a negative contribution to the 95 C sink (Zhao et al., 2021). Therefore, for old forests, it is necessary to properly adopt practices such as thinning, selective cutting, sanitary cutting, and reforestation (Zhao et al., 2021) to maintain a healthy state, avoid C release caused by the large-scale death of trees, and promote their regenera- 100 tion to function as a significant C sink. This would be critical for the maintenance and management of forest C sinks. Of course, such artificial regeneration practices would inevitably remove biomass C from forests, but it does not mean an equivalent amount of C release because C in harvesting 105 timbers is often turned into deposited C in wood products, e.g. furniture, house building, and instruments, which could exist for tens or even hundreds of years (Skog, 2008; van

Deusen, 2010; X. Zhang et al., 2018), providing sufficient growth time for young trees after regeneration. However, it should be noted that several studies have found that ageing forests could still maintain rising C densities, although their 5 biomass C accumulation rates gradually decreased (Cao et

al., 2012; Luyssaert et al., 2008; Yue et al., 2018; Zhao et al., 2014). Thus, the adoption of artificial regeneration practices in old forests also needs to be carefully evaluated to avoid unnecessary C release.

10 4.4 Uncertainty of estimations

The estimation involved in the study is presented with some uncertainties. In general, the national forest inventory data were assumed to have small errors of less than 5% (Fang et al., 2001). The survey accuracy of the forest area ¹⁵ and timber volume was over 90% (National Forestry and

- Grasslands Administration, 2019b). The method calculated biomass from surveyed stand volume data, and the R^2 of the BEF function (Eqs. 1 and 2) of the dominant tree species was higher than 0.80 (Table S1), suggesting that our estimates of
- ²⁰ forest biomass were statistically reliable **CES**. Previous studies have shown that the estimated error of forest biomass at the national scale using the BEF function is unlikely to exceed 3 % (Fang and Chen, 2001; Fang et al., 1996, 2002). A constant C conversion factor of 0.5 may introduce a system-
- ²⁵ atic error of -5.9% to **CSS** 2.5% (Ma et al., 2020). Despite these uncertainties, the results of this study provide relatively high accuracy and a comprehensive assessment of the forest C budget.

5 Conclusions

- ³⁰ In this study, we estimated forest biomass C storage and its changes in China over the past 40 years (1977–2018) and updated their estimates in the most recent decade (2009–2018) using the biomass expansion factor method and eight national forest inventories conducted every 5 years. We con-
- ³⁵ cluded that the Chinese forest biomass C pool increased by 3258 Tg C with an annual C sink of 88.0 Tg C yr⁻¹ from 1977 to 2018. The biomass C pool and C sink in the most recent 10 years (7525 Tg C and 154.8 Tg C yr⁻¹, 2009– 2018) were much higher than those of the previous 30 years
- ⁴⁰ (5447 Tg C and 63.3 Tg C yr⁻¹, 1977–2008), although the C sink strength displayed large variations in different periods. Afforestation practices, forest growth, and environmental changes were proposed as the main drivers of this significant C increase, especially in the most recent decade. Our
- ⁴⁵ study updates the previous estimates of China's forest C storage and its changes and provides an essential basis for policymaking for ecosystem services and the carbon neutrality target in China.

Data availability. .

Supplement. The supplement related to this article is available online at: https://doi.org/10.5194/bg-19-1-2022-supplement.

Author contributions. JF and CY designed the study. CY and YS analysed the data. ZG provided part of datasets. WS, YS, JZ, CJ, YF, SM, and JF wrote the manuscript and gave final approval for 55 publication.

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

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Acknowledgements. This study was funded by the National Natural Science Foundation of China (31988102 and 41871038). We also thank the National Forestry and Grass Administration for the data support.

Financial support. This research has been supported by the National Natural Science Foundation of China (grant nos. 31988102 and 41871038).

Review statement. This paper was edited by Ben Bond-Lamberty and reviewed by three anonymous referees.

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