



1 **Soil respiration database at different time scales in forest ecosystems across China**

2 Hongru Sun^{1,2}, Zhenzhu Xu¹, Bingrui Jia^{1*}

3 ¹*State Key Laboratory of Vegetation and Environmental Change, Institute of Botany,*

4 *Chinese Academy of Sciences, Beijing 100093, China*

5 ²*University of Chinese Academy of Sciences, Beijing 100049, China*

6 *Corresponding author:

7 Bingrui Jia

8 Institute of Botany, Chinese Academy of Sciences,

9 20 Nanxincun, Xiangshan, Haidian District, Beijing 100093, China

10 E-mail: jiabingrui@ibcas.ac.cn

11 Tel: 86-10-62836289

12 Fax: 86-10-82595962

13 **Abstract.** China's forests rank fifth in the world by area and cover a broad climatic
14 gradient from cold-temperate to tropical zones, and play a key role in the global carbon
15 cycle. Studies on forest soil respiration (R_s) are increasing rapidly in China over the last
16 two decades, but the resulting R_s data need to be summarized. Here, we compile a
17 comprehensive database of R_s in China's undisturbed forest ecosystems from literature
18 published up to December 31, 2018, including monthly R_s and the concurrently measured
19 soil temperature (N=8317), mean monthly R_s (N=5003), and annual R_s (N=634).
20 Detailed plot information was also recorded, such as geographical location, climate factors,
21 stand characteristics, and measurement description. We examine some aspects of the
22 database – R_s equations fitted with soil temperature, temperature sensitivity (Q_{10}),
23 monthly variations and annual effluxes in cold-temperate, temperate, subtropical and



24 tropical zones. We hope the database will be used by the science community to provide
25 a better understanding of carbon cycle in China's forest ecosystems and reduce
26 uncertainty in evaluating of carbon budget at the large scale. The database is publicly
27 available at <https://www.pangaea.de/tok/788910d8d3ae0a415c7bad2e7025a3f16f042a1b> (Sun
28 et al. 2021).

29

30 **Keywords:** Soil carbon flux, Carbon cycle, Temperature sensitivity, Forest, China

31 **1 Introduction**

32 Soil respiration (R_s) refers to the total amount of CO_2 released by undisturbed soil,
33 including autotrophic respiration and heterotrophic respiration, the former from plant
34 roots and their microbial symbionts, and the latter from microorganisms decomposing
35 litter and soil organic matter. As the second-largest terrestrial carbon flux, the recent
36 estimations of global annual R_s (80–98 Pg C year⁻¹) are above ten percent of the
37 atmospheric carbon pool (750 Pg C) (Bond-Lamberty and Thomson, 2010b;
38 Hashimoto et al., 2015; Raich et al., 2002; Warner et al., 2019), thus accelerating soil
39 respiration rates with climate warming have a strong potential to influence atmospheric
40 CO_2 levels. It is thus important to understand better soil respiration dynamics and
41 response to climate changes.

42 Forest area in China ranks fifth in the world (FAO, 2020) and covers a broad climatic
43 gradient, including cold-temperate, temperate, subtropical and tropical zones. In China,
44 most R_s measurements began only after 2001 (Chen et al., 2010), but have rapidly
45 increased during the last 20 years (Jian et al., 2020). Several studies have summarized
46 annual R_s in China's forest ecosystems, but with the small samples (e.g., N=50 in
47 Zheng et al., 2010; N=62 in Chen et al., 2008; N=120 in Zhan et al., 2012; N=139 in
48 Song et al., 2014). Yu et al. (2010) established a geostatistical model with a total of 390



49 monthly R_s data from different ecosystems in China. With 1782 monthly R_s in forest
50 ecosystems across China, Jian et al. (2020) analyzed the spatial patterns and temporal
51 trends from 1961 to 2014. However, amounts of R_s data are still unexploited, because
52 they were only displayed in the forms of monthly dynamics in the original papers' figures.
53 R_s data at a subannual timescales are important for upscaling global R_s (Jian et al.,
54 2018), which may derive different conclusions and deserve further exploration (Huang
55 et al., 2020).

56 The lack of the large-scale and observation-driven R_s data is a main constraining factor
57 in quantifying regional- to global-scale carbon budget (Bond-Lamberty and Thomson,
58 2010a; Rayner et al., 2005). R_s data and concurrently measured temperature thus
59 provide not only a solid base to understand the critical factors influencing R_s , but the
60 opportunity to better simulate R_s at the large scale. We attempted to compile a complete
61 forest R_s database at different temporal scales in China.

62 **2 Data and methods**

63 **2.1 Data sources**

64 The terms of “soil respiration”, “soil carbon (or CO₂) efflux”, or “soil carbon (or CO₂)
65 emission” were searched from publications before 2018 in the China Knowledge
66 Resource Integrated Database (<http://www.cnki.net/>), China Science and Technology
67 Journal Database (<http://www.cqvip.com>), ScienceDirect (<http://www.sciencedirect.com/>),
68 ISI Web of Science (<http://isiknowledge.com/>), and Springer Link
69 (<http://link.springer.com/>). Means, minimums and maximums of soil respiration during
70 the observation periods were usually given in these published studies, and monthly
71 patterns of soil respiration rates and the corresponding temperature were frequently
72 shown with figures. WEBPLOTDIGITIZER, a graphic digitizing software, was used to
73 take data from figures when values were not reported in the text (Burda et al., 2017).



74 **2.2 Data collection criteria**

75 The following criteria were used to ensure data consistency and accuracy: i) R_s was
76 measured in the field without obvious disturbances or manipulation experiments, e.g.
77 fire, cutting, nitrogen addition treatments, etc. ii) Forested swamps and commercial
78 plantations (e.g. orchard, rubber, etc.) were not examined. iii) R_s was measured either
79 by static chamber/gas chromatography (GC) or by dynamic chamber/infrared gas
80 analyzers (IRGA, model Li-6400, Li-8100, Li-8150 (LI-COR Inc., Lincoln, Nebraska,
81 USA)), which are the most popular methods and provide methodological consistency
82 (Sun et al., 2020; Wang et al., 2011; Yang et al., 2018; Zheng et al., 2010). Moreover,
83 the data has been carefully cross-checked by the authors and from different sources.

84 Based on these criteria, a total of 10288 monthly soil respiration data and 634 annual
85 soil respiration data were assembled from 568 publications. The dataset covers 28
86 provinces in China (18.61–52.86° N, 84.91–129.08° E) (Fig. 1). Meanwhile, the related
87 information was recorded, including geographical location (province, study site, latitude,
88 longitude and elevation), climate (mean annual temperature and mean annual
89 precipitation), stand description (forest type, origin, age, density, mean tree height and
90 diameter at breast height), measurement regime (method, time, frequency, collar area,
91 height and numbers) (Table 1). This forest region encompasses a large gradient of climate
92 regimes, mean annual temperature ranging from -5.4 to 23.8 °C and mean annual
93 precipitation ranging from 105 to 3000 mm.

94 **2.3 Data verification**

95 In this study, most of the R_s data (~82%) and the concurrent soil temperature at 5 cm
96 depth (T_5) and/or 10 cm depth (T_{10}) were extracted with WEBPLOTDIGITIZER, others
97 (e.g., minimum, maximum) were usually given in the original papers. To verify the



98 accuracy of the digital software, the means (R_s , T_5 , T_{10}) averaged from the extracted data
99 were compared with the corresponding means directly given in the original papers (Fig.
100 S1). The coefficients of determination (R^2) were all larger than 0.99, indicating that the
101 accuracy of WEBPLOTDIGITIZER is excellent.

102 **2.4 Monthly and annual soil respiration calculation**

103 Long-term continuous R_s could be monitored with Li-8100 or Li-8150, but there are few
104 published studies of such continuous data (Bond-Lamberty et al., 2020; Tu et al., 2015;
105 Wu et al., 2014; Yu et al., 2011). The typical days were usually selected to calculate mean
106 monthly R_s and the observation frequency was 1–12 days per month—high during the
107 growing season, but low in winter. R_s was measured throughout the day (16%) or at
108 representative time, e.g., 9:00 a.m.–11:00 a.m. (45%), 9:00 a.m.–12:00 a.m. (22%), etc.,
109 which had been validated to be close to the diurnal mean value (Xu and Qi, 2001; Yan et
110 al., 2006; Yang et al., 2018; Yao et al., 2011; You et al., 2013; Zheng et al., 2010). Annual
111 soil carbon efflux was integrated with soil respiration model (i.e. integration method) or
112 interpolated the average soil respiration rate between sampling dates (i.e. interpolation
113 method) (Shi et al., 2014). Finally, monthly R_s and annual soil carbon efflux were
114 converted to the common unit of $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ and $\text{g C m}^{-2} \text{ year}^{-1}$, respectively
115 (Bond-Lamberty and Thomson, 2010a).

116 **3 Results**

117 **3.1 Relationship between soil respiration rate and soil temperature**

118 Temperature is often the main factor determining soil respiration rates. There were
119 6341 and 2878 samples of paired R_s & T_5 and R_s & T_{10} in the database, respectively.
120 There were significantly exponential relationships of R_s with T_5 and T_{10} in forest
121 ecosystems across China, which could explain about 48% and 52% of the R_s variations,



122 respectively (Fig. S2). The exponential correlations were all significant in four climatic
123 zones, and the coefficients of determination for tropical ecosystems ($R^2=0.225-0.291$)
124 were smaller than those in other three zones ($R^2=0.516-0.934$) (Fig. 2).

125 Temperature sensitivity (Q_{10}) is defined as the factor by which R_s is multiplied when
126 temperature increases by 10 °C (Davidson and Janssens, 2006; Lloyd and Taylor, 1994).
127 Q_{10} could be calculated with the exponential equations between R_s and soil temperature.
128 At the national scale, the Q_{10} values in China's forest ecosystems from T_5 (-16.51–
129 33.58 °C) and T_{10} (-16.40–33.46 °C) were 2.05 and 2.17, respectively. The Q_{10} was the
130 largest in cold-temperate zone (T_5 : 3.74 & T_{10} : 3.32), secondary in temperate zone (T_5 :
131 2.69 & T_{10} : 3.00), and the smallest in subtropical zone (T_5 : 2.15 & T_{10} : 2.20) and
132 tropical zone (T_5 : 2.28 & T_{10} : 1.63).

133 **3.2 Monthly dynamics of soil respiration**

134 Monthly R_s appeared as a single-peak curve (Fig. 3). The largest values occurred in
135 August (4.18–4.36 $\mu\text{mol m}^{-2} \text{s}^{-1}$) in cold-temperate and temperate zones, larger than
136 the largest values in July (3.58–3.83 $\mu\text{mol m}^{-2} \text{s}^{-1}$) in subtropical and tropical zones.
137 The lowest values occurred in January in cold-temperate (0.20 $\mu\text{mol m}^{-2} \text{s}^{-1}$), temperate
138 (0.49 $\mu\text{mol m}^{-2} \text{s}^{-1}$), subtropical (1.10 $\mu\text{mol m}^{-2} \text{s}^{-1}$) and tropical zones (1.62 $\mu\text{mol m}^{-2}$
139 s^{-1}). Monthly variations were largest in cold-temperate and temperate zones,
140 secondary in subtropical zone, and smallest in tropical zone.

141 Annual mean R_s in January–December from low to high was cold-temperate (1.63
142 $\mu\text{mol m}^{-2} \text{s}^{-1}$), temperate (1.93 $\mu\text{mol m}^{-2} \text{s}^{-1}$), subtropical (2.47 $\mu\text{mol m}^{-2} \text{s}^{-1}$) and
143 tropical zones (2.57 $\mu\text{mol m}^{-2} \text{s}^{-1}$). Meanwhile, annual soil carbon emissions were
144 calculated with the annual mean R_s : 621.91 g C $\text{m}^{-2} \text{yr}^{-1}$ in cold-temperate zone, 733.31
145 g C $\text{m}^{-2} \text{yr}^{-1}$ in temperate zone, 937.15 g C $\text{m}^{-2} \text{yr}^{-1}$ in subtropical zone, and 973.35 g C



146 $\text{m}^{-2} \text{yr}^{-1}$ in tropical zone. Soil carbon emissions in growing season (May–October) and
147 winter (November–April) accounted for 85% and 15% in cold-temperate zone, 80%
148 and 20% in temperate zone, 69% and 31% in subtropical zone, 61% and 39% in tropical
149 zone. Subtropical and tropical zones still keep high soil respiration rates in winter,
150 which is the main source of their larger annual soil carbon emissions.

151 **3.3 Annual soil carbon effluxes**

152 There were 634 annual soil carbon effluxes, and most of the observations were
153 conducted in subtropical zone (61%) and temperate zone (32%) (Fig. 4). Mean annual
154 soil carbon emission was $851.88 \text{ g C m}^{-2} \text{ yr}^{-1}$ in China's forest ecosystems, ranging
155 from $260.10 \text{ g C m}^{-2} \text{ yr}^{-1}$ to $2058.00 \text{ g C m}^{-2} \text{ yr}^{-1}$. Mean annual soil carbon emissions in
156 tropical, subtropical, temperate and cold-temperate zones were 1042.01, 928.91,
157 697.85 and $684.29 \text{ g C m}^{-2} \text{ yr}^{-1}$, respectively. The former two was significantly higher
158 than the latter two, but the differences were not significant between tropical and
159 subtropical zones, and between temperate and cold-temperate zones. The differences
160 were not significant for EBF, ENF and DNF among different climate zones. DBF in
161 temperate and subtropical zones was similar ($\sim 750.00 \text{ g C m}^{-2} \text{ yr}^{-1}$), both of which were
162 larger than that in cold-temperate zone ($284.20 \text{ g C m}^{-2} \text{ yr}^{-1}$). MF in subtropical zone
163 ($977.35 \text{ g C m}^{-2} \text{ yr}^{-1}$) had significantly higher emissions than that in temperate zone
164 ($733.44 \text{ g C m}^{-2} \text{ yr}^{-1}$).

165 Evergreen forests were usually larger than deciduous ones in the same climatic zone,
166 for example, ENF ($866.98 \text{ g C m}^{-2} \text{ yr}^{-1}$) and DNF ($734.56 \text{ g C m}^{-2} \text{ yr}^{-1}$) in cold-
167 temperate zone, ENF ($699.96 \text{ g C m}^{-2} \text{ yr}^{-1}$) and DNF ($555.15 \text{ g C m}^{-2} \text{ yr}^{-1}$) in temperate



168 zone, EBF (1073.50 g C m⁻² yr⁻¹) and DBF (755.41 g C m⁻² yr⁻¹) in subtropical zone.
169 Broad-leaved forests showed significantly larger annual fluxes than coniferous forests
170 in temperate zone (DBF: 748.59 g C m⁻² yr⁻¹ vs. DNF: 555.15 g C m⁻² yr⁻¹) and
171 subtropical zone one (EBF: 1073.50 g C m⁻² yr⁻¹ vs. ENF: 717.50 g C m⁻² yr⁻¹).
172 However, DNF (734.56 g C m⁻² yr⁻¹) was larger than DBF (284.20 g C m⁻² yr⁻¹) in cold-
173 temperate zone, which was from high-latitude Great Xing'an Mountains (~51° N) and
174 high-altitude Gongga Mountain (2800-2950 m), respectively. Additionally, bamboo is
175 a special type in subtropical areas, exhibiting the highest soil carbon emissions
176 (1133.55 g C m⁻² yr⁻¹).

177 **4 Discussion**

178 **4.1 Temperature sensitivity (Q_{10}) of soil respiration**

179 Q_{10} is a key parameter in modelling the effects of climate warming on soil carbon
180 release. The Q_{10} calculated with the exponential equations of T_5 and T_{10} were 2.05 and
181 2.17 at the national scale (Fig. S2), which were lower than the averaged Q_{10} from
182 different studies in the syntheses of China's forest ecosystems (T_5 : 2.28–2.51 and T_{10} :
183 2.74–3.00, Peng et al., 2009; Song et al., 2014; Xu et al., 2015; Zheng et al., 2009) and
184 global forest ecosystems (T_5 : 2.55–2.70 and T_{10} : 3.01–3.31, Wang et al., 2010 a; b).
185 Our results were close to the Q_{10} of 2 commonly used in many biogeochemical models
186 (e.g., Cox et al., 2000; Sampson et al., 2007) and the mean Q_{10} of 2.11 estimated with
187 inverse modeling in forest soils across China (Zhou et al., 2009).

188 Temperature was the most important limiting factor for soil microbial activity and
189 root growth in cold regions, thus, R_s was more sensitive to temperature changes (Lloyd
190 and Taylor, 1994; Peng et al., 2009; Zheng et al., 2009; Zheng et al., 2020). The Q_{10}



191 increased from tropical zone to cold-temperate zone in this study, and varied from 1.63
192 to 3.74. The correlations between R_s and soil temperature were lowest in tropical zone
193 ($R^2=0.225-0.291$, Fig. 2d). The difference of the mean R_s between tropical moist
194 forests ($1260 \text{ g C m}^{-2} \text{ yr}^{-1}$) and tropical dry forests ($673 \text{ g C m}^{-2} \text{ yr}^{-1}$) was about 2-fold
195 (Raich and Schlesinger, 1992), indicating that soil moisture might play more important
196 roles.

197 **4.2 Comparisons of monthly and annual soil carbon effluxes**

198 The lowest monthly R_s occurred in January, and the largest values occurred in August
199 in cold-temperate and temperate zones and in July in subtropical and tropical zones
200 (Fig. 3). Similarly, monthly R_s of global terrestrial ecosystems reached their minima in
201 February and peaked in July and August (Hashimoto et al., 2015; Raich et al., 2002).
202 Due to the limitation of low temperature, winter observations of R_s were relatively
203 fewer in the cold-temperate and temperate zones. The R_s in winter (November–April)
204 was usually assumed to account for 20% of the total annual R_s (Geng et al., 2017; Yang
205 and Wang, 2005), which was in agreement with the proportion in temperate zone, but
206 greater than 15% in cold-temperate zone.

207 Annual soil carbon emission had been synthesized in forest ecosystems across China,
208 and the mean was $745.34 \text{ g C m}^{-2} \text{ yr}^{-1}$ (Zheng et al., 2010), $764.11 \text{ g C m}^{-2} \text{ yr}^{-1}$ (Zhan
209 et al., 2012), $917.73 \text{ g C m}^{-2} \text{ yr}^{-1}$ (Song et al., 2014) and $975.50 \text{ g C m}^{-2} \text{ yr}^{-1}$ (Chen et
210 al., 2008), and the mean of $851.88 \text{ g C m}^{-2} \text{ yr}^{-1}$ in the present study was in the mid-
211 range. The mean annual R_s in China's forest ecosystems was slightly lower than the
212 mean R_s of $990.00 \text{ g C m}^{-2} \text{ yr}^{-1}$ in global forest ecosystems (Chen et al., 2010). Warner
213 et al. (2019) modelled global R_s and found that the smallest and greatest annual soil
214 carbon emissions were in DNF (Mean= $344.10 \text{ g C m}^{-2} \text{ yr}^{-1}$) and EBF (Mean= 1310.47
215 $\text{g C m}^{-2} \text{ yr}^{-1}$), respectively. Compared with the predicted annual R_s , DNF in cold-



216 temperate (Mean=734.56 g C m⁻² yr⁻¹) and temperate zones (Mean= 555.15 g C m⁻² yr⁻¹)
217 had larger values, but those of EBF in subtropical (Mean=1073.50 g C m⁻² yr⁻¹) and
218 tropical zones (Mean=1065.09 g C m⁻² yr⁻¹) were lower (Fig. 4).

219 Mean annual soil carbon emissions from 634 annual *Rs* and 5003 mean monthly *Rs*
220 were 684.29 and 621.91 g C m⁻² yr⁻¹ in cold-temperate zone, 697.85 and 733.31 g C m⁻²
221 yr⁻¹ in temperate zone, 928.91 and 937.15 g C m⁻² yr⁻¹ in subtropical zone, and
222 1042.01 and 973.35 g C m⁻² yr⁻¹ in tropical zone (Fig. 4 and Fig. 3). The differences
223 between the directly averaged annual *Rs* and the accumulative mean monthly *Rs* were
224 small in four climatic zones, ranging from -8.24 g C m⁻² yr⁻¹ to 68.66 g C m⁻² yr⁻¹. Mean
225 annual soil carbon emissions in temperate, subtropical and tropical ecosystems were
226 745 g C m⁻² yr⁻¹, 776 g C m⁻² yr⁻¹ and 1286 g C m⁻² yr⁻¹ at the global scale, respectively
227 (Bond-Lamberty and Thomson, 2010a), which were comparable with our results.

228 **4.3 Improvements of the database**

229 The common measurement methods were selected, including Li-6400, Li-8100, Li-
230 8150 and gas chromatography, which had been proved to be consistent (Wang et al.,
231 2011; Yang et al., 2018; Zheng et al., 2010). The sample sizes of annual *Rs* were 50–
232 139 (Chen et al., 2008; Song et al., 2014; Zhan et al., 2012; Zheng et al, 2010) and 634
233 in the current study, and increased above 4-fold. The global soil respiration database
234 (SRDB-V5) collected 523 undisturbed annual *Rs* in China’s forest ecosystems (Jian et
235 al., 2021), but all methods were included, e.g. alkali absorption, gas chromatography
236 and various infrared gas analyzers. Alkali absorption method could underestimate *Rs*
237 (Chen et al., 2008; Jian et al., 2020). The total samples of mean monthly *Rs* were 5003,
238 which was much larger than the other database’s monthly samples of 1782 in China’s
239 forest ecosystems (Jian et al., 2020; Steele and Jian, 2018). Additionally, we extended
240 the database with the digital software (WEBPLOTDIGITIZER) from the monthly



241 dynamics figures of the original papers, including the paired R_s & T_5 (N=6341) and R_s
242 & T_{10} (N=2878). Predicting soil respiration from soil temperature has gained extensive
243 acceptance (Shi et al., 2014; Song et al., 2014; Sun et al., 2020). These data could be
244 used to establish the large-scale soil respiration equation and acquire the key parameters
245 of carbon cycle. Compared with the above-mentioned monthly or annual databases, this
246 study collected all available R_s data at different time scales. Bamboo forests were
247 seldom considered in the previous databases (Chen et al., 2008; Steele and Jian, 2018;
248 Zhan et al., 2012; Zheng et al., 2010), which exhibited the highest soil carbon emissions
249 (Mean=1133.55 g C m⁻² yr⁻¹, Fig. 4). With the area increasing at a high rate of 3.1%
250 per year (Song et al., 2017), bamboo forests would play an important role in regional
251 and even national carbon cycle.

252 **5 Data availability**

253 The soil respiration database in China's forest ecosystems used to produce the results
254 in this study is free to the public for scientific purposes and can be downloaded at
255 <https://www.pangaea.de/tok/788910d8d3ae0a415c7bad2e7025a3f16f042a1b> (Sun et
256 al. 2021).

257 **6 Conclusions**

258 In this study, we collected in situ R_s measurements with common infrared gas analyzers
259 (i.e. Li-6400, Li-8100, Li-8150) or gas chromatography to assemble a comprehensive
260 and uniform database of China's forest ecosystems at different time scales. Besides the
261 R_s data directly given in the original papers, the monthly patterns of R_s and the
262 concurrently measured soil temperature at 5 cm and/or 10 cm depth in the figures were
263 digitized. Meanwhile, we have made a preliminary analysis of the data. The results
264 showed that soil temperature could explain 22.5%–93.4% of the R_s variations.



265 Temperature sensitivity (Q_{10}) was about 2.05–2.17 at the national scale, increasing
266 from 1.63 in tropical zone to 3.74 in cold-temperate zone. Monthly R_s showed a single-
267 peak curve, and the largest values occurred in August (4.18–4.36 $\mu\text{mol m}^{-2} \text{s}^{-1}$) in cold-
268 temperate and temperate zones, larger than the largest values in July (3.58–3.83 μmol
269 $\text{m}^{-2} \text{s}^{-1}$) in subtropical and tropical zones. Mean annual soil carbon emissions decreased
270 from tropical (1042.01 $\text{g C m}^{-2} \text{yr}^{-1}$), subtropical (928.91 $\text{g C m}^{-2} \text{yr}^{-1}$), temperate
271 (697.85 $\text{g C m}^{-2} \text{yr}^{-1}$) to cold-temperate zones (684.29 $\text{g C m}^{-2} \text{yr}^{-1}$). This study provides
272 basic data and scientific basis for quantitative evaluation of soil carbon emissions from
273 forest ecosystems in China.

274 **Author contributions.** BJ designed the soil respiration database and searched the
275 papers until 2018. HS and BJ collected and digitized soil respiration data and compiled
276 the associated information. HS and BJ prepared the manuscript. ZX provided many
277 useful suggestions and reviewed the paper.

278 **Competing interests.** The authors declare that they have no conflict of interest.

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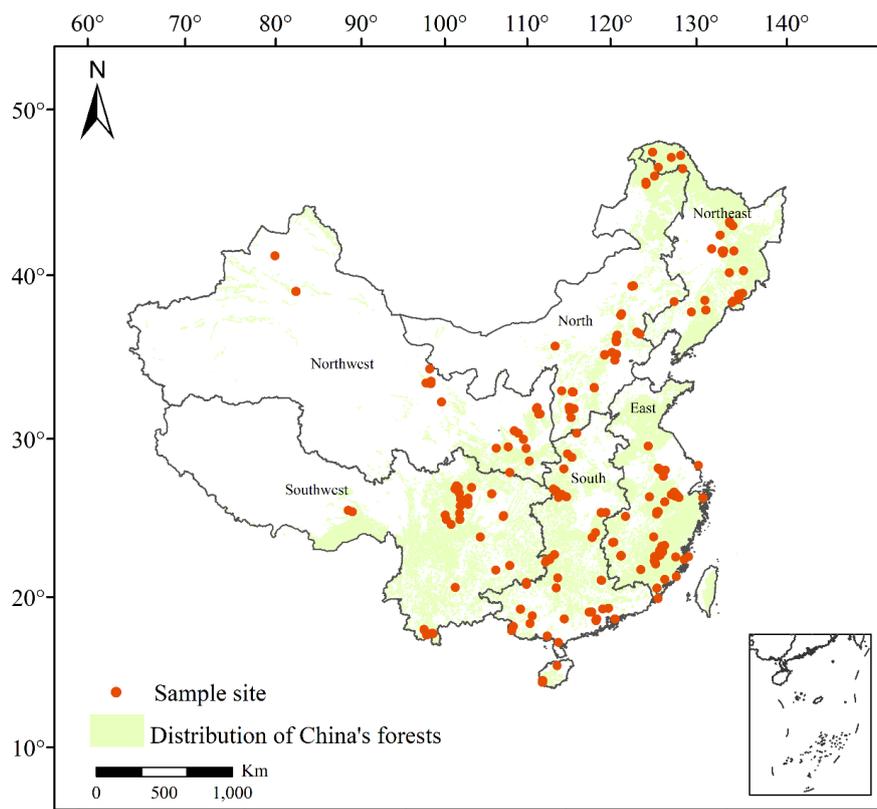


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447 **Table 1.** Variable information of soil respiration database in China's forest ecosystems,
 448 available at <https://www.pangaea.de/tok/788910d8d3ae0a415c7bad2e7025a3f16f042a1b>. N/A
 449 refers to values that are not applicable.

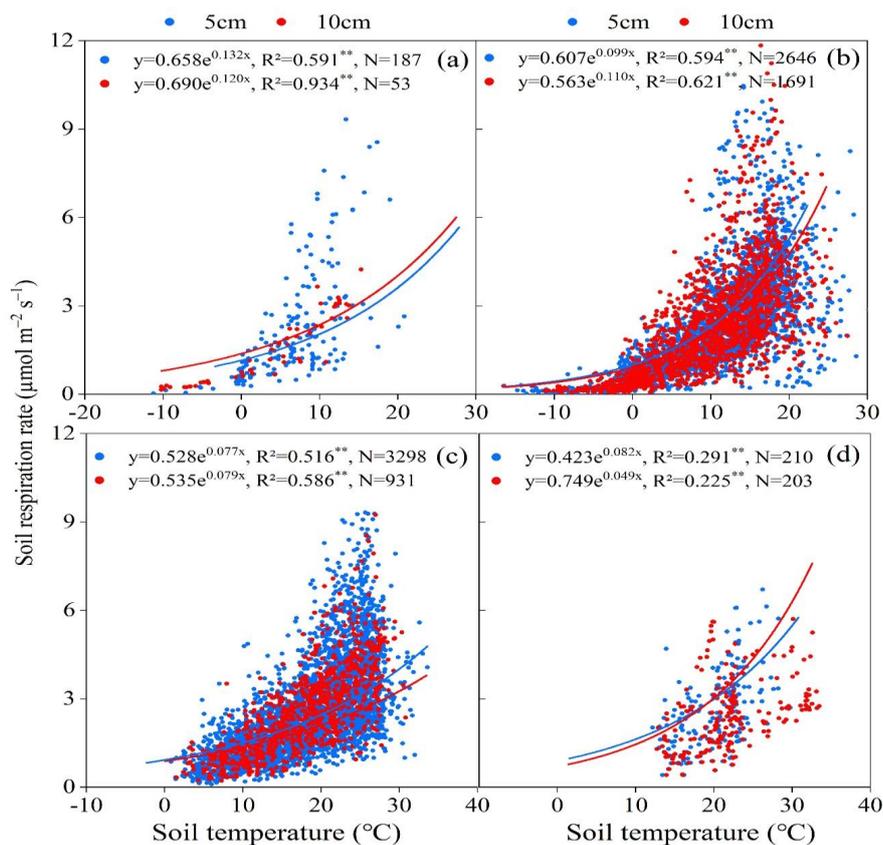
Column	Description	Unit	Number	Range
ID	Unique identification number of each record	N/A	11297	1–11297
Province	Province location of study site	N/A	28	N/A
Study site	Name of study site	N/A	155	N/A
Latitude	Latitude (N) of study site	°	988	18.61–52.86
Longitude	Longitude (E) of study site	°	988	84.91–129.08
Altitude	Altitude of study site	m	988	7–4200
MAT	Mean annual temperature	°C	988	-5.4–23.8
MAP	Mean annual precipitation	mm	988	105–3000
Forest type	Forest community characterized by the dominant tree species, or the ecological similarities (e.g. life form and biotope)	N/A	180	N/A
Origin	Stand origin was classified into planted and natural (i.e. secondary, primary) forests	N/A	4	N/A
Age	Stand age	years	769	2–~400
DBH	Mean diameter at breast height	cm	610	2.40–51.96
H _{tree}	Mean tree height	m	538	2.50–48.00
Density	Stem density and/or canopy coverage	trees ha ⁻¹	548	209–17000,0.23–0.98
Instrument	Measurement instrument of <i>R_s</i> , i.e. gas chromatography, infrared gas analyzers (Li-6400, Li-8100, Li-8150)	N/A	4	N/A
Time	Observation time of <i>R_s</i>	Hour: Minute	749	0:00–23:00
Frequency	Observation frequency of <i>R_s</i> , i.e. days per month	days	961	0.5–31
Area	Observation area of <i>R_s</i> , i.e. area of soil collar or base	cm ²	976	50–2500
Height	Height of soil collar or chamber	cm	828	4–50
Replication	Numbers of soil collar or chamber	N/A	968	1–768
Month	Observation month of <i>R_s</i>	Month, Year	10288	Jan.,2000–Mar.,2018
<i>R_s</i>	Soil respiration rate	μmol m ⁻² s ⁻¹	10288	0.01–11.84
<i>T₅</i>	Soil temperature at 5cm depth concurrently measured with <i>R_s</i>	°C	6341	-16.51–33.58
<i>T₁₀</i>	Soil temperature at 10cm depth concurrently measured with <i>R_s</i>	°C	2878	-16.40–33.46
Mode	The ways to obtain <i>R_s</i> data, 1: extracted with WEB PLOTDIGITIZER, 2: directly given in the original study	N/A	2	1–2
Period	Period of annual soil carbon efflux	Month, Year	631	Jan.,2001–Mar.,2018
Annual <i>R_s</i>	Annual soil carbon efflux	g C m ⁻² year ⁻¹	634	260.10–2058.00
Method	Method to calculate annual soil carbon efflux, i.e. integration method and/or interpolation method	N/A	3	N/A
Reference	Data sources	N/A	568	N/A



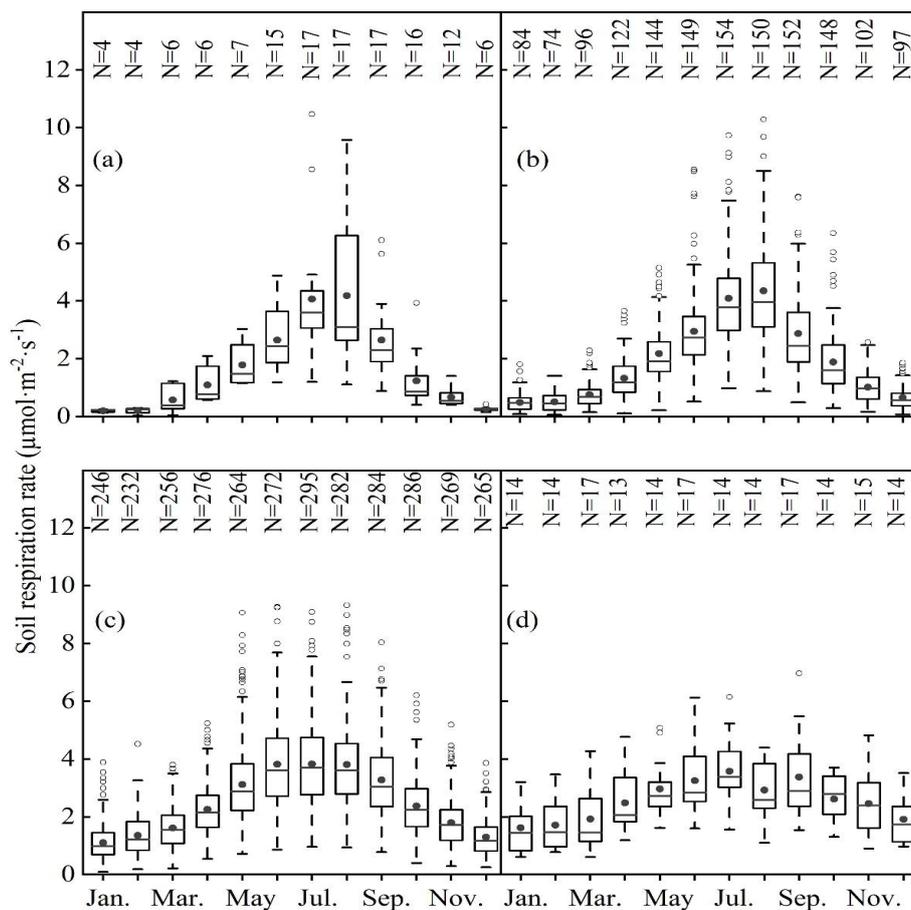
451

452 **Figure 1.** Distribution of study sites used to develop the forest soil respiration database

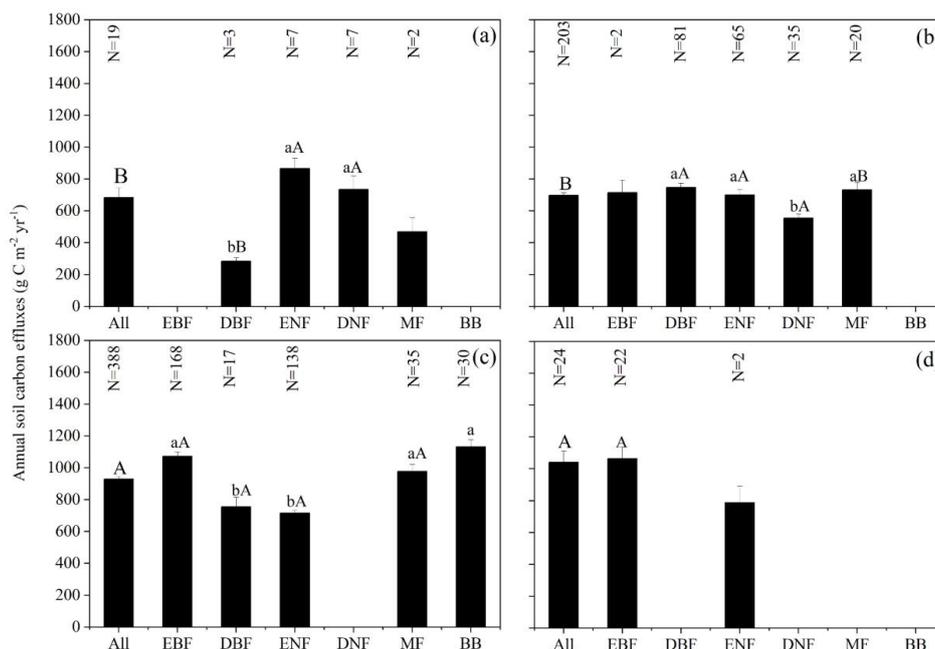
453 in China.



454 **Figure 2.** Exponential relationships of forest soil respiration rates with soil temperature
455 at 5 cm depth and 10 cm depth in cold-temperate (a), temperate (b), subtropical (c) and
456 tropical zones (d). P value below 0.01 was described by **.



457 **Figure 3.** Monthly patterns of forest soil respiration rates in cold-temperate (a),
 458 temperate (b), subtropical (c) and tropical zones (d). Solid circle: mean value; Solid
 459 horizontal line: median; Box: 25th to 75th percentiles; Whisker: 1.5 times interquartile
 460 range; Open circle: data points beyond the whiskers. The samples per month were listed
 461 in the upper part of the figure.



462 **Figure 4.** Comparisons of annual soil carbon effluxes (mean \pm standard error) among
 463 different forest types across China in cold-temperate (a), temperate (b), subtropical (c)
 464 and tropical zones (d). Lowercase letters are the comparisons of different forest types
 465 in each climatic zone, while capital letters are the comparisons of the same forest type
 466 in different climatic zones. The samples were listed in the upper part of the figure, and
 467 the samples larger than 3 were compared. EBF: evergreen broadleaf forest, DBF:
 468 deciduous broadleaf forest, ENF: evergreen needleleaf forest, DNF: deciduous
 469 needleleaf forest, MF: broadleaf and needleleaf mixed forest and BB: Bamboo forest.