

Stable isotopic constraints on global soil organic carbon turnover

Chao Wang¹, Benjamin Z. Houlton², Dongwei Liu¹, Jianfeng Hou^{1,3}, Weixin Cheng^{1,4}, Edith Bai^{1,5,*}

¹ICAS Key Laboratory of Forest Ecology and Management, Institute of Applied Ecology, Chinese Academy of Sciences, Shenyang, 110164, China.

²Department of Land, Air and Water Resources, University of California, Davis, CA, 95616, USA.

³College of Resources and Environment, University of Chinese Academy of Sciences, Beijing, 100049, China.

⁴Department of Environmental Studies, University of California, Santa Cruz, 1156 High Street, Santa Cruz, CA, 95064, USA.

⁵School of Geographical Sciences, Northeast Normal University, Changchun, 130024, China.

Correspondence to: Edith Bai (baie@iae.ac.cn)

Abstract. Carbon dioxide release during soil organic carbon (SOC) turnover is a pivotal component of atmospheric CO₂ concentrations and global climate change; however, reliably measuring SOC turnover rates at large spatial and temporal scales remains challenging. Here we use a natural carbon isotope approach, defined as beta (β), which was quantified from the $\delta^{13}\text{C}$ of vegetation and soil reported in the literature (177 separate soil profiles), to examine large-scale controls of climate, soil physical properties and nutrients over patterns of SOC turnover across terrestrial biomes worldwide. We report a significant relationship between β and calculated soil C turnover rates (k), which were estimated by dividing soil heterotrophic respiration rates by SOC pools. $\ln(-\beta)$ exhibits a significant linear relationship with mean annual temperature, but a more complex polynomial relationship with mean annual precipitation, implying strong-feedbacks of SOC turnover to climate changes. Soil nitrogen (N) and clay content correlate strongly and positively with $\ln(-\beta)$, revealing the additional influence of nutrients and physical soil properties on SOC decomposition rates. Furthermore, a strong ($R^2 = 0.76$; $p < 0.001$) linear relationship between $\ln(-\beta)$ and estimates of litter and root decomposition rates suggests similar controls over rates of organic matter decay among the generalized soil C stocks. Overall, these findings demonstrate the utility of soil $\delta^{13}\text{C}$ for independently benchmarking global models of soil C turnover and thereby improving predictions of multiple global change influences over terrestrial C-climate feedback.

1 Introduction

Soil contains a large amount of organic carbon (C) and plays a crucial role in regulating Earth's C cycle and climate system (Schmidt et al., 2011; Reichstein et al., 2013). Approximately 1500 Gt of soil organic carbon (SOC) is stored in the upper meter of global mineral soil (Scharlemann et al., 2014), which is equivalent to ~160 years-worth of current fossil fuel CO₂ emissions. Disagreement exists, however, over the residence time of this vulnerable C stock and its relationship to factors of ongoing change, particularly climate changes and widespread nitrogen pollution (Reay et al., 2008; Reichstein et al., 2013). Biogeochemical models rely heavily on turnover rates of discrete SOC pools (active, intermediate, and recalcitrant) derived from lab incubation studies (Davidson and Janssens, 2006; Xu et al., 2016). In practice, however, SOC pools fall along a continuum of characteristic turnover times (from days to centuries; (Schmidt et al., 2011; Lehmann and Kleber, 2015)), in a given ecosystem site. Furthermore, lab-derived estimates of SOC turnover disrupts the sensitive balance between plant-soil-microbe interactions in ecosystems, adding questions on the reliability of such techniques when applied to real-world conditions.

The rate of SOC turnover is an important parameter for process-based ecosystem models (Davidson and Janssens, 2006; Schimel et al., 1994) and those used to forecast the global carbon cycle and climate system in the future (Friedlingstein et al., 2006). Global biogeochemical models often use climatic factors such as precipitation and temperature to predict SOC turnover rates (Schimel et al., 1994; Nishina et al., 2014). While several studies reported positive relationships between temperature and SOC turnover (Chen et al., 2013; Trumbore et al., 1996; Bird et al., 1996; Trumbore, 1993; Carvalhais et al., 2014), however, others

38 have called the generality of such relationships into question (Giardina and Ryan, 2000). This discrepancy could be due to
39 interactions among factors which are difficult to separate in the field, for example, among soil temperature, soil moisture and
40 nutrient controls over SOC decomposition (Davidson and Janssens, 2006). Nitrogen (N) in particular can affect SOC
41 decomposition by changing microbial community structure, microbial activity or both (Curiel et al., 2007). Incorporation of factors
42 besides climate is crucial for improving model performance and predicting the feedback-response of the terrestrial carbon cycle to
43 climate change (Nishina et al., 2014).

44 In addition, questions remain regarding whether the turnover of different C stocks behaves fundamentally similarly. For
45 example, climate is considered to be a “master regulator” of leaf litter (Zhang et al., 2008), root (Gill and Jackson, 2000; Silver and
46 Miya, 2001) and soil organic C pools (Davidson and Janssens, 2006). However, recent findings have pointed out that soil microbial
47 community composition may play a more important role in litter decomposition rates than climate or litter quality (Bradford et al.,
48 2016; Keiser and Bradford, 2017). Due to the different microbial communities among leaves, roots and soils, and different chemical
49 composition of such pools, rates of C turnover have the potential to vary widely across generalized classes of C stocks.

50 Stable carbon isotope composition ($\delta^{13}\text{C}$) provides relatively non-disruptive insights into the turnover of SOC (Garten et al.,
51 2000; Accoe et al., 2002; Powers and Schlesinger, 2002; Bird et al., 1996). For sites with reasonably stable vegetation stocks,
52 measures of vertical soil-profile $\delta^{13}\text{C}$ can provide constraints on SOC turnover rates in ecosystems (Acton et al., 2013; Garten et
53 al., 2000; Wynn et al., 2006). Soil $\delta^{13}\text{C}$ generally increases from shallow to deep mineral soils in relatively well-drained systems,
54 concomitant with decreasing SOC concentrations (Fig. S1). The vertical distribution of the $\delta^{13}\text{C}$ reflects microbial preferences for
55 ^{12}C vs. ^{13}C in decomposing substrates (Garten et al., 2000), which, in turn, increases the $^{13}\text{C}/^{12}\text{C}$ of residual organic C fractions
56 with a kinetic isotope effect defined by ϵ (Fig. S1). Therefore, SOC $\delta^{13}\text{C}$ tends to increase with depth along vertical soil profiles
57 until it reaches a maximum value at which point a steady-state is achieved (Kohl et al., 2015; Accoe et al., 2002; Brunn et al.,
58 2014; Garten et al., 2000; Wynn et al., 2006; Brunn et al., 2016). These trends result in a negative linear relationship between the
59 log-transformed SOC concentration and soil $\delta^{13}\text{C}$ (Acton et al., 2013; Garten et al., 2000; Garten and Hanson, 2006; Powers and
60 Schlesinger, 2002). The slope of the linear regression between soil $\delta^{13}\text{C}$ and the log-transformed SOC concentration is defined as
61 beta (β), which has been proposed as a proxy for SOC turnover rate in a select number of sites (Acton et al., 2013; Garten et al.,
62 2000; Powers and Schlesinger, 2002). β has also been assessed in set of regional scale analyses (Acton et al., 2013; Brunn et al.,
63 2014); however, whether β values can be used to constrain rates and controls on SOC turnover is yet to be explored at the global
64 scale. Furthermore, the Suess effect (the atmospheric isotopically depleted fossil fuel CO_2 lowering atmospheric $\delta^{13}\text{C}\text{-CO}_2$)
65 (Boström et al., 2007; Wynn et al., 2006) and the mixing of different C sources (Acton et al., 2013; Diocion and Kellman,
66 2008; Wynn, 2007) may also influence the profile of soil $\delta^{13}\text{C}$ with depth. Identifying the relative influence of different factors in
67 vertical profiles of soil $\delta^{13}\text{C}$ is essential to applying this proxy to patterns of SOM turnover.

68 Here, we examine the efficacy of β as a proxy for SOC turnover rates by synthesizing soil profile data from sites around the
69 world (Fig. 1). To understand the overall utility of C isotope composition for constraining SOC turnover rates, we explore the
70 relationship between β and modeled SOC decomposition constant k and environmental factors, particularly climate, soil clay
71 content and nutrient availability. We also compare the variation of β with that of root and litter turnover rates across latitude
72 (thermal) gradient to examine whether and how the decomposition of generalized C pools varies as a function of likely controls.

73 2 Materials and Methods

74 2.1 Data compilation

75 Using the key words of ‘carbon isotope & vertical profile’, ‘ $\delta^{13}\text{C}$ & soil depth profile’, or ‘soil carbon turnover & stable isotope’
76 on the Web of Science source, we assembled a total of 150 soil profiles from 53 journal papers (Fig. 1; A list of the literature
77 sources is given in Table S1). Only soil profiles under pure C_3 vegetation without significant human disturbance were selected. For
78 each profile, we collected carbon isotope ($\delta^{13}\text{C}$), organic carbon (SOC) and N concentration of leaf/litter and mineral soil layers at
79 different depths if the data is available, and more than four $\delta^{13}\text{C}$ values should be provided within the top 1 meter. Where data were
80 not available in tables, Data Thief software (<http://www.datathief.org/>) was used to acquire values from figures. We also noted the
81 experiment location (latitude and longitude), biome types, mean annual precipitation (MAP), and mean annual temperature (MAT)
82 for each soil profile. In cases where climate variables were not reported (15 soil profiles out of 177), we used the WorldClim data
83 (<http://www.worldclim.com/>), which have average monthly temperature and precipitation between 1970 to 2000 with a resolution
84 of $\sim 1 \text{ km}^2$ at the global scale, to reconstruct climate values based on latitude and longitude coordinates in ArcGIS version 10.0
85 using Spatial Analysis tool (ESRI, Redlands, CA).

86 In addition, a previous reported arid and semi-arid grassland transect along 3000 km with 27 sampling locations was added
87 into the dataset (Wang et al., 2017). Those sampling sites are dominated by C_3 plants and cover approximately 16° longitude
88 ranging from $104^\circ 52' \text{ E}$ to $120^\circ 21' \text{ E}$ and 10° latitude ranging from $40^\circ 41' \text{ N}$ to $50^\circ 03' \text{ N}$. The MAP ranges from 90 mm to 420
89 mm and MAT ranges from -2° C to $+7^\circ \text{ C}$. At each location, five $1 \text{ m} \times 1 \text{ m}$ sub-plots (or one $5 \text{ m} \times 5 \text{ m}$ sub-plot in areas with
90 shrub as the dominating plants) were setup within a $50 \text{ m} \times 50 \text{ m}$ plot. Twenty soil cores (0 - 100 cm) in each $1 \text{ m} \times 1 \text{ m}$ sub-plot
91 were collected and divided into 0-10 cm, 10-20 cm, 20-40 cm, 40-60 cm and 60-100 cm depth segments and bulked to form one
92 composite sample for each segment per sub-plot. Leaf samples of five dominating genera (*Stipa*, *Leymus*, *Caragana*, *Reaumuria*
93 and *Nitraria*) were sampled for carbon isotope analysis if these genera were present in the sub-plots.

94 In laboratory, leaf samples were washed with deionized water to remove dust particles and then dried at 65° C for 48 h. Both
95 soil and leaf samples were ground in a ball mill and stored in a plastic bag. Soil carbonate was removed from soil samples using
96 0.5 M HCl . Organic carbon concentration and isotope composition of soil and leaf were carried out at the Stable Isotope Faculty
97 of University of California, Davis.

98 2.2 Beta calculation

99 A negative linear regression between the \log_{10} -transformed SOC concentration and $\delta^{13}\text{C}$ for each soil depth profile was conducted
100 (Fig. S1). The slope of the this linear regression is defined as beta (β) value (Acton et al., 2013; Garten et al., 2000; Powers and
101 Schlesinger, 2002).

102 2.3 Soil decomposition rate constant (k)

103 We applied the steady-state assumption in our study. While this is not always valid – especially in highly disturbed human altered
104 environments such as agriculture – this is a reasonable approximation in natural ecosystems, where SOC turnover on decadal to
105 century to millennial time-scales. The carbon decomposition rate constant (k) was estimated as the ratio between soil heterotrophic
106 respiration (R_{H}) and soil organic carbon stock (SOC) (Sanderman et al., 2003): $k = R_{\text{H}} / \text{SOC}$. The SOC stock for each soil profile
107 was extracted from a global soil organic carbon map (within 1 m depth), which was created by IGBP-DIS (1998) with a resolution
108 of 0.5 by 0.5 degree; the mean annual soil total respiration (R_{s}) was extracted from a long-term dataset with a resolution of 0.5 by
109 0.5 degree (Raich et al., 2002). We added the two datasets and the coordinate of soil profiles in ArcGIS (version 10.0, ESRI,

110 Redlands, CA) to extract SOC stock and respiration rates for each profile using the Spatial Analysis tool. Then, we used the linear
111 relationship between soil respiration (R_S) and R_H to calculate R_H (Bond - Lamberty et al., 2004): $\ln(R_H) = 1.22 + 0.73 \times \ln(R_S)$.

112 2.4 Data analysis

113 Negative β value and decomposition rate constant k were log-transformed to perform statistical tests. Larger $\ln(-\beta)$ translated to
114 faster SOC decomposition rates (Acton et al., 2013; Powers and Schlesinger, 2002). Soil $\ln(-\beta)$ was analyzed and summarized
115 across different biome types. Soil $\ln(-\beta)$ was also compared with litter and root decomposition rate along latitude at the global
116 scale. Two-variable regression analysis was first performed to explore the relationship between $\ln(-\beta)$ and $\ln(k)$, or $\ln(-\beta)$ and
117 climate variables (MAP and MAT) as well as soil edaphic factors (N and clay content). Multiple regression analysis was then used
118 to examine the relationship between $\ln(-\beta)$ and these variables (MAP, MAT, soil N and clay content). Akaike information criterion
119 (AIC) was used to estimate the quality of model when increasing the number of parameters.

120 3 Results

121 3.1 Worldwide patterns of β

122 A total of 177 soil profiles from all continents other than Antarctica were encapsulated in our compiled dataset (Fig. 1). Carbon
123 isotope composition ($\delta^{13}C$) increased with soil depth in the majority of examined profiles and was strongly correlated with the
124 logarithm of SOC (Fig. S1). $\ln(-\beta)$ was significantly positively related with site-based estimates of the soil C decomposition
125 constant, $\ln(k)$ with $R^2 = 0.34$ (Fig. 2).

126 The values for $\ln(-\beta)$ ranged from -0.50 to 2.20 across sites (non-transformed β values ranged from -0.60 to -7.41, Table S1).
127 Highest mean $\ln(-\beta)$ was observed in tropical forests (Fig. 3), followed by deserts, temperate forests, and temperate grasslands,
128 with a mean value of 1.15, 0.70, 0.58 and 0.49, respectively. MAP among those four biomes increased from desert < temperate
129 grassland < temperate forest < tropical savanna < tropical forest and MAT increased from temperate grassland < temperate forest
130 < desert < tropical forest (Fig. S2).

131 Along the latitude gradient, $\ln(-\beta)$ decreased from the equator to poles, but was higher at 20-30° N compared to the 10-20° N
132 latitudinal band (Fig. 4a). The mean decomposition rate of leaf-litter and root C displayed similar latitudinal patterns ($R^2 = 0.76$;
133 $p < 0.001$, Fig. 4b).

134 3.2 Controls on β across ecosystems

135 The $\ln(-\beta)$ and MAT displayed a strong, positive relationship across the global dataset ($R^2 = 0.43$; $P < 0.001$; Fig. 5a). $\ln(-\beta)$ did
136 not show a simple linear correlation with MAP, but instead showed a polynomial relationship with a tipping point at MAP = 3000
137 mm (Fig. 5b). When MAP was less than 3000 mm, $\ln(-\beta)$ was positively correlated with MAP ($R^2 = 0.23$, $P < 0.001$); $\ln(-\beta)$
138 decreased with the increasing of MAP in areas receiving > 3000 mm of MAP. A quadratic equation provided the best fit to the
139 relationship between $\ln(-\beta)$ and MAP for all sites ($R^2 = 0.24$, $P < 0.001$; Fig. 5b). Soil N explained 20% of the variations in $\ln(-\beta)$
140 ($P < 0.001$; Fig. 5c). Moreover, a quadratic equation best described the relationship between soil clay and $\ln(-\beta)$, with $R^2 = 0.49$ (P
141 < 0.001 ; Fig. 5d). AIC analysis showed that the full-factors model (i.e., MAT, MAP, soil N and clay) accounted for more of the
142 variation in $\ln(-\beta)$ than any other regression model in the global data set (Table 1).

143 4 Discussion

144 Our global data synthesis reveals significant relationships between $\ln(-\beta)$ and the turnover of soil, litter and root C pools at

145 geographically broad scales (Fig. 2 and 4). These findings build on site-based observations and regional assessments (Accoe et al.,
146 2002;Garten et al., 2000;Powers and Schlesinger, 2002;Brunn et al., 2014;Brunn et al., 2016), and suggest that C isotope
147 composition is a useful proxy for understanding generalized patterns of SOC turnover and the underlying controls over soil C
148 metabolism. That our results hold across all major soil C pools implies that SOC, root and litter turnover share common controls,
149 in particular those related to climate and soil nitrogen contents. These findings suggest that decomposition of belowground and
150 aboveground soil C may have similar responses to global climate change, such as global warming and increasing atmospheric N
151 deposition. Furthermore, our results highlight the potential of incorporating natural stable C isotopes in global biogeochemical and
152 Earth system models to constraint soil and litter decomposition rates that are vital to climate change forecasts.

153 Within the terrestrial biosphere, our findings point to highest mean $\ln(-\beta)$ in tropical forest ecosystems, which is consistent
154 with previous studies (Carvalhais et al., 2014). At the other extreme, our analysis suggests that slowest mean rates of SOC
155 decomposition occur in temperate grassland (Fig. 3), consistent with results from previous simulation modeling (Carvalhais et al.,
156 2014;Schimel et al., 1994). Relatively slow decomposition rates have been observed for plant litter decay in arid grassland sites
157 (Zhang et al., 2008), and largely reflects strong moisture controls on decomposition. In addition, microbial biomass and microbial
158 activities are much lower in arid/semi-arid vs. mesic or humid sites (Fierer et al., 2009), thus leading to low rates of SOC and litter
159 decomposition.

160 **4.1 Climate and nutrient dependences of β**

161 The differences of β value among different biomes reflects several controlling variables – especially mean annual temperature,
162 mean annual precipitation, soil N contents, and clay content. Of particular importance are temperature-driven controls over β , in
163 which MAT explains 43% of the variation of $\ln(-\beta)$ in our global data compilation (Table 1; Fig. 5a). A recent meta-analysis, which
164 included 24 soil profiles across a range of cool temperate to tropical forest sites, reported similarly strong temperature-dependencies
165 of β (Acton et al., 2013). Our findings broaden this perspective to a global range of terrestrial biomes and climates, and indicate
166 that, with increasing MAT, SOC turnover is substantially accelerated. This result agrees with previous studies which have identified
167 temperature as the strongest regulator of soil C decomposition among all known controls (Carvalhais et al., 2014;Schimel et al.,
168 1994), and is consistent with global C-climate feedback models, which project accelerated rates of CO_2 efflux from the land
169 biosphere with climate warming (Ciais et al., 2014).

170 Our study also points to significant relationships between β and precipitation-climates, which are more complex than those
171 observed for MAT. Rather, we find an inflection point in β in our global data set at MAP \sim 3000 mm (Fig. 5b). This relationship
172 reveals negative effects of moisture on change of soil $\delta^{13}\text{C}$ in very wet climates. A cross-system compilation of the smaller though
173 more dynamic litter pool shows a similar pattern of decreasing decay rates in regions with MAP $>$ 3000 mm compared to drier
174 sites (Zhang et al., 2008). In addition, Schuur (2001) showed that leaf and root decomposition rates declined significantly with
175 increasing precipitation along a highly constrained rainfall sequence in Hawaiian forest sequence (from 2020 mm $<$ MAP $<$ 5050
176 mm), thereby resulting in slower rates of nutrient mineralization and declines in net primary production (NPP) in the wettest sites.
177 The consistencies between our study and past work suggests that precipitation affects the decomposition of SOC and litter in
178 similar ways, slowing decomposition rates when MAP is very high and anaerobic conditions dominate (i.e. MAP $>$ 3000 mm;
179 Schuur 2001).

180 In addition to climate, nutrients influence the magnitude of $\ln(-\beta)$ in our compilation, with SOC turnover rates generally
181 increasing with soil N concentrations across ecosystem sites (Table 1, Fig. 5c). Although soil N has been suggested as an important
182 control over SOC decomposition in previous work (Schimel et al., 1994), our study is one of the few to confirm the existence of
183 such a relationship at the global scale. Positive correlations between litter decomposition rates and litter N contents during the early

184 stages of decay have been reported previously (Berg, 2000). Past work has also suggested that high N availability enhances soil
185 degrading enzyme activities (Fioretto et al., 2007).

186 Finally, our results suggest that soil physical factors, particularly soil clay content, plays a role in $\ln(-\beta)$ and soil organic C
187 turnover (Fig. 5d), consistent with previous expectations (Schimel et al., 1994; Xu et al., 2016). In sites where clay content is < 50%
188 (i.e., sandy soils), for example, $\ln(-\beta)$ increases with the soil clay content; however, when clay content is > 50% (loamy or clayey
189 soils), no clear relationship between $\ln(-\beta)$ and clay content is observed (Fig. 5d). The change in this relationship could be explained
190 by the higher “preservation capacity” of clayey soils (Vogel et al., 2014).

191 SOC turnover is an important parameter for process-based models and Earth system models (Schimel et al., 1994; Davidson
192 and Janssens, 2006), and models used to forecast the carbon cycle and climate system into the future (Friedlingstein et al., 2006).
193 Global biogeochemical models commonly use climatic factors as predictors of SOC turnover rates (Carvalhais et al., 2014). In
194 contrast, our results point to factors beyond climate singly, soil N content and soil texture, in altering organic C turnover across the
195 terrestrial biosphere. Taken together, for instance, our multiple regression analysis considering all factors (i.e., MAT, MAP, soil
196 N and clay) explains nearly 70% of variation of $\ln(-\beta)$ ($R^2 = 0.66$, $P < 0.001$; Table 1), suggesting the high dependence of SOC
197 turnover on these factors. We therefore suggest the need for models that include all of these factors when forecasting global C
198 cycle response to change.

199 In addition, our findings suggest that the C isotope composition of the soil can help to improve global C model performance.
200 A common problem in global C research is finding consistent and sufficiently integrated metrics against which the performance of
201 different biogeochemical models can be quantitatively analyzed (Tian et al., 2015). The strong relationships we observe between
202 β and SOC turnover suggest that this natural-isotope proxy can be used to ground-truth large-scale patterns of model-simulated
203 soil C dynamics. Future work to collect and analyze C isotope data in vertical soil profiles, which is a relatively inexpensive process,
204 can further extend the regional coverage of β and help benchmark SOC turnover estimates among global model simulations. This
205 is important given the potential for SOC turnover to respond to multiple global changes and produce significant feedbacks on
206 climate at the global scale (Carvalhais et al., 2014; Lehmann and Kleber, 2015).

207 **4.2 Alternative controls over soil $\delta^{13}\text{C}$ with depth**

208 Several processes have been proposed to explain the widely observed pattern of increasing $\delta^{13}\text{C}$ from shallow to deep soils beyond
209 those related to kinetic isotope fractionation during decomposition. For example, atmospheric $\delta^{13}\text{C}$ -CO₂ has been decreasing since
210 the industrial revolution owing to the combustion of ¹³C-depleted fossil fuels, which could lead to lower $\delta^{13}\text{C}$ in surface soils vs.
211 deeper horizons (Friedli et al., 1987). However, the magnitude of this effect is small (i.e., 1.4 - 1.5‰) compared to the substantial
212 variation of soil $\delta^{13}\text{C}$ along depth profiles (~3.5‰) in our dataset. Based on a 100-year-old soil archive (i.e. soil collected before
213 extensive fossil fuel CO₂ emissions) and modern samples collected from a common site in the Russian steppe, Torn *et al.* (2002)
214 demonstrated that $\delta^{13}\text{C}$ profiles of modern and pre-industrial soils were similar. Thus, the $\delta^{13}\text{C}$ values of SOM in soil profiles that
215 can be attributed to the changes in the $\delta^{13}\text{C}$ of atmospheric CO₂ should be small.

216 In addition, bioturbation and consequent mixing of C from different sources has the potential to alter soil $\delta^{13}\text{C}$ profiles (Acton
217 et al., 2013; Ehleringer et al., 2000). The $\delta^{13}\text{C}$ of root material is generally higher than that of above ground biomass, such as leaves
218 (Powers and Schlesinger, 2002), and so the $\delta^{13}\text{C}$ of SOC at the soil surface may be lower than deep soils. Because microbes,
219 invertebrates, and other soil fauna are typically enriched in $\delta^{13}\text{C}$ compared to source-substrates, biological migration and physical
220 mixing of soils may alter relationships between soil C concentrations and $\delta^{13}\text{C}$ (Wynn et al., 2006). Kohl et al. (2015) suggested
221 that increased proportions of soil bacteria (¹³C enriched) relative to fungi (¹³C depleted) biomass might also contribute to increasing
222 $\delta^{13}\text{C}$ with depth; however the process-based modeling predicts the opposite pattern (Acton et al., 2013).

223 Moreover, in extremely wet sites, it is likely that leaching of dissolved organic carbon (DOC) from soils to streams affects the
224 relationship between decomposition and isotope effect expression (Powers and Schlesinger, 2002). Previous studies have shown
225 that DOC increases with increasing soil depth (Kaiser et al., 2001). Because DOC is generally ^{13}C -enriched (Kaiser et al., 2001),
226 increasing DOC leaching into very wet sites would be expected to induce a larger change in soil $\delta^{13}\text{C}$ with depth, and hence,
227 increasing $\ln(-\beta)$.

228 **4.3 Uncertainty on carbon isotope method**

229 It is important to stress that we applied steady-state assumptions in estimating turnover, which would not hold for all sites,
230 especially highly disturbed or human altered environments, such as agricultural soils. Application of our results to ecosystems that
231 are far from steady-state should be approached with caution. Further, we used multiple data sources to estimate soil carbon store
232 and respiration for profiles in our dataset and then calculated the k values, which might add uncertainty to the comparison between
233 $\ln(-\beta)$ and $\ln(k)$. Finally, the coordinates that we used to extract climate data for some soil profiles (15 soil profiles out of 177)
234 from a world climate dataset may not match the exact locations of these soil profiles.

235 **5 Conclusion**

236 Our analysis provides a globally integrative tool for understanding variations of SOC turnover rate, which can be applied spatially
237 based on estimates of factors such as climate and soil properties. Compared with other methods, utilization of C isotope
238 composition ratios in soil profile provides an independent approach that does not rely on disruption of plant-soil-microbe
239 interactions. It has the added benefit of integrating over longer time scales (decade to centuries), and thus provides a common
240 measurement for model-based benchmarking and calibration schemes.

241 **Author contribution.** C.W. and E.B. conceived and wrote the paper with contributions from B.Z.H. C. W., W. C., D. L., and J.
242 H. conducted the field and laboratory works. C.W. E.B. and D.W.L. compiled data from peer-reviewed publications and conducted
243 the modelling. All co-authors interpreted the results.

244 **Acknowledgements.** This work was financially supported by the National Basic Research Program of China (973 program;
245 2014CB954400), the National Natural Science Foundation of China (41601255, 31522010 and 31600358), the Key Research
246 Program of Frontier Sciences, CAS (QYZDB-SSWDQC006), and the Youth Innovation Promotion Association CAS to Chao
247 Wang.

248
249 **Competing interests:** The authors declare that they have no conflict of interest

250 **References**

- 251 Accoe, F., Boeckx, P., Cleemput, O. V., Hofman, G., Zhang, Y., and Guanxiong, C.: Evolution of the $\delta^{13}\text{C}$
252 signature related to total carbon contents and carbon decomposition rate constants in a soil profile under
253 grassland, *Rapid Commun. Mass Sp.*, 16, 2184-2189, 2002.
- 254 Acton, P., Fox, J., Campbell, E., Rowe, H., and Wilkinson, M.: Carbon isotopes for estimating soil
255 decomposition and physical mixing in well-drained forest soils, *J. Geophys. Res.: Biogeosci.*, 118, 1532-
256 1545, 2013.
- 257 Berg, B.: Litter decomposition and organic matter turnover in northern forest soils, *Forest Ecol. Manag.*, 133,
258 13-22, 2000.
- 259 Bird, M. I., Chivas, A. R., and Head, J.: A latitudinal gradient in carbon turnover times in forest soils, *Nature*,
260 381, 143-146, 1996.
- 261 Bond - Lamberty, B., Wang, C., and Gower, S. T.: A global relationship between the heterotrophic and
262 autotrophic components of soil respiration?, *Glob. Change Biol.*, 10, 1756-1766, 2004.
- 263 Boström, B., Comstedt, D., and Ekblad, A.: Isotope fractionation and ^{13}C enrichment in soil profiles during
264 the decomposition of soil organic matter, *Oecologia*, 153, 89-98, 2007.
- 265 Bradford, M. A., Berg, B., Maynard, D. S., Wieder, W. R., Wood, S. A., and Cornwell, W.: Understanding the
266 dominant controls on litter decomposition, *J. Ecol.*, 104, 229-238, 2016.
- 267 Brunn, M., Spielvogel, S., Sauer, T., and Oelmann, Y.: Temperature and precipitation effects on $\delta^{13}\text{C}$ depth
268 profiles in SOM under temperate beech forests, *Geoderma*, 235-236, 146-153, 2014.
- 269 Brunn, M., Condrón, L., Wells, A., Spielvogel, S., and Oelmann, Y.: Vertical distribution of carbon and
270 nitrogen stable isotope ratios in topsoils across a temperate rainforest dune chronosequence in New Zealand,
271 *Biogeochemistry*, 129, 37-51, 2016.
- 272 Carvalhais, N., Forkel, M., Khomik, M., Bellarby, J., Jung, M., Migliavacca, M., u, M., Saatchi, S., Santoro,
273 M., Thurner, M., Weber, U., Ahrens, B., Beer, C., Cescatti, A., Randerson, J. T., and Reichstein, M.: Global
274 covariation of carbon turnover times with climate in terrestrial ecosystems, *Nature*, 514, 213-217, 2014.
- 275 Chen, S., Huang, Y., Zou, J., and Shi, Y.: Mean residence time of global topsoil organic carbon depends on
276 temperature, precipitation and soil nitrogen, *Global Planet. Change*, 100, 99-108, 2013.
- 277 Ciais, P., Sabine, C., Bala, G., Bopp, L., Brovkin, V., Canadell, J., Chhabra, A., DeFries, R., Galloway, J.,
278 and Heimann, M.: Carbon and other biogeochemical cycles, in: *Climate Change 2013: The Physical Science*
279 *Basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on*
280 *Climate Change*, Cambridge University Press, 465-570, 2014.
- 281 Curiel, Y., J., Baldocchi, D. D., Gershenson, A., Goldstein, A., Misson, L., and Wong, S.: Microbial soil
282 respiration and its dependency on carbon inputs, soil temperature and moisture, *Glob. Change Biol.*, 13,
283 2018-2035, 2007.
- 284 Davidson, E. A., and Janssens, I. A.: Temperature sensitivity of soil carbon decomposition and feedbacks to
285 climate change, *Nature*, 440, 165-173, 2006.

- 286 Diochon, A., and Kellman, L.: Natural abundance measurements of ^{13}C indicate increased deep soil carbon
287 mineralization after forest disturbance, *Geophys. Res. Lett.*, 35, L14402, 2008.
- 288 Ehleringer, J. R., Buchmann, N., and Flanagan, L. B.: Carbon isotope ratios in belowground carbon cycle
289 processes, *Ecol. Appl.*, 10, 412-422, 2000.
- 290 Fierer, N., Strickland, M. S., Liptzin, D., Bradford, M. A., and Cleveland, C. C.: Global patterns in
291 belowground communities, *Ecol. Lett.*, 12, 1238-1249, 2009.
- 292 Fioretto, A., Papa, S., Pellegrino, A., and Fuggi, A.: Decomposition dynamics of *Myrtus communis* and
293 *Quercus ilex* leaf litter: Mass loss, microbial activity and quality change, *Appl. Soil Ecol.*, 36, 32-40, 2007.
- 294 Friedli, H., Siegenthaler, U., Rauber, D., and Oeschger, H.: Measurements of concentration, $^{13}\text{C}/^{12}\text{C}$ and
295 $^{18}\text{O}/^{16}\text{O}$ ratios of tropospheric carbon dioxide over Switzerland, *Tellus B*, 39, 80-88, 1987.
- 296 Friedlingstein, P., Cox, P., Betts, R., Bopp, L., Von Bloh, W., Brovkin, V., Cadule, P., Doney, S., Eby, M., and
297 Fung, I.: Climate-carbon cycle feedback analysis: Results from the C4MIP model intercomparison, *J. Climate*,
298 19, 3337-3353, 2006.
- 299 Garten, C. T., Cooper, L. W., Post III, W., and Hanson, P. J.: Climate controls on forest soil C isotope ratios
300 in the southern Appalachian Mountains, *Ecology*, 81, 1108-1119, 2000.
- 301 Garten, C. T., and Hanson, P. J.: Measured forest soil C stocks and estimated turnover times along an elevation
302 gradient, *Geoderma*, 136, 342-352, 2006.
- 303 Giardina, C. P., and Ryan, M. G.: Evidence that decomposition rates of organic carbon in mineral soil do not
304 vary with temperature, *Nature*, 404, 858-861, 2000.
- 305 Gill, R. A., and Jackson, R. B.: Global patterns of root turnover for terrestrial ecosystems, *New phytol.*, 147,
306 13-31, 2000.
- 307 IGBP-DIS (1998) SoilData (V.0). A program for creating global soil-property databases, IGBP Global Soils
308 Data Task, France.
- 309 Kaiser, K., Guggenberger, G., and Zech, W.: Isotopic fractionation of dissolved organic carbon in shallow
310 forest soils as affected by sorption, *Eur. J. Soil Sci.*, 52, 585-597, 2001.
- 311 Keiser, A. D., and Bradford, M. A.: Climate masks decomposer influence in a cross-site litter decomposition
312 study, *Soil Biol. Biochem.*, 107, 180-187, 2017.
- 313 Kohl, L., Laganière, J., Edwards, K. A., Billings, S. A., Morrill, P. L., Van Biesen, G., and Ziegler, S. E.:
314 Distinct fungal and bacterial $\delta^{13}\text{C}$ signatures as potential drivers of increasing $\delta^{13}\text{C}$ of soil organic matter
315 with depth, *Biogeochemistry*, 124, 13-26, 2015.
- 316 Lehmann, J., and Kleber, M.: The contentious nature of soil organic matter, *Nature*, 528, 60-68, 2015.
- 317 Nishina, K., Ito, A., Beerling, D., Cadule, P., Ciais, P., Clark, D., Falloon, P., Friend, A., Kahana, R., and Kato,
318 E.: Quantifying uncertainties in soil carbon responses to changes in global mean temperature and
319 precipitation, *Earth Syst. Dynam.*, 5, 197-209, 2014.
- 320 Powers, J. S., and Schlesinger, W. H.: Geographic and vertical patterns of stable carbon isotopes in tropical
321 rain forest soils of Costa Rica, *Geoderma*, 109, 141-160, 2002.

- 322 Raich, J. W., Potter, C. S., and Bhagawati, D.: Interannual variability in global soil respiration, 1980–94,
323 *Glob. Change Biol.*, 8, 800-812, 2002.
- 324 Reay, D. S., Dentener, F., Smith, P., Grace, J., and Feely, R. A.: Global nitrogen deposition and carbon sinks,
325 *Nat. Geosci.*, 1, 430-437, 2008.
- 326 Reichstein, M., Bahn, M., Ciais, P., Frank, D., Mahecha, M. D., Seneviratne, S. I., Zscheischler, J., Beer, C.,
327 Buchmann, N., Frank, D. C., Papale, D., Rammig, A., Smith, P., Thonicke, K., van der Velde, M., Vicca, S.,
328 Walz, A., and Wattenbach, M.: Climate extremes and the carbon cycle, *Nature*, 500, 287-295, 2013.
- 329 Sanderman, J., Amundson, R. G., and Baldocchi, D. D.: Application of eddy covariance measurements to the
330 temperature dependence of soil organic matter mean residence time, *Global biogeochem. cycles*, 17,
331 GB001833, 2003.
- 332 Scharlemann, J. P., Tanner, E. V., Hiederer, R., and Kapos, V.: Global soil carbon: understanding and
333 managing the largest terrestrial carbon pool, *Carbon Manag.*, 5, 81-91, 2014.
- 334 Schimel, D. S., Braswell, B., Holland, E. A., McKeown, R., Ojima, D., Painter, T. H., Parton, W. J., and
335 Townsend, A. R.: Climatic, edaphic, and biotic controls over storage and turnover of carbon in soils, *Global*
336 *biogeochem. cycles*, 8, 279-293, 1994.
- 337 Schmidt, M. W., Torn, M. S., Abiven, S., Dittmar, T., Guggenberger, G., Janssens, I. A., Kleber, M., Kögel-
338 Knabner, I., Lehmann, J., and Manning, D. A.: Persistence of soil organic matter as an ecosystem property,
339 *Nature*, 478, 49-56, 2011.
- 340 Schuur, E. A.: The effect of water on decomposition dynamics in mesic to wet Hawaiian montane forests,
341 *Ecosystems*, 4, 259-273, 2001.
- 342 Silver, W., and Miya, R.: Global patterns in root decomposition: comparisons of climate and litter quality
343 effects, *Oecologia*, 129, 407-419, 2001.
- 344 Tian, H., Lu, C., Yang, J., Banger, K., Huntzinger, D. N., Schwalm, C. R., Michalak, A. M., Cook, R., Ciais,
345 P., Hayes, D., Huang, M., Ito, A., Jain, A. K., Lei, H., Mao, J., Pan, S., Post, W. M., Peng, S., Poulter, B., Ren,
346 W., Ricciuto, D., Schaefer, K., Shi, X., Tao, B., Wang, W., Wei, Y., Yang, Q., Zhang, B., and Zeng, N.: Global
347 patterns and controls of soil organic carbon dynamics as simulated by multiple terrestrial biosphere models:
348 Current status and future directions, *Global biogeochem. cycles*, 29, 775-792, 2015.
- 349 Torn, M. S., Lapenis, A. G., Timofeev, A., Fischer, M. L., Babikov, B. V., and Harden, J. W.: Organic carbon
350 and carbon isotopes in modern and 100 year old soil archives of the Russian steppe, *Glob. Change Biol.*, 8,
351 941-953, 2002.
- 352 Trumbore, S. E.: Comparison of carbon dynamics in tropical and temperate soils using radiocarbon
353 measurements, *Global biogeochem. cycles*, 7, 275-290, 1993.
- 354 Trumbore, S. E., Chadwick, O. A., and Amundson, R.: Rapid Exchange Between Soil Carbon and
355 Atmospheric Carbon Dioxide Driven by Temperature Change, *Science*, 272, 393-396, 1996.
- 356 Vogel, C., Mueller, C. W., Höschen, C., Buegger, F., Heister, K., Schulz, S., Schloter, M., and Kögel-Knabner,
357 I.: Submicron structures provide preferential spots for carbon and nitrogen sequestration in soils, *Nat.*
358 *Commun.*, 5, 2947, 2014.
- 359 Wang, C., Wei, H., Liu, D., Luo, W., Hou, J., Cheng, W., Han, X., and Bai, E.: Depth profiles of soil carbon

- 360 isotopes along a semi-arid grassland transect in northern China, *Plant Soil*, 417, 43-52, 2017.
- 361 Wynn, J. G., Harden, J. W., and Fries, T. L.: Stable carbon isotope depth profiles and soil organic carbon
362 dynamics in the lower Mississippi Basin, *Geoderma*, 131, 89-109, 2006.
- 363 Wynn, J. G.: Carbon isotope fractionation during decomposition of organic matter in soils and paleosols:
364 implications for paleoecological interpretations of paleosols, *Palaeogeogr. Palaeoclimatol. Palaeoecol.*, 251,
365 437-448, 2007.
- 366 Xu, X., Shi, Z., Li, D., Rey, A., Ruan, H., Craine, J. M., Liang, J., Zhou, J., and Luo, Y.: Soil properties
367 control decomposition of soil organic carbon: Results from data-assimilation analysis, *Geoderma*, 262, 235-
368 242, 2016.
- 369 Zhang, D., Hui, D., Luo, Y., and Zhou, G.: Rates of litter decomposition in terrestrial ecosystems: global
370 patterns and controlling factors, *J. Plant Ecol.*, 1, 85-93, 2008.

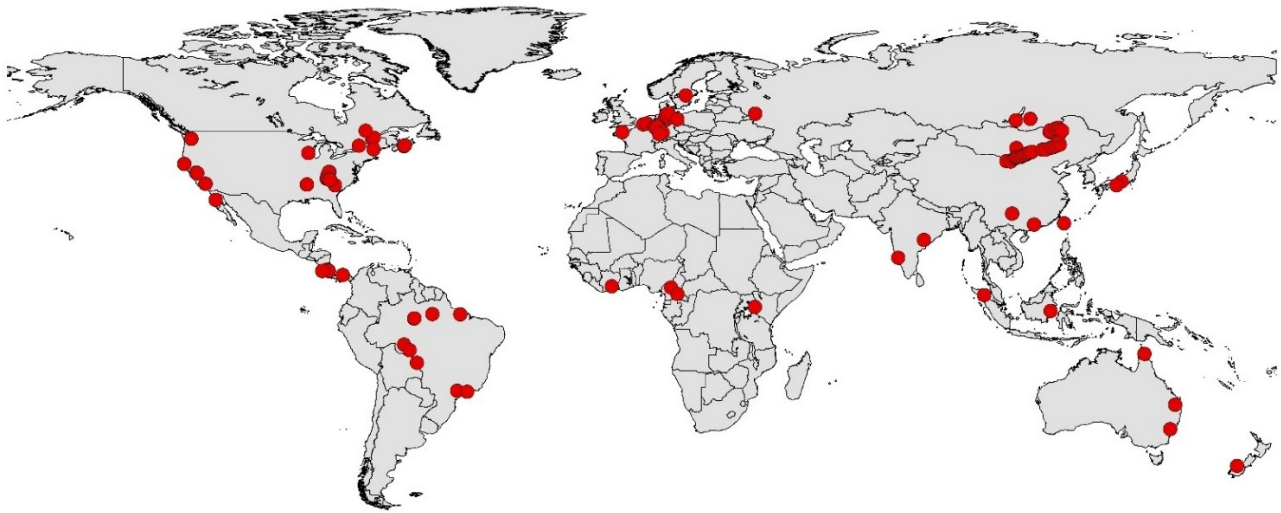
371 **Table 1:** Multiple regressions between $\ln(-\beta)$ and climate and other edaphic factors at global scale.

Variables	R^2	n	AIC
$\ln(-\beta) = 0.046 \text{ MAT} - 0.058 \text{ MAP} + 0.225$	0.43 ^{***}	176	167.30
$\ln(-\beta) = 0.053 \text{ MAT} - 0.137 \text{ MAP} + 0.384 \text{ N} + 0.150$	0.55 ^{***}	104	64.71
$\ln(-\beta) = 0.046 \text{ MAT} - 0.222 \text{ MAP} + 0.849 \text{ N} + 0.006 \text{ Clay} + 0.069$	0.66 ^{***}	70	28.95

372 MAT: Mean Annual Temperature (°C); MAP: Mean Annual Precipitation (m); N: Soil nitrogen concentration (%) in topsoil. Clay:

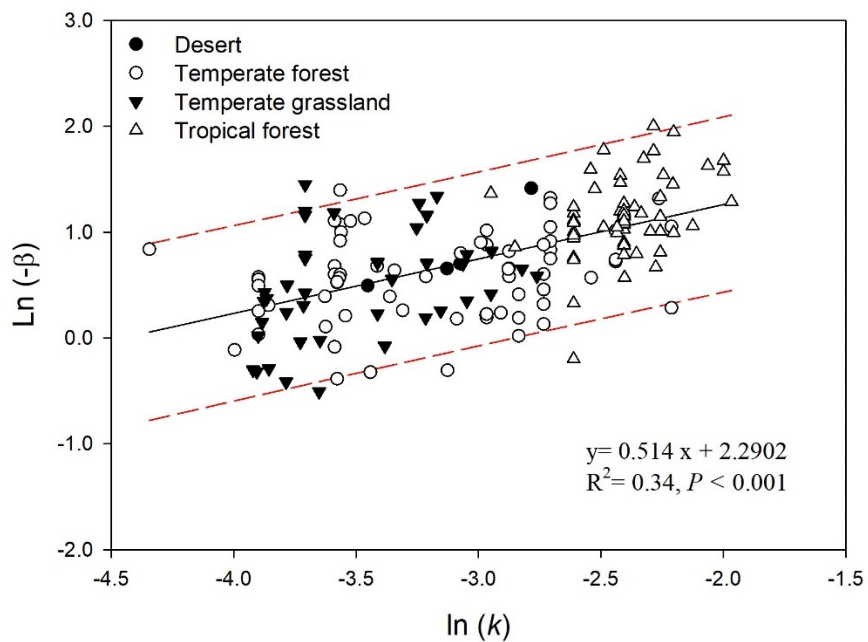
373 Soil clay concentration (%) in topsoil. n is the number of data, and R^2 is the coefficient of determination for the regression line.

374 AIC: Akaike information criterion. *** represents significant at p less than 0.001.



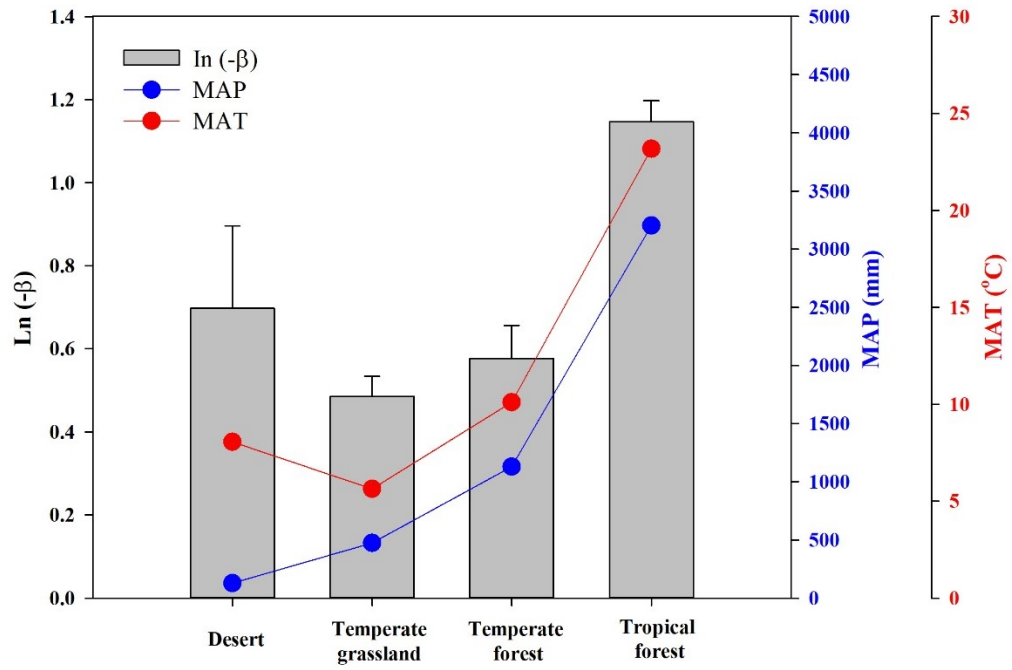
375

376 **Figure 1.** Locations of the 177 soil profiles used to calculate β values in this study.



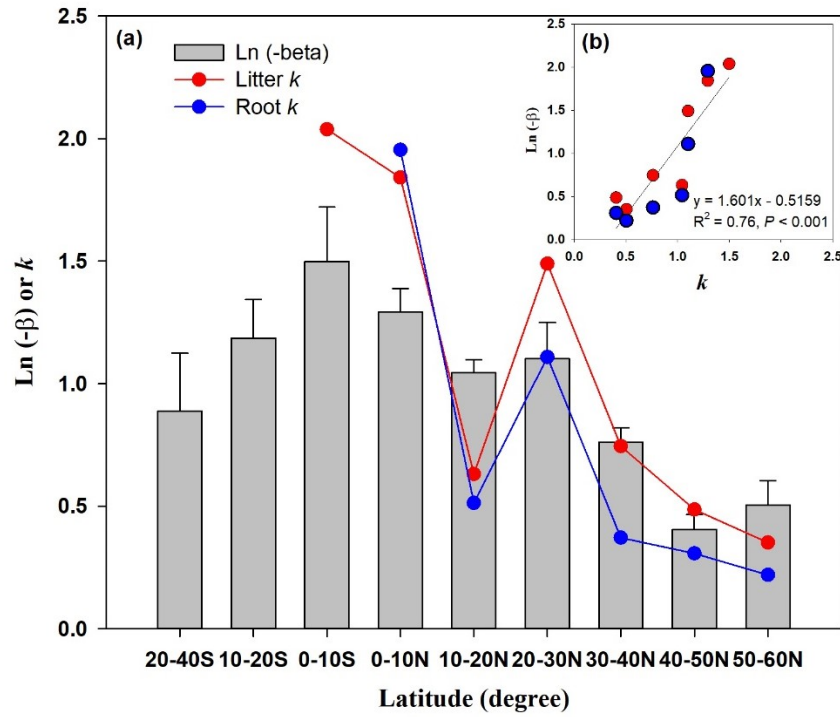
377

378 **Figure 2.** Link between β value and modeled soil carbon turnover rate (k), which was estimated as the ratio between
 379 soil heterotrophic respiration and soil carbon stocks. Solid line is regression line and dashed lines denote 95%
 380 prediction interval.



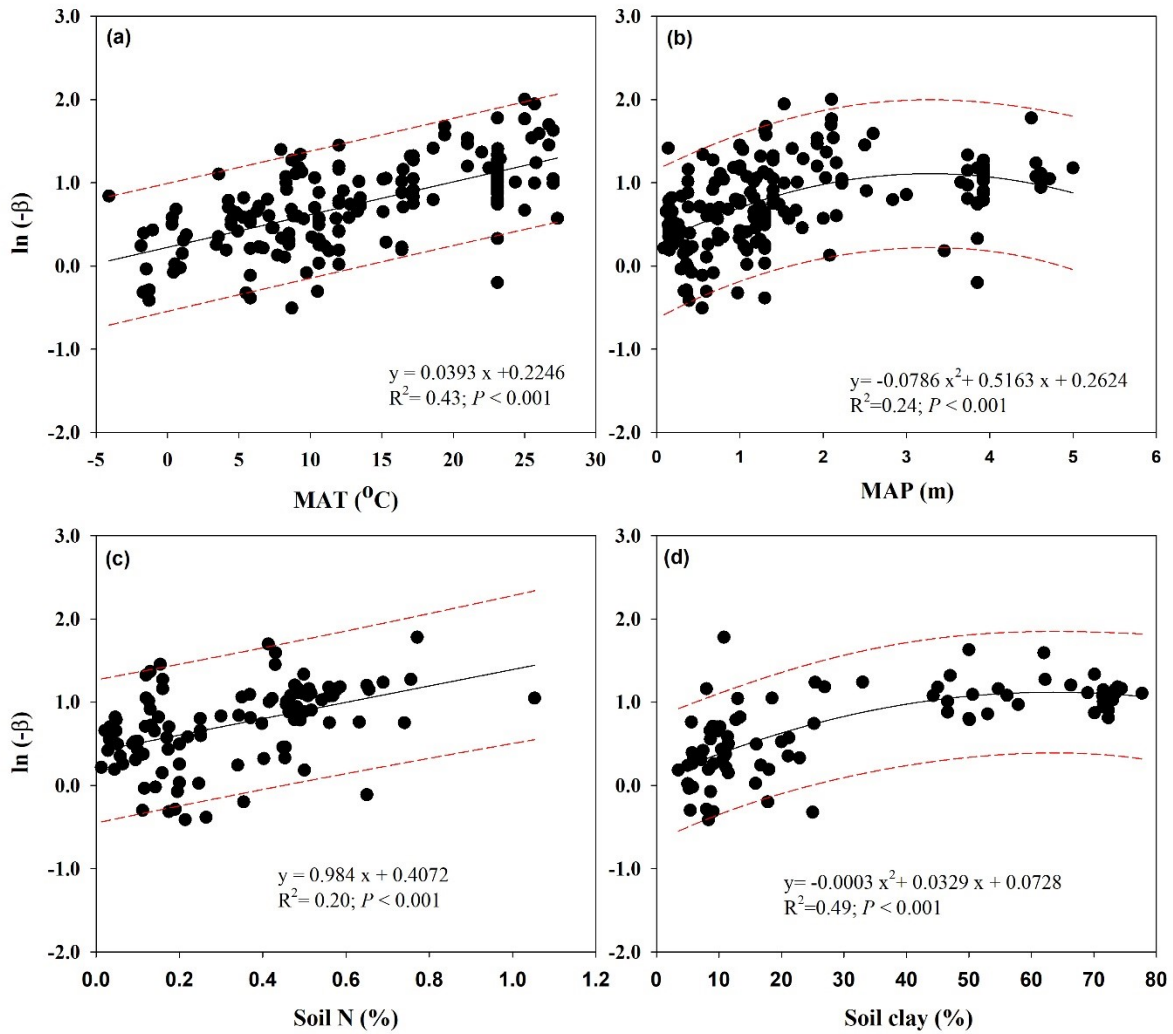
381

382 **Figure 3.** Variations of mean β with biome types. Blue and red points present MAP and MAT for each biome,
 383 respectively.



384

385 **Figure 4.** Variation of mean β value across latitude (bar chart), litter decomposition rate k (yr^{-1} , red dots, Zhang et al.,
 386 2008) and root decomposition rate k (yr^{-1} , blue dots, Silver and Miya, 2001) at the global scale. The inner panel is the
 387 regression between soil β value and litter and root decomposition rate k .



388

389 **Figure 5.** Beta varied with climatic and edaphic factors. Relationships between $\ln(-\beta)$ and MAT (a), MAP (b), soil N
 390 (c), and clay concentration (d) for global dataset. Solid line is regression line and dashed lines denote 95% prediction
 391 interval.