Temperature and moisture are minor drivers of regional-scale soil organic carbon dynamics

B. González-Domínguez1,2, P. A. Niklaus2, M. S. Studer1, F. Hagedorn1, L. Wacker4, N. Haghipour4,5, S. Zimmermann3, L. Walthert4, C. McIntyre4,5,6 & S. Abiven1

Storing large amounts of organic carbon, soils are a key but uncertain component of the global carbon cycle, and accordingly, of Earth System Models (ESMs). Soil organic carbon (SOC) dynamics are regulated by a complex interplay of drivers. Climate, generally represented by temperature and moisture, is regarded as one of the fundamental controls. Here, we use 54 forest sites in Switzerland, systematically selected to span near-independent gradients in temperature and moisture, to disentangle the effects of climate, soil properties, and landform on SOC dynamics. We estimated two SOC turnover times, based on bulk soil 14C measurements (τ14C) and on a 6-month laboratory soil incubation (τi). In addition, upon incubation, we measured the 14C signature of the CO2 evolved and quantified the cumulated production of dissolved organic carbon (DOC). Our results demonstrate that τi and τ14C capture the dynamics of contrasting fractions of the SOC continuum. The 14C-based τ14C primarily reflects the dynamics of an older, stabilised pool, whereas the incubation-based τi mainly captures fresh readily available SOC. Mean site temperature did not raise as a critical driver of SOC dynamics, and site moisture was only significant for τi. However, soil pH emerged as a key control of both turnover times. The production of DOC was independent of τi and not driven by climate, but primarily by the content of clay and, secondarily by the slope of the site. At the regional scale, soil physicochemical properties and landform appear to override the effect of climate on SOC dynamics.

Earth System Models (ESMs) have become a primary tool to project future climate1. Given that soil organic carbon (SOC) is one of the largest terrestrial C pool (~3000 Pg C)2; and relevant to terrestrial carbon-climate feedbacks, the adequate representation of soils is critical for these models. SOC dynamics is commonly implemented in ESMs as a function of climate3,4, typically represented by the interplay of temperature and moisture. Soil incubations and short-term (<10 years) field studies have generally indicated that SOC reservoirs are vulnerable to warming5,6. Moisture has also been recognized as important for predicting the evolution of SOC stocks and fluxes, but its influence has been less thoroughly researched compared to temperature7,8. The evaluation of well-established ESMs has shown that they yield widely different projections of SOC stocks and dynamics and that their results poorly fit observations9,10. A major cause of uncertainties relates to the global carbon cycle9 and it has been suggested that they could be partly reduced if the simulation of SOC turnover was improved9,10. For this, the dependences of SOC dynamics on key drivers (e.g. climate, clay content) should be better understood and implemented in carbon models10,11.

A common limitation of observational studies on this topic is the inherent covariation of drivers in time and/or space (e.g. site temperature and moisture, precipitation and soil pH), as this covariation across natural gradients makes it very challenging to disentangle the effect of drivers12-14. The complex orography of Switzerland,
with the Alps separating the northern and southern regions, generates a large diversity of climates ranging from artic to mediterranean. To unravel the regional-scale effect of temperature and moisture on SOC dynamics, we developed a statistics-based strategy to select 54 study sites spread over Switzerland with maximized orthogonality of temperature and moisture (Fig. 1). We collected field data, ran a laboratory soil incubation and measured $^{14}$C contents to test concurrently various putative drivers of SOC dynamics. Our main hypothesis was that climate plays a primary role, followed by soil properties and landform.

We selected the 54 study sites from the forest soils database of the Swiss Federal Institute for Snow, Forest and Landscape Research (WSL) containing information on 1,050 profiles\(^{15}\) (Figs 1 and S4). To represent climate, we...
included site temperature (1981–2010 average air mean monthly temperature) and moisture (1981–2010 dryness index, details in Methods). For soil properties and landform, we gained pH, clay content (%), slope (%) and orientation (N-E-S-W) data and reduced them to two orthogonal variables (PCo1, PCo2) by principal coordinate analysis. As the Swiss biogeographical regions39 (Fig. 1) differ in biogeophysical characteristics that may also affect SOC stocks and turnover rates, we fitted them as an independent block factor during data analysis. The 54 study sites were selected so that the effects of biogeographic region, temperature, moisture, and the aggregated soil properties and landform-related variables (PCo1, PCo2) were near-orthogonal, i.e. their effects were independent in the multiple linear regression models summarized by analysis of variance17,18 (details in Methods).

From each of the 54 sites, we collected three mineral soil samples from three non-overlapping areas of a plot (~1,600 m²). Each of these samples was a composite obtained by mixing eight 0–20 cm depth soil cores. We determined the bulk soil radiocarbon signature by accelerator mass spectrometry39 on a mixture of material from these three bulk soil composites. Further, we incubated fresh soil samples (sieved ≤2 mm) in microlysimeters, which permitted the flushing of the soils during the incubation to yield water extracts35. For 181 days, we repeatedly measured SOC mineralisation as the evolution of CO₂ and the production of dissolved organic carbon (DOC) by determining organic carbon in soil water extracts. We measured 14CO₂ evolved upon incubation by accelerator mass spectrometry. We ran the incubation in the dark at constant temperature and moisture (25°C and water potential of −20 kPa).

SOC dynamics reflects the balance between the long-term above and belowground organic carbon (OC) inputs, as well as SOC loss processes. Assuming system homogeneity, we calculated two separate estimates of turnover time and compared them. First, bulk soil 14C measurements were used to calculate turnover time τ14C, based on near steady-state conditions (i.e. inputs ≈ outputs) and not accounting for lag times38. Second, SOC mineralisation data from the laboratory soil incubation was used to estimate turnover time τi as the ratio of total SOC stock and output flux32 (details in Methods). As the dissolution of organic substrates is a requirement for their transfer across microbial membranes31, sorption to minerals24,25, and large-scale transport across ecosystems and export from soils26,27, DOC constitutes a small but decisive component of the SOC cycle. Thus, we also quantified the DOC flux as the cumulative production throughout the 181-day incubation.

Results and Discussion

Turnover of contrasting ‘fractions’ of the SOC continuum. Our data show that the 14C-based τ14C is at least one order of magnitude larger than the incubation-based estimate τi (Fig. 2), reinforcing the marked discrepancy found between both calculation methods in a review by Feng et al.38. The 14C-based τ14C ranges from 67 to 511 years and represents the average turnover time of the mixture of all OC molecules that constitute bulk SOC39. In general, τ14C reflects SOC cycling on time scales ranging from years to millennia, which in mineral soils typically constitutes most of the stock50 and where SOC stabilisation processes prevail. As the terrestrial carbon cycle does not only respond to a mean climate, but also to a time series of weather conditions and extreme events31–33, the long-term nature of τ14C makes this metric suited to capture the cumulative effect of such events. On the other side, the incubation-based τi based on the OC output flux over the course of the incubation, ranges from 5 to 29 years, in agreement with the radiocarbon signature of the CO₂ evolved, which carries the atmospheric 14C signature of the last ~5–20 years (Fig. 3). Therefore, τi, primarily but not exclusively, reflects plant-derived readily available OC31 not yet stabilised by mineral interactions or occlusion within aggregates31. In addition, our estimates of τ14C and τi comply with the concepts of mean age and mean transit time, respectively.

Sierra et al. (2016) presented these metrics to contribute to resolving the misconception in literature between the terms turnover time, age and residence time. Further, the discrepancy between the radiocarbon signature of the bulk soil and of the CO₂ evolved (Fig. 3) evidences the heterogeneity of the soil systems. In fact, in a truly homogeneous system τ14C would equal τi (i.e. mean age = mean transit time). Quantitatively, during the 181-day incubation, soils lost 1–9% OC as CO₂ and DOC, indicating that under perturbation a labile fraction of SOC will be potentially lost rapidly.

Todd-Brown et al.38 simulated global SOC turnover for the period 1995–2005 with ESMs part of the Coupled Model Intercomparison Project phase 5 (CMIP5). As many of these models do not explicitly report the depth of OC in the soil profile, the authors assumed that simulations referred to the top 1 m. Their calculated turnover times ranged between ~11 and 37 years, shorter than our estimates of τ14C (67–511 years), but matching those of τi (5–29 years). As our samples correspond to the top 20 cm of mineral soil, and τi is based on laboratory soil incubations, our results suggest that at least some ESMs focus on the turnover of the most labile SOC.

Radiocarbon and incubation-based turnover times τ14C and τi were not strongly correlated (Pearson = 0.40, p < 0.01; Fig. 2), suggesting that the OC fractions these estimates reflect may be influenced by different drivers of SOC dynamics. Lastly, part of this discrepancy may be also related to the presence of pyrogenic organic carbon (PyOC). For example, data on PyOC by hydrogen pyrolysis on bulk soil samples of the 54 sites39, show that the two extreme points, with large τ14C (~380 years) but low τi (~6 years) (Fig. 2), lie at the highest end of the range of probability distribution of PyOC (decile 7th and 10th, Fig S1). When we exclude these two sites from the analysis, the overall trend approaches the curvilinear relationship between mean residence time determined by 14C dating and by a 13C C3–C4 crop experiment in a study by Paul et al.37. The authors of this study concluded that the differences between the two turnover time estimations were because they respectively reflected the dynamics of a slow versus an active pool.

Regional drivers of SOC turnover times. We did not find an effect of site temperature on τ14C (Tables 1 and S1). Field warming experiments show a direct relationship between temperature and the short-term turnover of SOC, but there is evidence that this increase is ephemeral and that with sustained warming, the CO₂ loss from soil tends to decline to ‘prior warming’ levels within a few years (≥10 years)38–40. Several mechanisms can explain thermal acclimation: (1) the depletion of microbially accessible OC pools together with the low sensitivity
of an older pool; (2) increased OC inputs and outputs; and (3) the thermal acclimation of the microbial community (i.e. physiology, biomass, community composition) and enzymatic activities. As for temperature, site moisture did not significantly influence \( \tau_{14C} \) (Tables 1 and S1), indicating that acclimation might have also occurred, or that, over long timescales, \( \tau_{14C} \) was not differently affected by the moisture regimes considered in this study. As such, these results suggest that at the regional scale, the influence of climate on the decadal-centennial turnover of SOC is not apparent, implying that the effect of a changing climate might be similar across climatic gradients.

**Figure 3.** Radiocarbon signatures of bulk soil organic carbon and CO\(_2\) evolved during the last 31 days of a 181-day laboratory soil incubation (0–20 cm depth mineral soil) (Fig. 1). Points are \( \Delta^{14}C \) measured values corrected for 2014, year of field sampling. Uncertainty associated with \( \Delta^{14}C \) bulk soil organic carbon are analytical errors and to \( \Delta^{14}C \) mineralised CO\(_2\) are based on a mixing model assuming constant contamination. Grey discontinuous lines represent atmospheric \(^{14}CO_2\) (North Hemisphere zone 275) and serve as a reference to approximate when the CO\(_2\) evolving from the samples was mostly fixed by vegetation. Values > 0 indicate the presence of Bomb \(^{14}C\); values < 0 indicate that samples have been isolated from atmospheric exchange long enough for considerable radioactive decay to occur and values = 0 indicate that the signature of the samples is equal to the one of the 1890 wood standard which represents pre-industrial atmospheric \(^{14}CO_2\).

**Table 1.** Results of regression analyses to investigate the influence of climate on \( \tau_{14C} \) and DOC. Soil organic carbon (SOC) dynamics: \( \tau_{14C} \) and \( \tau_i \) are respectively the log-transformed \(^{14}C\) and incubation-based turnover times (years). DOC is the cumulative dissolved organic carbon produced during the 181-day incubation relative to the total organic carbon content of the mineral bulk soil (0–20 cm depth) at the beginning of the experiment (g kg\(^{-1}\)OC). Drivers: BGR (biogeographical region), MI (site moisture two-level categorical variable: moist, dry), TI (site temperature two-level categorical variable: warm, cold), MI \( \times \) TI (interaction of moisture and temperature or vice versa), PCo1 and PCo2 (each of the two orthogonal variables obtained by principal coordinates analysis that aggregate soil properties and landform-related variables). Symbols refer to p-values: *** < 0.001, ** < 0.01, * < 0.05, < 0.1, 'n.s.' not significant. Crossed out values indicate that significance is not meaningful because it is not consistent across the two model structures tested for each metric of SOC dynamics. For extended results, refer to Table S1.
Table 2. Results of regression analyses to investigate the influence of soil properties and landform on $\tau_{14C}$, $\tau_i$ and DOC. Soil organic carbon (SOC) dynamics: $\tau_{14C}$ and $\tau_i$ are respectively the log-transformed $^{14}$C and incubation-based turnover times (years). DOC is the cumulative dissolved organic carbon produced during the 181-day incubation relative to the total organic carbon content of the bulk mineral soil (0–20 cm depth) at the beginning of the experiment (g kg$^{-1}$OC). Drivers: pH, clay (%), slope (%) and orientation (degrees). Symbols refer to p-values: ***, ***, **, * < 0.001, ** < 0.01, * < 0.05, < 0.1, 'n.s.' not significant. Crossed out values indicate that significance is not meaningful because it is not consistent across the three model structures tested for each indicator of SOC dynamics. For extended results, refer to Table S2.

| SOC dynamics | ANOVA structure | Significance of drivers |
|--------------|-----------------|------------------------|
|              | pH Clay Slope Orientation |              |
| $\log \tau_{14C}$ | Driver + BGR + MI × TI | * n.s. n.s. n.s. |
|               | BGR + Driver + MI × TI | * n.s. < 0.05 n.s. |
|               | BGR + MI × TI + Driver | * n.s. < 0.05 n.s. |
| $\log \tau_i$ | Driver + BGR + MxTI | *** *** n.s. n.s. |
|               | BGR + Driver + MI × TI | *** *** n.s. n.s. |
|               | BGR + MI × TI + Driver | *** *** n.s. n.s. |
| DOC          | Driver + BGR + MI × TI | *** n.s. n.s. |
|               | BGR + Driver + MI × TI | *** n.s. n.s. |
|               | BGR + MI × TI + Driver | *** n.s. n.s. |

Although $\tau_i$ is assumed to be linked to climate and land cover type$^{46,47}$, and despite it captured the dynamics of the most labile SOC fraction (Fig. 3), it was not significantly explained by site temperature (Tables 1 and S1). Therefore, either temperature plays a limited effect on $\tau_i$, or soils from cold sites are more sensitive to warming than those from warm sites, thus masking the effect of on-site temperature on the CO$_2$ evolved during the incubation$^{48}$. Acclimation processes might have also buffered the response of $\tau_i$ to site temperature; however, as $\tau_i$ averages ~10 years, the time available for acclimation to have occurred is shorter than in the case of $\tau_{14C}$, which averages ~172 years. Overall, these results indicate that at the regional scale, the influence of site temperature on the yearly-decadal turnover of SOC is not evident and that the net effect of warming might be similar across temperature gradients. Furthermore, $\tau_i$ was significantly related to site moisture (Tables 1 and S1) thus, the water balance appears to be more important than temperature for the turnover of the most labile SOC$^{29}$. Our data reveals shorter $\tau_i$ (i.e. higher specific respiration) in dry sites than in moist sites. In the laboratory, when soils were under optimum conditions for decomposition, SOC that had accumulated in the field under dry conditions was mineralised quickly. This pattern evidences that in the field SOC mineralisation was predominantly limited by water availability, and manifests higher sensitivity of enzymes to moisture in dry sites$^{49}$.

The aggregated indicator of soil properties and landform PCo1 explained a significant part of the variation of $\tau_i$ and $\tau_{14C}$ (Tables 1 and S1). Further analyses revealed that soil pH was the underlying driver of $\tau_i$ and, to a lesser extent, of $\tau_{14C}$ (Tables 2, S2 and S3). Soil pH is involved in many processes affecting SOC availability and stabilisation and, it is expected that turnover times become shorter with increasing pH values$^{48}$. The influence of site moisture status on $\tau_i$ but not on $\tau_{14C}$, together with the stronger influence of pH on $\tau_i$ than on $\tau_{14C}$, contribute to explain the lack of correlation between both turnover times (Fig. 2). These results confirm the occurrence of a spatio-temporal gap between the two SOC fractions these turnover times represent and demonstrate that drivers do not equally govern their dynamics (Fig. 4). Additionally, although in temperate forest soils pH is closely linked to the parent material$^{41}$, we observed a significant but weak correlation between water balance and soil pH (Pearson = −0.35, p < 0.05), highlighting the influence of climate on determining soil pH at the global scale$^{52}$.

Regional drivers of DOC leaching. In the incubation assay, the cumulative DOC leached only accounted for ~1% of total OC losses (i.e. CO$_2$ plus DOC). In agreement with results of previous studies, neither site temperature nor moisture emerged as significant drivers of this flux$^{53,54}$ (Tables 1 and S1). Our analyses revealed clay content, pH and site slope as strong controls of DOC production. However, as differences in mineralogy often go along with changes in soil pH$^{55}$, further statistical analyses showed that of the two covarying variables, clay and pH (Pearson = 0.39, p < 0.05), clay was driving the release of DOC (Tables 2, S2 and S3). We found lower DOC losses at higher clay contents, demonstrating the importance of organomineral interactions for the stabilisation of soil organic matter$^{48}$.

With respect to the site slope, even though laboratory experiments are not able to represent many mechanisms that affect DOC fluxes under field conditions, such as the spatio-temporal variability of water transport, we found that steeper sites yielded smaller DOC amounts. We explain this result by a larger water run-off at steeper sites that limits the accumulation of soluble OC in the profile. Finally, despite organic substrates must be in dissolved form for enzymatic degradation, we did not find a correlation between DOC leaching and $\tau_i$ (Fig. S2). Results from other studies have also indicated that DOC production is not directly related to SOC turnover$^{43,34,47}$, probably because only a small fraction of the total DOC is readily bio-available$^{48}$.

To conclude, to our knowledge this is the first study on drivers of SOC dynamics that combines radiocarbon and incubation data adopting an experimental framework that statistically de-correlates the effects of temperature and moisture over a large network of sites. Our results suggest that $^{14}$C and incubation-based turnover times primarily capture the dynamics of an older, stabilised and a younger, readily available SOC fraction, respectively.
Climatic controls did not affect the long-term $^{14}$C-based turnover time, challenging the SOC turnover mechanisms embodied in carbon models in which temperature and moisture are critical for decomposition kinetics. Nevertheless, for the shorter-term incubation-based turnover time, encompassing the decomposition of recently fixed carbon, the legacy effect of site moisture prevailed over the one of temperature. Further, our data revealed a prevailing control of soil pH over both turnover times. The production of DOC was independent of the incubation-based turnover time and neither controlled by climatic drivers, but by clay content. Our data endorses the action of soil physicochemical properties (e.g. pH, clay content) and landform as key drivers of SOC dynamics.

**Methods**

**Selection of study sites.** We followed a stepwise procedure to select the 54 study sites so that the effects of biogeographic region, site temperature, site moisture, and the aggregated soil properties and landform-related variables (PCo1, PCo2) were near-orthogonal (Fig. S3).

A common limitation of studies on the drivers of SOC dynamics is the co-variation of controls in time and/or space. This makes partitioning their relative influences challenging. For example, if we take a set of random samples, it would be very difficult to statistically distinguish between the effect of decreasing temperature and increasing moisture, as in the temperate zone colder sites also tend to be moister. The knowledge-driven selection of study sites for soil research is not new, but its application is still far from common practice. In this work, we developed a statistics-based strategy to select, from the soil database of the Swiss Federal Institute for Forest, Snow and Landscape Research (WSL), a set of 54 sites spread over Switzerland with minimised co-variation of the putative drivers (i.e. maximising orthogonality/statistical independence). The main hypothesis of this study is that sites air temperature and moisture play a primary role in SOC dynamics, followed by soil properties and landform. Accordingly, the stepwise implementation of the selection strategy was performed as follows with the R software (version 3.3.2):

1. **Definition of the exhaustive population within the study area.** By May 2014, the WSL soil database contained data on 1,050 profiles covering the main biogeoregions (i.e. Jura, Midland, Pre-Alps, Alps and Southern Alps) and a vast range of climatic conditions. 3% of the soil profiles in the database belong to the Long-term Forest Ecosystem Research Programme (LWF, German acronym). LWF sites are of special interest due to the sustained and intensive research activities and availability of data. Studies conducted in forested areas have shown that strong topographic gradients are an important control of erosion and deposition patterns. For this reason, to assume that C losses due to erosion are negligible relative to the rate of SOC mineralisation, we excluded sites, excepting LWF sites, with slopes larger than 50% (26.57 degrees). Once we applied this filter, the working database counted 709 entries.

2. **Definition of the putative drivers.** To represent climate, we included sites temperature (1981–2010 average air mean monthly temperature) and moisture (1981–2010 dryness index). The dryness index is a proxy to site moisture status based on the cumulative number of dry months. A dry month is one on which the precipitation of the month is smaller than its potential evapotranspiration (Penman). For soil properties and landform, we reduced the pH value, clay content (%), slope (%) and orientation (N-E-S-W) to two orthogonal variables (PCo1, PCo2) by principal coordinates analysis. For the selection of the sites, all data was sourced from the WSL soil database.

3. **Partition of the population into four equiprobable categories based on main hypothesised drivers.** As we hypothesise that climate plays a primary role in SOC dynamics, we partitioned the population into four

---

Figure 4. Schematic synthesis of the scope of $^{14}$C ($\tau_{^{14}C}$) and incubation-based ($\tau_i$) turnover times. Full colours represent the breadth of this study and patterns the outlook when considering comparable studies on temperate forest topsoils. We illustrate the drivers of soil organic carbon (SOC) dynamics considered in this study and highlight in italic, those that significantly explained turnover times. Stars and arrows' thickness represent the strength of these significances. Significance code of p-values: ***<0.001, **<0.01, *<0.05.
equiprobable categories based on the median value of site temperature and moisture.

4. **Spatial distribution of the population based on biogeographical regions.** Swiss biogeographical regions are the product of a statistical classification based on floristic and faunal distribution patterns, thus they are also connected to geophysical characteristics. To ensure the spatial distribution of the study sites, we subdivided each climatic category upon biogeographical regions. Some combinations of climate-biogeographical region did not occur or were under-represented. For example, we did not find the combination of moist and warm sites in the Alpine region or moist sites in the Southern Alps region.

5. **Partition of the population into four categories based on secondary hypothesised drivers.** We reduced pH, clay (%), slope (%) and orientation (N-E-S-W) into two variables (PCo1 and PCo2) product of a principal coordinates analysis performed on the population dataset (PCoA; also known as classical or metric multidimensional scaling)\(^ {14,64}\). PCoA is an ordination method that attempts to position objects in a Euclidean space of reduced dimensionality while preserving their distance relationships of dissimilarity (or similarity)\(^ {64}\). As a result, in the graphical representation of the Euclidean space, the further are the sites plotted to each other, the more dissimilar they are and vice-versa.

6. **Non-automated selection of study sites from two-dimensional Euclidean spaces defined by PCo1 and PCo2.** To select the 54 sites, we positioned the data from each climate-biogeographical region combination into two-dimensional Euclidean spaces defined by PCo1 and PCo2. From here, we selected equidistant sites from the origin of coordinates. Due to missing points, some combinations were not sampled. These gaps are expected to have an only minor influence on the regression analyses because they are spread rather than clustered over the full range of potential combinations.

**Field campaign.** Between July and September 2014, we collected from each of the 54 sites three mineral soil samples from three non-overlapping areas of a 40 × 40 m² plot immediate to the original WSL soil profile. Each of these samples was a composite product of mixing eight 0–20 cm depth soil cores collected with a 5 cm diameter Humax corer. This sampling strategy enabled us to account for soil spatial variability. After collection, we transported the soils in portable fridges to the laboratories where we sieved them by hand (≤ 2 mm) and stored them at 3.5 °C until further use.

**Laboratory soil incubation.** In February 2015, we incubated fresh soil samples (sieved to ≤ 2 mm; 40 g equivalent dry mass; adjusted to 0.8 g cm\(^{-3}\) bulk density) in sterilised glass microlysimeters\(^ {39}\) which permitted the flushing of the samples during the incubation to yield soil water extracts. Microlysimeters were placed into 2-litre airtight glass jars that contained 20 ml of distilled water in a separate beaker to ensure the headspace was moist. After a 10-day pre-incubation, run under the same conditions as the incubation assay, we measured SOC mineralisation for 181 days while keeping soil moisture (−20 kPa) and temperature (25 °C) constant. We adjusted soil moisture with a suction pump and controlled it periodically by weighing the samples. Our goal was not to estimate the turnover time of soils in the field but to establish comparable conditions close to optimum for microbial activity driving SOC mineralisation. As the laboratory incubation ran at 25 °C, it meant an average temperature increase of 17.3 °C compared to field temperatures (range of temperature increase from 13.2–24.0 °C). We intended to apply this large increase in temperature to override different transient warming responses that might have occurred when warming soil from cold and warm sites\(^ {48}\).

**Radiocarbon measurements.** In this study, we measured the radiocarbon content of (1) bulk soil organic carbon and (2) CO\(_2\)-C evolved from bulk soil. Measurements were performed on a MINI CARBON DATING System (MICADAS, Ionplus, Switzerland)\(^ {19}\) featuring a gas ion source\(^ {67}\) and coupled to an Elemental Analyser (EA vario MICRO cube, Elementar, Germany) at the Laboratory of Ion Beam Physics, ETH Zurich.

1. We measured the radiocarbon signature of bulk soil organic carbon on mixtures of the three composite samples collected at each of the 54 study sites. Prior analysis, we acidified (HCl 37% Trace-Metal purity, 60 °C, 24 h) and neutralized (NaOH pellets, 72 h, 60 °C) soils to remove potential inorganic carbon. Samples were combusted in the EA and the gas generated measured in the MICADAS.

2. We measured the radiocarbon signature of the CO\(_2\)-C evolved on mixtures of the NaOH traps of the three composites containing the CO\(_2\)-C evolved during the last 31 days of the 181-day incubation. We selected the last sampling period of the incubation in order to minimise the effect of soil disturbance as affected by the incubation preparation. After mixing, we acidified the NaOH samples with H\(_2\)PO\(_4\) (85% purity) and measured the gas generated in the MICADAS. Incubated blanks (i.e. jars without soil) were used to correct the radiocarbon signatures and to incorporate a total uncertainty based on a mixing model assuming constant contamination\(^ {68}\).
Radiocarbon values are reported in $\Delta^{14}C$. We corrected the standards OX-I (SRM-4990, NIST) and OX-II (SRM-4990C, NIST) for the decay undergone from 1950 to 2014, year of field sampling.

**$^{14}$C and incubation-based turnover times.** Various observational techniques have been developed to evaluate SOC dynamics and the magnitude and variability of these estimations is largely method dependent\(^{28,37,69,70}\). In fact, one single satisfactory technique does not exist and each of them may be suited to capture the dynamics of SOC at different timescales\(^{69,71}\). Turnover time is a metric of SOC dynamics that at steady-state often refers to the ratio of total carbon stock to input or output flux\(^2\). In a homogeneous system at steady-state, it indicates the time it would take for a reservoir to empty if there were no further inputs\(^2\). When data are not interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pool can lead to misconception. Soils with similar mean turnover time may present distinct distributions of turnover among pools, thus running the risk of being interpreted adequately, the assumption of a single homogeneous pooled
References

1. Luo, Y. et al. Toward more realistic projections of soil carbon dynamics by Earth system models. *Global Biogeochem. Cycles* **30**, 40–56 (2016).

2. Köchy, M., Hiederer, R. & Freibauer, A. Global distribution of soil organic carbon – Part 1: Masses and frequency distributions of SOC stocks for the tropics, permafrost regions, wetlands, and the world. *Soil** 1, 351–365 (2015).

3. Jones, C. D. et al. CMIP and MIPANDash: The Coupled Climate–Carbon Cycle Model Intercomparison Project: experimental protocol for CMIP6. *Geosci. Model Dev.* **9**, 2883–2880 (2016).

4. Piao, S. et al. Net carbon dioxide losses of northern ecosystems in response to autumn warming. *Nature* **451**, 49–52 (2008).

5. Cramer, J. M., Fierer, N. & McLaughlin, K. K. Widespread coupling between the rate and temperature sensitivity of organic matter decay. *Nat. Geosci.* **3**, 854–857 (2010).

6. Lützow, M. & Kögel-Knabner, I. Temperature sensitivity of soil organic matter decomposition—what do we know? *Biol. Fertil. Soils* **46**, 1–15 (2009).

7. Moyano, F. E. et al. The moisture response of soil heterotrophic respiration: interaction with soil properties. *Biogeoosciences* **9**, 1173–1182 (2012).

8. Keiluweit, M., Waniek, T., Kleber, M., Nico, P. & Fendorf, S. Anaerobic microsites have an unaccounted role in soil carbon stabilization. *Nat. Commun.* **8**, 1771 (2017).

9. Friedlingstein, P. et al. Uncertainties in CMIP5 climate projections due to carbon cycle feedbacks. *J. Clim.* **27**, 511–526 (2014).

10. Todd-Brown, K. E. O. et al. Causes of variation in soil carbon simulations from CMIP5 Earth system models and comparison with observations. *Biogeosciences* **10**, 1717–1736 (2013).

11. Doetterl, S. et al. Soil carbon storage controlled by interactions between geochemistry and climate. *Nat. Geosci.* **8** (2015).

12. Davidson, E. A., Belk, E. & Boone, R. D. Soil water content and temperature as independent or confounded factors controlling soil respiration in a temperate mixed hardwood forest. *Glob. Chang. Biol.* **4**, 217–227 (1998).

13. Wu, Z., Dijkstra, P., Koch, G. W., Peñuelas, J. & Hungate, B. A. Responses of terrestrial ecosystems to temperature and precipitation change: a meta-analysis of experimental manipulation. *Glob. Chang. Biol.* **17**, 927–942 (2011).

14. Ciais, P. et al. Europe-wide reduction in primary productivity caused by the heat and drought in 2003. *Nature* **437**, 529–533 (2005).

15. Walthert, L., Graf Pannatier, E. & Meier, E. S. Shortage of nutrients and excess of toxic elements in soils limit the distribution of soil-sensitive tree species in temperate forests. *For. Ecol. Manage.* **297**, 94–107 (2013).

16. Gonneth, Y., Wohlgemuth, T., Sansonnens, B. & Buttler, A. *Die biogeographischen Regionen der Schweiz. Umwelt-materialien documents environment* (2001).

17. Schmid, B., Baruffol, M., Wang, Z. & Niklaus, P. A. A guide to analyzing biodiversity experiments. *J. Plant Ecol.* **10**, 91–110 (2017).

18. Quinn, G. P. & Keough, M. J. *Experimental design and data analysis for biologists.* (Cambridge University Press, 2002).

19. Wacker, L. et al. MICADAS: routine and high-precision radiocarbon dating. *Radiocarbon* **52**, 252–262 (2010).

20. Gonzalez-Dominguez, R., Studer, M. S., Hagedorn, F., Niklaus, P. A. & Ahlven, S. Leaching of soils during laboratory incubations does not affect soil organic carbon mineralisation but solubilisation. *Plos One* **12**, e0174725 (2017).

21. Harkness, D. D., Harrison, A. F. & Bacon, P. J. *The Temporal Distribution of 'Bomb' 14C in a Forest Soil.* *Radiocarbon* **28**, 328–337 (1986).

22. Sierra, C. A., Müller, M., Metzler, H., Manzoni, S. & Trumbore, S. E. The muddle of ages, turnover, transit, and residence times in the carbon cycle. *Glob. Chang. Biol.* https://doi.org/10.1111/gcb.13556 (2016).

23. Marschner, B. & Kalbitz, K. Controls of bioavailability and biodegradability of dissolved organic matter in soils. *Geoderma* **113**, 211–235 (2003).

24. Guggenberger, G. & Kaiser, K. Dissolved organic matter in soil: challenging the paradigm of sorptive preservation. *Geoderma* **113**, 293–310 (2003).

25. Kaiser, K. & Kalbitz, K. Cycling downwards – dissolved organic matter in soils. *Soil Biochem.* **52**, 29–32 (2012).

26. Battin, T. J. et al. Biophysical controls on organic carbon fluxes in fluvial networks. *Nat. Geosci.* **1**, 95–100 (2008).

27. Hagedorn, F., Schleppi, P., Waldner, P. & Flühler, H. Export of Dissolved Organic Carbon and Nitrogen from Gleysol Dominated Catchments: The Significance of Water Flow Paths. *Biogeochemistry* **50**, 137–161 (2000).

28. Feng, W. et al. Methodological uncertainty in estimating carbon turnover times of soil fractions. *Soil Biochem.* **100**, 118–124 (2016).

29. Giardina, C. P., Litton, C. M., Crow, S. E. & Asner, G. P. Warming-related increases in soil CO2 efflux are explained by increased below-ground carbon flux. *Nat. Clim. Chang.* **4**, 822–827 (2014).

30. Trumbore, S. Age of soil organic matter and soil respiration: radiocarbon constraints on belowground C dynamics. *Ecol. Appl.* **10**, 399–411 (2000).

31. Reichstein, M. et al. Climate extremes and the carbon cycle. *Nature* **500**, 287–295 (2013).

32. Heinemann, M. & Reichstein, M. Terrestrial ecosystem carbon dynamics and climate feedbacks. *Nature* **451**, 289–92 (2008).

33. Carvalhais, N. et al. Global covariation of carbon turnover times with climate in terrestrial ecosystems. *Nature*, https://doi.org/10.1038/nature13731 (2014).

34. Trumbore, S. Radiocarbon and Soil Carbon Dynamics. *Annu. Rev. Earth Planet. Sci.* **37**, 47–66 (2009).

35. Schrumpf, M. et al. Storage and stability of organic carbon in soils as related to depth, occlusion within aggregates, and attachment to minerals. *Biogeoosciences* **10**, 1675–1691 (2013).

36. Reisser, M. et al. Drivers of Pyrogenic Carbon content and its long lasting influence on Soil Organic Carbon in forest soils (in prep.) (2018).

37. Paul, E. A., Collins, H. P. & Leavitt, S. W. Dynamics of resistant soil carbon of midwestern agricultural soils measured by naturally occurring 14C abundance. *Geoderma* **104**, 239–256 (2001).

38. Luo, Y., Wan, S., Hui, D. & Wallace, L. L. Acclimatization of soil respiration to warming in a tall grass prairie. *Nature* **413**, 622–625 (2001).

39. Melillo, J. M. et al. Long-term pattern and magnitude of soil carbon feedback to the climate system in a warming world. *Science* **307**, 358, 101–105 (2005).

40. Bradford, M. A. et al. Thermal adaptation of soil microbial respiration to elevated temperature. *Ecol. Lett.* **11**, 1316–1327 (2008).

41. Melillo, J. M. et al. Soil warming and carbon-cycle feedbacks to the climate system. *Science* **307**, 2173–2176 (2002).

42. Elasais, P. E. et al. The response of heterotrophic CO2 flux to soil warming. *Glob. Chang. Biol.* **11**, 167–181 (2005).

43. Knorr, W., Preintze, L., House, J. & Holland, E. Long-term sensitivity of soil carbon turnover to warming. *Nature* **433**, 2003–2006 (2005).

44. Allison, S. D., Wallenstein, M. D. & Bradford, M. A. Soil-carbon response to warming dependent on microbial physiology. *Nat. Geosci.* **3**, 336–340 (2010).

45. Blagodatskaya, E., Blagodatsky, S., Khomyakov, N., Myachina, O. & Kuzyakov, Y. Temperature sensitivity and enzymatic mechanisms of soil organic matter decomposition along an altitudinal gradient on Mount Kilimanjaro. *Sci. Rep.* **6**, 22240 (2016).

46. Zheng, Z.-M. et al. Temperature sensitivity of soil respiration is affected by prevailing climatic conditions and soil organic carbon content: A trans-Asia China based case study. *Soil Biochem.* **41**, 1531–1540 (2009).

47. Poepplau, C. et al. Temporal dynamics of soil organic carbon after land-use change in the temperate zone - carbon response functions as a model approach. *Glob. Chang. Biol.* **17**, 2415–2427 (2011).
48. Koven, C. D., Hugelius, G., Lawrence, D. M. & Wieder, W. R. Higher climatological temperature sensitivity of soil carbon in cold than warm climates. *Nat. Clim. Chang.* 7, 817–822 (2017).
49. Hawkes, C. V., Waring, B. G., Rocca, J. D. & Kivlin, S. N. Historical climate controls soil respiration responses to current soil moisture. *Proc. Natl. Acad. Sci.* 114, 6322–6327 (2017).
50. Leifeld, J. et al. Control of soil pH on turnover of belowground organic matter in subalpine grassland. *Biogeochemistry* 112, 59–69 (2013).
51. Blaser, P., Walthert, L., Zimmermann, S., Graf Pannatier, E. & Luster, J. Classification schemes for the acidity, base saturation, and acidification status of forest soils in Switzerland. *J. Plant Nutr. Soil Sci.* 171, 163–170 (2008).
52. Sanderman, J., Maddern, T. & Baldock, J. Similar composition but differential stability of mineral retained organic matter across four classes of clay minerals. *Biogeochemistry* 121, 409–424 (2014).
53. Neff, J. C. & Hooper, D. U. Vegetation and climate controls on potential CO₂, DOC and DON production in northern latitude soils. *J. Environ. Qual.* 43, 1584–1597 (2014).
54. Mikutta, R. et al. Biogeochemistry of mineral–organic associations across a long-term mineralogical soil gradient (0.3–4100 kyr), Hawaiian Islands. *Geochim. Cosmochim. Acta* 73, 2034–2060 (2009).
55. Six, J. & Jastrow, J. D. Organic matter turnover. *Encycl. Soil Sci.* 53, 325–338 (1966).
56. Wollum, A. & Gomez, J. A conductivity method for measuring microbially evolved carbon dioxide. *Ecology* 51, 155–156 (1970).
57. Addiscott, T. M. & Wagenet, R. J. A Simple Method for Combining Soil Properties that Show Variability. *Soil Sci. Soc. Am. J.* 49, 1365 (1985).
58. R Core Team. R: A language and environment for statistical computing (2016).
59. McCorkle, E. P. et al. Tracing the source of soil organic matter eroded from temperate forest catchments using carbon and nitrogen isotopes. *Chem. Geol.* 445, 172–184 (2016).
60. R Core Team. R: A language and environment for statistical computing (2016).
61. McCorkle, E. P. et al. Tracing the source of soil organic matter eroded from temperate forest catchments using carbon and nitrogen isotopes. *Chem. Geol.* 445, 172–184 (2016).
62. Stacy, E. M., Hart, S. C., Hunsaker, C. T., Johnson, D. W. & Berhe, A. A. Soil carbon and nitrogen erosion in forested catchments: implications for erosion-induced terrestrial carbon sequestration. *Biogeosciences* 12, 4861–4874 (2015).
63. Wollum, A. & Gomez, J. A conductivity method for measuring microbially evolved carbon dioxide. *Ecology* 51, 155–156 (1970).
64. Nadellhofer, K. J. & Fry, B. Controls on natural nitrogen-15 and carbon-13 abundances in forest soil organic matter. *Soil Sci. Soc. Am. J.* 52, 1633–1640 (1988).
65. Ruff, M. et al. A Gas Ion Source for Radiocarbon Measurements at 200-kV. *Radiocarbon* 49, 307–314 (2007).
66. Hanke, U. M. et al. Comprehensive radiocarbon analysis of benzene polycarboxylic acids (BPCAs) derived from pyrogenic carbon in environmental samples. *Radiocarbon* 59, 1103–1116 (2017).
67. Di bona, C. P. & Riesen, M. G. Evidence that decomposition rates of organic carbon in mineral soil do not vary with temperature. *Nature* 404, 858–61 (2000).
68. Tipping, E., Chamberlain, P. M., Bryant, C. L. & Buckingham, S. Soil organic matter turnover in British deciduous woodlands, quantified with radiocarbon. *Geoderma* 155, 10–18 (2010).
69. McFarlane, K. J. et al. Comparison of soil organic matter dynamics at five temperate deciduous forests with physical fractionation and radiocarbon measurements. *Biogeochemistry* 112, 457–476 (2013).
70. Six, J. & Jastrow, J. D. Organic matter turnover. *Encycl. Soil Sci.* 936–942 (2002).
71. Trumbore, S. E. Potential responses of soil organic carbon to global environmental change. *Proc. Natl. Acad. Sci. USA* 94, 8284–8291 (1997).
72. Radiocarbon and climate change - Mechanisms, applications and laboratory techniques. (Springer International Publishing), https://doi.org/10.1007/978-3-319-25643-6 (2016).
73. Torn, M. S., Swanston, C. W., Castanha, C. & Trumbore, S. E. Storage and turnover of organic matter in soil. In *Biogeochemical Processes Involving Natural Nonliving Organic Matter in Environmental Systems* (eds Senesi, N., Xing, B. & Huang, P. M.) 219–272 (Springer International Publishing), https://doi.org/10.1007/978-3-319-25643-6 (2016).
74. Davidson, E. A., Trumbore, S. E. & Amundson, R. Soil warming and organic carbon content. *Nature* 408, 789–790 (2000).
75. Hua, Q., Barbetti, M. & Rakowski, A. Z. Atmospheric Radiocarbon for the Period 1950–2010. *Radiocarbon* 55, 2059–2072 (2013).
76. Remund, J., Rühm, B. & Huguenin-Landl, B. Klimadaten für die Waldmodellierung für das 20. und 21. Jahrhundert. Schlußbericht des Projektes im Forschungsprogramm Wald und Klimawandel. (ETH, Zurich, 2014).
77. Tipping, E., Chamberlain, P. M., Bryant, C. L. & Buckingham, S. Soil organic matter turnover in British deciduous woodlands, quantified with radiocarbon. *Geoderma* 155, 10–18 (2010).
78. McFarlane, K. J. et al. Comparison of soil organic matter dynamics at five temperate deciduous forests with physical fractionation and radiocarbon measurements. *Biogeochemistry* 112, 457–476 (2013).
79. Chen, S., Huang, Y., Zou, J. & Shi, Y. Mean residence time of global topsoil organic carbon depends on temperature, precipitation and soil nitrogen. *Glob. Planet. Change* 100, 99–108 (2013).

**Acknowledgements**
We thank the Swiss National Science Foundation (project-no. 143026) for funding this study in the frame of the National Research Programme 68. (http://www.nrp68.ch/en). We thank the Geography Department of the University of Zurich for their support and the Swiss Federal Institute for Forest, Snow and Landscape Research (WSL) and the Swiss Long-term Forest Ecosystem Research Programme (LWF) for giving us access to the research sites and to the soil database.

**Author Contributions**
Conceptualization: B.G.D., S.A., P.A.N. Data curation: B.G.D., S.Z., L.W. Formal analysis: B.G.D., P.A.N. Funding acquisition: S.A., P.A.N. Investigation: B.G.D. Methodology: B.G.D., S.A. Project administration: B.G.D. Resources: S.A., P.A.N. Software: B.G.D. Supervision: B.G.D., S.A. Validation: B.G.D., S.A., P.A.N. Visualization: B.G.D. Writing – original draft: B.G.D. Writing – review & editing: B.G.D., S.A., M.S.S., P.A.N., N.H., C.M., L.W., S.Z., L.W., F.H.

**Additional Information**
Supplementary information accompanies this paper at https://doi.org/10.1038/s41598-019-42629-5.

**Competing Interests:** The authors declare no competing interests.
