Dear Sébastian,

I would like to thank you very much for supervising the revisions on my paper and for soliciting both excellent reviews from Jerome Balesdent and Reviewer #2. Both reviewers demonstrated to be experts in the field and I very much appreciate their comments. We have addressed them thoroughly and I believe this majorly improved the paper.
Overall, both reviewers were very positive about the dataset and the interpretations which they both felt merited publication. However, a major point of critique was the modelling. We acknowledge that this part of the paper was insufficiently clear, and heavily relied on the commented code which was in the Supplemental. We rectified this section following the suggestions by Mr. Balesdent and Reviewer #2, and subsequently did an appropriate overhaul of this section. We would like to clarify that we merely switched from the manual, iterative. time-consuming optimization in excel (e.g. Herold et al., 2015) to a faster, automated form in Matlab. We did not develop a new model such as the likes of CENTURY or DAYCENT. We also benchmarked all our automated results to the manual excel, and found that they agree. As reviewer #2 stressed, the focus of this paper should be on the exceptional dataset and our revisions reflect this. We have also added all additional requested information such as quantified residual errors after the optimization.

In accordance with the submission guidelines, the code and datasets will be deposited in FAIR-aligned data repositories. For the review process, we have also included the Matlab codes for the reviewers in the supplement.

Please find detailed replies to comments in our point-by-point replies to the reviewers.

Thank you again very much,

Kind regards on behalf of all my co-authors,

Tessa

Response to Jerome Balesdent

Understanding the dynamics of carbon in deep soil layers is an important issue, and this study uses an excellent sequence and provides a rare dataset: soil 14C measure- ment at two dates using archived samples brings a precious information of C dynamics. One of the interesting results is the demonstration of the occurrence of rock-derived carbon. Another concerns the age of water extractable carbon. The analytical meth- ods are high standard and highly relevant. I therefore consider it is worth publishing the data in Biogeochemistry. Unfortunately, there are major concerns that need revi- sion. The most important is that the mathematical and numerical interpretations look inappropriate, and this leads the authors to give conclusions that are in contrast with what the data show, whereas some unprecedented results could be derived. I finally suggest two alternative solutions: either the authors drop the modelling part and make a semi-quantitative interpretation of the data, either they use another model. I also noticed miscellaneous improvements to be done. The discussion should be updated according to these major points. The title and summary are nevertheless appropriate.

Dear Prof. Balesdent,

Thank you very much for your positive feedback and thorough review. We very much appreciate that you value the importance of the data for the wider Biogeosciences community. Your comments about the turnover time modelling are also very insightful and the issues have subsequently been addressed. There was indeed a semantic issue which caused problems, so we incorporated all of your feedback. We realized that most of the modelling was explained in the code in the SI, and that therefore the text in the main text was absolutely inadequate in order to explain our calculations. Consequently, paper and especially the discussion was updated according to these major points. As you indicated, the title and summary remained appropriate.

We want to thank you again for your helpful review, which has further improved this paper. Please find detailed replies below.

1. The chosen model is unlikely to simulate observed data.

Most of samples below 10 cm show an increase in Δ14C between 1990's and 2010's, by several 10‰ (Figure 3), and even some above 10 cm do. As seen in the FIGURE below, which was built for this review, the 14C content of well mixed compartments directly fed from atmospheric C has DECREASED with time since the 1990's (or in- creased by less than 4‰ for slow pools). The sum of two parallel pools cannot have a Δ14C increased between 1995 and 2014.

FIGURE: Simulated Δ14C of a well-mixed compartment under steady state as a func- tion of compartment turnover rate, for two dates of sampling.

Thank you for these comments. Indeed, rapidly turning-over compartments have decreased in ^{14}C in the last two decades, whilst the slower compartments have increased in ^{14}C signature (e.g. Figure 3a) (as you also indicate in your comments below). As you indicated in your supplemental, there was indeed a semantic issue with the turnover time definition when estimating the size of the two respective pools, which has now been adapted. Our apologies for the confusion.

I finally understood (from 14C data in Figure 3 and turnover time data in Table S5) that the the "mosty reliable' kWSOC value is more or less the arithmetic mean of two kWSOC values, one calculated in the 1990's and the other in the 2010's. The authors must invoke other processes to explain an increasing Δ14C. These processes may act together and interact:

- Transit of carbon in another horizons or pool before entering the observed layer. This might be associated with either bioturbation or DOC production from an above layer, movement, and insolubilization. The data tend to indicate that carbon movement is a significant cause of the increase in Δ14C across the sequence.

- non-steady state, e.g. increased bioturbation due to warming, change in NPP and/or decay rates.

To me, the fact that the Δ14C of WSOC of all samples (except Othmarsingen 0-5 cm and Lausanne 0-5 cm) inceased is a proof that WSOC is a by-product of SOM aged several 10th of years (usual age of OH horizons), and not directly fed by vegetation decomposition. This would be a bright finding and merit appropriate modelling.

Thank you for these comments as well. We did not sufficiently explain how we estimate the turnover of the WEOC or 'labile' pool using the ^{14}C time-series. We have addressed this now in the method section, by detailing the different steps and the error calculation (Equations 1-4). In short, we do not take the arithmetic mean, but rather use the standard equations (e.g. Herold et al., 2014; Torn et al., 2009) to find the likeliest turnover time considering both data points of the time series. Instead of the usual excel-based method, we do this in MatLab because it is automated and more repeatable. The solution which has the lowest calculated residual square root mean error (RSME) is automatically chosen, as opposed to a manual iteration.

Thank you for highlighting the importance of potentially DOC-driven transport of young(er) carbon through the deep soil, we have included this in our discussion. We have also now included your suggestion in Section 4.1.3 to highlight that WEOC is likely not fed by vegetation decomposition but rather is derived from several decades-old SOM.

2. Consistency in model implementation (to be confirmed).

I tried to calculate by myself turnover time values, based on 14C data in Figure 3 and turnover time data in Table S5, and didn't find the author's results. This may arise from the fact that the basic differential equations of the model (equation 5 = SI.7) looks false, or at least do not correspond to authors' hypotheses. Equation SI.7 states:

$F(t) = k \cdot Fatm(t) + m1 \cdot F(t - 1).(1 - \lambda - k1) + m2 \cdot F(t-1).(1 - \lambda - k2)$
This equation indicates that the flux of 14C leaving the system (out of desintegration)

is:

$(m1.k1 + m2.k2).F(t -1)$, i.e., $k.F(t)$

Since the corresponding flux of carbon is $k = m1.k1 + m2.k2$, this equation says that the 14C activity of carbon leaving the system is $F(t – 1)$. So the equation would IM- PLICITELY considers that the activity of the flux out is the same as that of the compart- ment itself. This is typically the assumption of a so-called 'well mixed' compartment, and is not the case of a system with two compartments. It would only accept the solu- tion $k1 = k2$. Making this implicit assumption is a current mistake or at least a source of disagreement in isotope geochemisty. As a consequence, I guess that the authors have calculated a mean turnover time corresponding to a single compartment for bulk carbon, and an independent specific turnover time of WSOC. The error might be linked with my point 3 below. See a proposal for the correct equation as an appendix of this review. The authors are invited to check how eq SI.7 was implemented and how the couple (k2 , m1) was inferred from bulk F14C.

Thank you, there are two main things raised in this comment:

A. Modelling Structure

Indeed, we have calculated a mean turnover time corresponding to a single compartment for bulk carbon, and one independent specific turnover time of WSOC. We have clarified this in the text.

B. Model consistency

Thank you for your suggestions and example for Figure SI.7, we have implemented all of your suggestions (Eq. 6). More details can be found below.

We would like to clarify that we merely transformed the usual excel file people use to find turnover time to MatLab-driven optimization, because it saves time, is repeatable, unbiased and error can be quantified. We have now also quantified all our errors (See SI). Furthermore, the code can easily be used as well for longer time-series (i.e. > 2 timepoints). We benchmarked our results to the Excel-based method, and the results agree.

3. Mathematical (and semantic) misuse of "turnover time'.

Let us call the turnover time of carbon in the compartment $T = 1/k$ Mathematically, the carbon input to the system is $m1/T1 + m2/T2$. The size of the compartment is $m1 + m2$. So, the turnover time, which is the ratio of pool size to the input, is:

$T = (m1 + m2)/(m1/T1 + m2/T2)$

In Table SI.5, which presents the main result, i.e. the values of turnover time, the authors calculated the bulk turnover time as:

$T = (m1.T1 + m2.T2)/(m1 + m2)$, which is wrong.

What authors call "turnover time" is in fact the MEAN AGE of carbon, which is different of the mean turnover time in non-well mixed compartments. The error in not only semantic because it possibly have interfered in model and 14C equation (point 2). Sierra et al. (2016), whom you cite lines 161-162, recommends the use of "age", not "turnover time" for this variable. See also Manzoni et al.(2009).

Indeed, there was a (semantic) inconsistency regarding turnover time between the Main text and the SI, which we have now addressed and corrected. We also implemented your equation. We have the ^{14}C-determined 'turnover time' for the bulk soil, whilst stating that we assume a steady state. We have also clarified our definition in the text, following Manzoni et al. (2009) as well.

4. Data availability.

The authors must provide in SI a table including the primary data, i.e., Δ14C, C stock by horizon, WEOC stocks. Reference that were used to estimate atmospheric Δ14C (post bomb and pre-bomb) should be indicated (e.g. Reimer , Hua etc.)

We have included an excel file with all the raw data regarding $\Delta^{14}C$ and stocks the WEOC material. The WEOC concentration is low (< 1 %) and can be found in SI Table 4. We had indicated the provenance of our pre- and post-bomb data already in the method section, but we have now further clarified it.

5. Hypothesis on WSOC as the labile pool.

Line 180-182 and 190-191: A major (if not the major) assumption of the model is that the dynamic pools has the same decay rate as that of WEOC. The 'dynamic' pools contains as much as 88% of soil C (on the average 34%), whereas WEOC only a few %. Assigning the constant k of WEOC to the dynamic pool is therefore a surprising and very heavy hypothesis. (see also point 1.)

Alternatively, the study may have targetted the study of WSOC dynamics for itself, e.g., considered that both WSOC and bulk C are heterogenous pools, each with a labile and a more stable component, but in varied proportion. Many other models use particulate organic matter (i.e. either sand-size primary organic particles or light OM, which has been described as having a good fit with labile carbon

Yes indeed, it was our assumption that the measured WEOC could be representative of the dynamic pool. There are studies that hypothesize WEOC could be indicative of a larger dynamic pool (Baisden and Parfitt, 2007; Koarashi et al., 2012). But indeed, this is a heavy assumption. We have therefore decreased the importance of the two-pool model in the paper, and highlighted this assumption. Indeed, both the WEOC and bulk themselves can heterogeneous pools, hence we also looked at biomarkers in another study (e.g. Van der Voort et al., 2017, Diverse Soil Carbon Dynamics Expressed at the Molecular Level, GRL). Looking at other fraction would be a worthwhile topic for future work.

6. Conclusions on correlation with MAP.

Projecting conclusions on the effect of MAP on the basis of a "wet" sequence, i.e., where the water deficit is probably low if not nil, may look brash. The driest site is 800 mm, but with a MAT 1.3C and probably a small PET. Furthermore (Lines 360-361), authors state that 'The only climate-related driver which appears to be significant is precipitation' whereas the r2 coefficient between MAP and turnover 0-20 cm is 0.04! I would recommend here to cite Carvalhais et al. (2013) and Mathieu et al. (2015), who highlighted the role of precipitation in SOM stabilization or ecosystem carbon turnover.
I sfinally uggest to mederate the conclusions, but maybe discuss the role of precipitation on DOC movement (see point 1).

Thank you for these insights. As suggested, we have highlighted the role of precipitation as SOM stabilizer and interaction with DOC movement, and tempered our statements about precipitation. Indeed, Switzerland is a wet country! Your own 2018 paper also could show the important role of evapotranspiration but we unfortunately do not have this data. Also, we adapted the phrasing of line 360-61, the role of precipitation is pronounced for the deep soil.

7. Presentation of model and equations.

The presentation of both the model and the optimization process is obscure throughout the text and should be more precise, in either text or SI. In the cases with four radio- carbon dates (2 sampling dates x two fractions), the optimization of three dynamic pa- rameters is not a formal solution, but a best fit.

Indeed, we have now mentioned this specifically in the text.

The type of adjustment (least squares ?) and a criterion of the fit (e.g., RMSE) should be indicated.

This has been included in the main text instead of the SI, we use RSME.

Harmonize the name of variables throughout the text and SI. For consistency with SI, please use m insteaf of F in eqn (3), (4) and (4); and possibly F instead of R. Also use the same character k in SI and main text. Harmonize M (Figure S2) and m, etc.

Thank you, this has now been adjusted.

How were single points managed ? (Line 194-195. " Due to limited availability of archived samples, there are only single time points available for some samples as indicated in Fig. 4.")

This has been clarified in the main text, we solve the standard radiocarbon decay equations (e.g. Torn et al., 2009, $R_{sample,t} = k \times R_{atm,t} + (1 - k - \lambda) \times R_{sample(t-1)}$). This is done more traditionally in Excel, we did the same using a Matlab optimization.

8. Miscellaneous.

lines 51-52 note the pioneer studies by Jenkinson et al (1992) on long-term experi- ments. The models by Braakkeke et al. (2014) also simulates 14C profiles in rather similar podzols, using WSOC as well, and may receive more attention in the discussion section. Also note (e.g. Line 34) the conclusions of Mathieu et al. (2015) concerning soil versus climate drivers of 14C, and (lines 39-40) the recent paper by Balesdent et al. (2018), which improved the understanding of the significance of deep soil C to the global C cycle.

Thank you, I have incorporated these literature suggestions. I had already cited Braakhekke et al.

Move lines 126-128 (WEOC) to the end of 2.1. (WEOC extraction). Note that extraction with Na 0.86 M is not exactly Water extraction, since it moves some exchangeable calcium, disperses clays and therefore moves sorbed organic compounds that would not have been mobilized by water.

Indeed, we followed Hagedorn et al., 2014 when preparing the extraction, and have this stated this clearly in the method section.

Line 252 ’ Deeper soil bulk stock and turnover positively...’ and table S5: avoid "turnover " alone standing for "turnover time" in such sentences, because the common sense of turnover is turnover rate, i.e., the inverse of turnover time. This may lead to a reverse understanding of correlations.

Indeed! We adapted this now.

Line 262. Balesdent et al. (2018) reported that 21% of world subsoil C (30-100 cm) is less than 50 years old.

We have included this.

The amount of WEOC (while not used in the modelling experiment) would be welcome.

We have included this in the SI Table 4. Concentrations are low (< 1%)

Surprisingly, the section of Material and methods indicates that NPP and its compo- nents were measured, which is a rare information in SOM studies. As a result, authors have an indicator of the true turnover time of soil C, i.e. the ratio of Soil C stock to C input is known, that they do not use.

Indeed, there is NPP data, but we were recommended by the field experts that although it was representative for the tree vegetation, we had better not use it for estimating soil flux, as there would be too many assumptions to be considered. We did include the data, so others are free to use it.

Figure 4 contains the main primary result of the study. Polices Should be enlarged. The square signs for Aptal WEOC 1997 are misleading. Table S5 is the main final result and should take place in the main document.

We have adapted this figure slightly. Following your critique about the assumption of using WEOC as a dynamic pool we reduced the importance of the fraction modelling in the paper, so we opted to keep it in the SI.

Note that the bi-exponential age distribution is factually the age distribution of C in current "four pools" models such as RothC (or Century). All coupling of these models with radiocarbon more or less managed bi-exponential age distribution and 14C; e.g., Jenkinson et al. (1992).

Yes, we are familiar with Century (RothC), but feel applying them would be beyond the scope of this paper.

9. Appendix

The differential equation should consider F1 and F2 the 14C fraction in pools 1 and 2, respectively, as illustrated in your Fig S1.

Input flux to pool1 is k1.m1; input flux to pool2 is k2.m2

$F1(t) = k1.Fatm(t) + (1 - k1 - \lambda).F1(t - 1)$ $F2(t) = k2.Fatm(t) + (1 - k2 - \lambda).F2(t - 1)$ which give: $F(t) = m1F1(t) + m2.F2(t) = k.Fatm(t) + m1.(1 - k1 - \lambda).F1(t - 1) + m1.(1 – k2 - \lambda).F2(t - 1)$ And needs numerical resolution of F1 and F2.

Thank you, we implemented this.

10. Cited references Balesdent J., Basile-Doelsch I, Chadoeuf J., Cornu S., Derrien D. Fekiacova Z., Hatté C. Atmosphere-soil carbon transfer as a function of soil depth. Nature, 559, 599–602. (2018) doi.org/10.1038/s41586-018-0328-3

Jenkinson D.S., D.D. Harkness, E.D. Vance, D.E. Adams and A.F. Harrison. Calculating net primary production and annual input of organic matter to soil from the amount and radiocarbon content of soil organic matter. Soil Biol. Biochem. 24(4):295-308 (1992)

Manzoni, S., Katul, G. G. & Porporato, A. Analysis of soil carbon transit times and age distributions using network theories. J. Geophys. Res. 114, G04025 (2009)

Mathieu J., Hatté C., Parent E., Balesdent J. Deep soil carbon dynamics are driven more by soil type than by climate: a worldwide meta-analysis of radiocarbon profiles. Global Change Biology 21, 4278-4292. (2015) doi:10.1111/gcb.13012.

Thank you, we implemented these papers

11. Figure.

Simulated Δ14C of a well-mixed compartment under steady state as a function of compartment turnover rate, for two dates of sampling. Compartment has a single C7

exponential distribution of ages; system start 8050 BP; atmospheric Δ14C after Reimer et al. (2009) and Hua et al. (2013); Northern hemisphere zone N2; May-August.

Interactive comment on Biogeosciences Discuss., https://doi.org/10.5194/bg-2018-361, 2018.

Response to Reviewer #2

This study aims at investigating the dynamics of carbon as a function of soil depth in five sites of the Swiss Alps. To reach this goal the authors realised 14C measurements on samples collected in late 90's and in 2014. Soils were sampled at different depths and a water extractable fraction was extracted. The authors derived C turnover rates from 14C data using a two-pool model. They identify a substantial fraction of fast-cycling C at depth and further investigate potential edaphic and climatic drivers of turnover. The data gathered in this study are of great interest, but at this stage, the manuscript suffers from too severe limitations to be published.

Thank you very much for your positive feedback regarding the quality of the dataset and insights which can be gained from it. You indicated that the main limitation was the two-pool modelling, so we addressed this, details below. We have also addressed the other issues that have been raised. Thanks again for your helpful review, it helped further improve this paper.

In particular, the authors should decide what is precisely their objective: do they want to provide insights on deep C cycling or to offer a new method to compute turn-over time using 14C data? I would suspect the readers of Biogeosciences to be really interested in the first option, as there are only a limited number of studies on this topic (as claimed in l 276 of the discussion).

Thank you for posing this question. Indeed, our objective is to provide insights on deep soil C cycling, and not to develop a new model such as the likes of Century or RothC. We have clarified this and further simplified the modelling. We merely switch from an excel-based manual, iterative, time-consuming optimization with limited error quantification to an automated form in excel with error quantification.

Nevertheless, the data on C turn-over along the soil profile are mainly presented as supplementary, while there is a strong focus on methodological aspects in the main text.

We present the ^{14}C data and ^{14}C turnover data in graphs in the main texts (Figures 3-5), and the raw data can be found the SI. We have augmented our graphs.

The discussion should also be improved. Too many repetitions of the results in 4.1.1 and 4.1.2; 4.1.3 repeats some facts of 4.1.2. 4.2:

Thank you, we of course avoid repetition, we removed the overlapping content. The different sections do refer to the same data, so re-addressing certain patterns is unavoidable.

I could not find clear information in the materials and methods section about how the data supporting this section were collected.

Thank you, actually in sections 2.1 and 2.2 we detail that our samples are part of the long-term ecosystem monitoring program (LWF) of the Swiss Federal Institute for Forest, Snow and Landscape research, and that our ancillary data derived from publications related to this program.

The introduction/rational should refer to the needs of information on petrogenic C. 4.3: you could condense your message as you expose the same arguments for bulk C and WEOC.

Thank you, we have included this.

Some references to recent publications on deep C dynamics are lacking (i.e. He et al., 2016; Mathieu et al 2016; Balesdent et al 2018) while they could improve the discussion.

Thank you, some of these papers were already included, and we have added the rest.

I finally encourage the authors to carefully examine the relevance - and the quality - of their illustrations (see some comments below). A better focus of both the text and illustrations would guarantee a better understanding of the message the authors could deliver from the very exceptional dataset they collected.

Please see the comments below, indeed, visuals are key!

Some additional comments
Could you indicate what is "Rsample,t" in Eq 1 and 2.

We have clarified this in the text

The model is based on the assumption that k1 is the turn-over of the WEOC pool. However, how do you justifythat m1 is not the size of the WEOC pool (please provide the C content of the WEOCin your MS). ?

Indeed, this is an assumption, we have adapted this in the text. We have included the WOEC concentration data in the Si, it is usually < 1%.

Clarify what do you mean by deep, and provide numerical value when you refer to depth in the text – currently you sometimes use it indifferently to refer to 30 cm or 80 cm, while the data strongly differ between both depths.

We mean > 20 cm (Mathieu et al., 2016), and have clarified this in the text.

Some Figures and Tables are offered to the readers while they are not utilised in the text: remove them (one example is Fig 3 - PS the information on the back curve is missing in the legend)

Thank you for noticing, we added this.

I do not understand Figure 2. How do you compute turn-over time using one individual time point?

We have clarified this in the text as well as the figure.

I suggest to remove Figure 5 as it is not precise – keep it for oral presentations - (what is vulnerable C?) and to provide Tables with exact numerical data in the main text.

Thank you, we have removed the portion about vulnerable carbon as suggested. As the heatmaps have accurate legends, we do believe it is precise enough to keep it in the paper.

Please provide the C content in for the samples measured for 14C. (Table 3 only show 3 different depths, while the data is available according to Fig 5)

Thank you, we have included the carbon stocks in the main text, which is most relevant when considering the turnover estimates. The carbon content data can be found in the SI as well as the Excel file with the raw data for this paper

You provide twice the particle size distribution (Tab 2 and 3).

Thank you, we have deleted the overlapping part. The difference between the tables is that Table 2 is an average Table 3 is per depth interval.

Some of your interpretations rely on soil waterlogging while this information is not clearly available (when you first mention waterlogged soil line224, the reader has not idea of which sites are concerned). In addition, I would not conclude that waterlogging is a driver of turnover by looking at the non-averaged values in Table S5.

Thank you, we have clarified this and adapted the interpretation.

Why are the radiocarbon signatures of WEOC different between waterlogged and nonwaterlogged soils in 3.1, while calculated turnover rates are not?

Waterlogged soils have slower turnover, both in the bulk and in the WEOC. We have explained in the discussion that this is likely due to the impact of mineralogy as impacted by the geology, interacting with the climate.

Change your title: your gradient is not only a climatic one but a geologic one as well, with strong implication on C cycling.

We have highlighted the geological aspect in the introduction and discussion.

Figure 6: the colour code is not the same than in other figures.

Indeed, this figure shows the depth profiles dug from pits, and not the plot-averaged samples, that's why we opted for a different colour code.

I do not understand Table S1: how do you compute single resolved 14C data?

Thank you, this was not clear, we have clarified this in the text.

Fig S2: what stands at -20cm depth?

It is the 20 cm thick humus layer – we have adapted this and clarified it in the text.

Table S5: figures are not aligned in the table what makes the reading a bit tricky. The caption is not in the same order than the columns. The title of the 5th column is not clear (=> proportion of labile pool would be better)

Thank you for highlighting this, we have adapted Table S5 accordingly.

Dynamics of deep soil carbon – insights from ^{14}C time-series across a climatic gradient

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Abstract. Quantitative constraints on soil organic matter (SOM) dynamics are essential for comprehensive understanding of the terrestrial carbon cycle. Deep soil carbon is of particular interest, as it represents large stocks and its turnover rates remain highly uncertain. In this study, SOM dynamics in both the top and deep soil across a climatic (average temperature ~1-9 °C) gradient are determined using time-series (~20 years) ^{14}C data from bulk soil and water-extractable organic carbon (WEOC). Analytical measurements reveal enrichment of bomb-derived radiocarbon in the deep soil layers on the bulk level during the last two decades. The WEOC pool is strongly enriched in bomb-derived carbon, indicating that it is a dynamic pool. Turnover time estimates of both the bulk and WEOC pool show that the latter cycles up to a magnitude faster than the former. The presence of bomb-derived carbon in the deep soil, as well as the rapidly turning WEOC pool across the climatic gradient implies that there likely is a dynamic component of carbon in the deep soil. Precipitation and bedrock type appear to exert a stronger influence on soil C turnover and stocks as compared to temperature.

1 Introduction

Within the broad societal challenges accompanying climate and land use change, a better understanding of the drivers of turnover of carbon in the largest terrestrial reservoir of organic carbon, as constituted by soil organic matter (SOM), is essential (Batjes, 1996; Davidson and Janssens, 2006; Doetterl et al., 2015; Prietzel et al., 2016). Terrestrial carbon turnover remains one of the largest uncertainties in climate model predictions (Carvalhais et al., 2014; He et al., 2016). At present, there is no consensus on the net effect that climate and land use change will have on SOM stocks (Crowther et al., 2016; Gosheva et al., 2017; Melillo et al., 2002; Schimel et al., 2001; Trumbore and Czimczik, 2008). Deep soil carbon is of particular interest because of its large stocks (Jobbagy and Jackson, 2000; Balesdent et al., 2018; Rumpel and Kogel-Knabner, 2011) and perceived stability. The stability is indicated by low ^{14}C content (Rethemeyer et al., 2005; Schrumpf et al., 2013; van der Voort et al., 2016) and low microbial activity (Fierer et al., 2003). Despite its importance, deep soil carbon has been sparsely studied and remains poorly understood (Angst et al., 2016; Mathieu et al., 2016; Rumpel and Kogel-Knabner, 2011). The inherent complexity of SOM and the multitude of drivers controlling its stability further impedes the understanding of this globally significant carbon pool (Schmidt et al., 2011). In this framework, there is a particular interest in the portion of soil carbon that could be most vulnerable to change, especially in colder climates (Crowther et al., 2016). Water-exactable organic carbon (WEOC) is seen as a dynamic and potentially vulnerable carbon pool in the soil (Hagedorn et al., 2004; Lechleitner et al., 2016). Radiocarbon (^{14}C) can be a powerful tool to determine the dynamics of carbon turnover over decadal to millennial timescales

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because of the incorporation of bomb-derived ^{14}C introduced in the atmosphere in the 1950's as well as the radioactive decay of ^{14}C naturally present in the atmosphere (Torn et al., 2009). Furthermore, ^{14}C can also be employed to identify petrogenic (or geogenic) carbon in the soil profile. Understanding the potential mobilization of stabilized petrogenic carbon is key because it could constitute an additional CO_2 source to the atmosphere (Hemingway et al., 2018). Time-series ^{14}C data is particularly insightful because it enables the tracking of recent decadal carbon. Furthermore, single time-point ^{14}C data can yield two estimates for turnover time, whilst time-series data yields a single turnover estimate (Torn et al., 2009). Given that the so-called "bomb radiocarbon spike" will continue to diminish in the coming decades, time-series measurements are increasingly a matter of urgency in order to take full advantage of this intrinsic tracer (Graven, 2015). Several case-studies have collected time-series ^{14}C soil datasets and demonstrated the value of this approach (Baisden and Parfitt, 2007; Prior et al., 2007; Fröberg et al., 2010; Mills et al., 2013, Schrumpf and Kaiser, 2015). However, these studies are sparse, based on specific single sites and have been rarely linked to abiotic and biotic parameters. Much more is yet to be learned about the carbon cycling through time-series observations in top- and subsoils along environmental gradients. Furthermore, to our knowledge, there are no studies with pool-specific ^{14}C soil time-series focusing on labile carbon.

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This study assesses two-pool soil carbon dynamics as determined by time-series (~20 years) radiocarbon across a climatic gradient. The time-series data is analyzed by a numerically optimized model with a robust error reduction to yield carbon turnover estimates for the bulk and dynamic WEOC pool. Model output is linked to potential drivers such as climate, forest productivity and physico-chemical soil properties. The overall objective of this study is to improve our understanding of shallow and deep soil carbon dynamics in a wide range of ecosystems.

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2 Materials and methods

2.1 Study sites, sampling strategy and WEOC extraction

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The five sites investigated in this study are located in Switzerland between 46-47° N and 6-10° E and encompass large climatic (mean annual temperature (MAT) 1.3-9.2°C, mean annual precipitation (MAP) 864-2126 mm $m^{-2}y^{-1}$) and geological gradients (Table 1). The sites are part of the Long-term Forest Ecosystem Research program (LWF) at the Swiss Federal Institute for Forest, Snow and Landscape Research, WSL (Schaub et al., 2011; Etzold et al., 2014). The soils of these sites were sampled between 1995 and 1998 (Walthert et al., 2002, 2003) and were re-sampled following the same sampling strategy in 2014 with the aim to minimize noise caused by small-scale soil heterogeneity. In both instances sixteen samples were taken on a regular grid on the identical 43 by 43 meters (~1600 m^2) plot (Fig. 1; see Van der Voort et al., 2016 for further details). For the archived samples taken between 1995 and 1998, mineral soil samples down to 40 cm depth (intervals of 0-5, 5-10, 10-20 and 20-40 cm) were taken on an area of 0.5 by 0.5 m (0.25 m^2). For samples >40 cm (intervals of 40-60, 60-80 and 80-100 cm), corers were used to acquire samples (n=5 in every pit, area ~2.8×10^{-3} m^2). The organic layer was sampled by use of a metal frame (30×30 cm). The samples were dried at 35-40°C, sieved to remove coarse material (2 mm), and stored in hard plastic containers under controlled climate conditions in the "Pedothek" at WSL (Walthert et al., 2002). For the samples acquired in 2014 the same sampling strategy was followed, and samples were taken on the exact same plot proximal (~10 m) to the legacy samples. For the sampling, a SHK Martin Burch AG HUMAX soil corer (~2×10^{-3} m^2) was used for all depths (0-100 cm). For the organic layer, a metal frame of 20×20 cm was used to sample. Samples were sieved (2 mm), frozen and freeze-dried using an oil-free vacuum-pump powered freeze dryer (Christ, Alpha 1-4 LO *plus*). For the time-series radiocarbon measurements, all samples covering ~1600 m^2 were pooled to one composite sample per soil depth using the bulk-density. In order to determine bulk-density of the fine earth of the 2014 samples, stones > 2 mm were assumed to have a density of 2.65 g/cm^3. For the Alptal site, sixteen cores were taken on a slightly smaller area (~1500 m^2) which encompasses the control plot of a nitrogen addition experiment (NITREX project) (Schleppi et al., 1998). For this site, no archived samples are available and thus only the 2014 samples were analyzed. Soil carbon stocks were estimated by multiplying SOC concentrations with the mass of soil calculated from measured bulk densities and stone contents for each depth interval (Gosheva et al., 2017). For the Nationalpark site, the soil carbon stocks from 80-100 cm were estimated using data from a separately dug soil profile (Walthert et al., 2003) because the HUMAX corer could not penetrate the rock-dense soil below 80 cm depth. In order to understand very deep soil carbon dynamics (i.e. >100 cm), this study also includes single-time point ^{14}C analyses of soil profiles that were dug down to the bedrock between 1995 and 1998 as part of the LWF programme on the same sites (Walthert et al., 2002). The sampling of the profiles has not yet been repeated.

2.2 Climate and soil data

Temperature and precipitation data are derived from weather stations close to the study sites that have been measuring for over two decades, yielding representative estimates of both variables and over the time period concerned in this study (Etzold et al., 2014). The pH values for all sites and concerned depth intervals were acquired during the initial sampling campaign (Walthert et al. 2002). At Alptal, pH values were determined as described in Xu et al. (2009), values of 10-15 cm were extrapolated to the deeper horizons because of the uniform nature of the Gley horizon. Exchangeable cations were extracted (in triplicate) from the 2-mm-sieved soil in an unbuffered solution of 1 M NH_4Cl for 1 hour on an end-over-end shaker using a soil-to-extract ratio of 1:10. The element concentrations in the extracts were determined by inductively coupled plasma atomic emission spectroscopy (ICP-AES) (Optima 3000, Perkin–Elmer). Contents of exchangeable protons were calculated as the difference between the total and the Al-induced exchangeable acidity as determined (in duplicate) by the KCl method (Thomas, 1982). This method was applied only to soil samples with a pH ($CaCl_2$) < 6.5. In samples with a higher pH, we assumed the quantities of exchangeable protons were negligible. The effective cation-exchange capacity (CEC) was calculated by summing up the charge equivalents of exchangeable Na, K, Mg, Ca, Mn, Al, Fe and H. The base saturation (BS) was defined as the percental fraction of exchangeable Na, K, Mg, and Ca of the CEC (Walthert et al., 2002, 2013). Net primary production (NPP) was determined by Etzold et al. (2014) as the sum of carbon fluxes by woody tree growth, foliage, fruit production and fine root production. Soil texture (sand, silt and clay content) on plot-averaged samples taken in 2014 have been determined using grain size classes for sand, silt and clay respectively of 0.05-2 mm, 0.002-0.05 mm and <0.002 mm according to Klute (1986). The continuous distribution of grain sizes was also determined after removal of organic matter (350 °C for 12 h) using the Mastersizer 2000 (Malvern Instruments Ltd.). Soil water potential (SWP) was measured on the same sites as described in Von Arx et al., (2013). In accordance with Mathieu et al., (2015), topsoil refers to the mineral soil up to 20 cm depth, and deep soil refers to mineral

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soil below 20 cm. Out of the five sites, two are hydromorphic (Gleysol and Podzol in Alptal and Beatenberg respectively), whilst the others are non-hydromorphic (Luvisol, Cambisol and Fluvisol in Othmarsingen, Lausanne and Nationalpark respectively).

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2.3 Isotopic (^{14}C, ^{13}C) and compositional (C, N) analysis

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Prior to the isotopic analyses, inorganic carbon in all samples was removed by vapour acidification for 72 hours (12M HCl) in desiccators at 60 °C (Komada et al., 2008). After fumigation, the acid was neutralised by substituting NaOH pellets for another 48 hours. All glassware used during sample preparation was cleaned and combusted at 450°C for six hours prior to use. Water extractable organic carbon (WEOC) was procured by extracting dried soil with of 0.5 wt% pre-combusted NaCl in ultrapure Milli-Q (MQ) water in a 1:4 soil:water mass ratio (adapted from Hagedorn et al., (2004), details in Lechleitner et al., (2016)).

In order to determine absolute organic carbon and nitrogen content as well as ^{13}C values, an Elemental Analyser-Isotope Ratio Mass Spectrometer system was used (EA-IRMS, Elementar, vario MICRO cube – Isoprime, Vison). Atropine (Säntis) and an in-house standard peptone (Sigma) were used for the calibration of the EA-IRMS for respectively carbon concentration, nitrogen concentration and C:N ratios and ^{13}C. High ^{13}C values were used to flag if all inorganic carbon had been removed by acidification.

For the ^{14}C measurements of the bulk soil samples were first graphitised using an EA-AGE (elemental analyser-automated graphitization equipment, Ionplus AG) system at the Laboratory of Ion Beam Physics at ETH Zürich (Wacker et al., 2009). Graphite samples were measured on a MICADAS (MIniturised radioCArbon DAting System, Ionplus AG) also at the Laboratory of Ion Beam Physics, ETH Zürich (Wacker et al., 2010). For three samples (Alptal depth intervals 40-60, 60-80 and 80-100 cm) the ^{14}C signature was directly measured as CO_2 gas using the recently developed online elemental analyzer (EA) - stable isotope ratio mass spectrometers (IRMS)–AMS system et ETH Zürich (McIntyre et al., 2016). Oxalic acid (NIST SRM 4990C) was used as the normalising standard. Phthalic anhydride and in-house anthracite coal were used as blank. Two in-house soil standards (Alptal soil 0-5 cm, Othmarsingen soil 0-5 cm) were used as secondary standards. For the WEOC, samples were converted to CO_2 by Wet Chemical Oxidation (WCO) (Lang et al., 2016) and run on the AMS using a Gas Ion Source (GIS) interface (Ionplus). To correct for contamination, a range of modern standards (sucrose, Sigma, δ13C = -12.4 ‰ VPDB, $F^{14}C$ = 1.053 ± 0.003) and fossil standards (phthalic acid, Sigma, δ13C = -33.6‰ VPDB, $F^{14}C$ <0.0025) were used (Lechleitner et al., 2016).

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2.4 Numerical optimization to find carbon turnover and size of the dynamic pool

2.4.1 Turnover based on a single ^{14}C measurement

The ^{14}C signature of a sample can be used to estimate turnover time of a carbon pool (Torn et al., 2009).

$$R_{sample,t} = k \times R_{atm,t} + (1 - k - \lambda) \times R_{sample(t-1)} \quad (1)$$

$$R_{sample,t} = \frac{\Delta^{14}C_{sample}}{1000} + 1 \quad (2)$$

In Eq. 1-2, the constant for radioactive decay of ^{14}C is indicated as λ, the decomposition rate k (inverse of turnover time) is the only unknown in this equation and is hence the variable for which the optimal value that fits the data is sought using the model. The R value of the sample is inferred from $\Delta^{14}C$, hence accounting for the sampling year, as shown in Eq. (2) (Herold et al., 2014; Solly et al., 2013). In order to avoid ambiguity, the term *turnover time* and not i.e. mean residence time is used solely in this manuscript (Sierra et al., 2016).

For the turnover time estimation, we assumed the system to be in steady state over the modeled period (~1×10[4] years, indicating soil formation since the last glacial retreat (Ivy-Ochs et al., 2009)), hence accounting both for radioactive decay and incorporation of the bomb-testing derived material produced in the 1950's and 1960's (Eq. 1.) (Herold et al., 2014; Torn et al., 2009). We assumed an initial fraction modern (F_m) of ^{14}C value of 1 at 10000 B.C.. For the period after 1900 atmospheric fraction modern (F_m) values of the Northern Hemisphere were used (Hua et al., 2013). This equation could be solved in Excel with manual iterations (e.g. Herold et al., 2014), or alternatively a numerical optimization can be used to find the best fit automatically. In this paper, we used a numerical optimization constructed in MATLAB version 2015a (The MathWorks, Inc., Natick, Massachusetts, United States) to find the best fit. The numerical optimization is exhaustive, meaning that every single turnover value from 1 to 10.000 years with an interval of 0.1 year is tested. The error is defined as the difference between the fitted value of R and the measured value (Eq. 3). The turnover value with the lowest error is then automatically selected.

$$Error_{single\ timepoint} = |R_{calculated} - R_{measured}| \quad (3)$$

The residual error of each fit are provided in the Supplemental Information (SI) Table 3. Turnover times determined with the numerical optimization match the manually optimized turnover modeling published previously (Herold et al., 2014; Solly et al., 2013).

2.4.2 Turnover based on two ^{14}C measurements

A single ^{14}C value could yield possible turnover values (Torn et al., 2009, Graven et al., 2015). If there is a time-series ^{14}C dataset, this problem can be eliminated. In this paper, we have time-series data of both the bulk soil, as well as the vulnerable fraction (WEOC). For all samples a time-series dataset is available, both data points are employed to give the best estimate of turnover time. The same numerical optimization (Eq. 1 and 2) as we did for a single time-point, except that we try to find the best fit for both time points whilst reducing the compounding residual mean square error (RSME, Eq. 4). As can be seen in Fig. 2a, single time points can yield

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two likely turnover times but when two datapoints are available, a single value can be found. The input data for Figure 2 can be found in SI Table 1. The results of the time-series turnover modelling for both the bulk and WEOC pool of the sub-alpine site Beatenberg are shown in Fig. 3.

$$Error_{two\ timepoints} = \sqrt[2]{|R_{calculated} - R_{measured}|^2_{time\ point\ 1} + |R_{calculated} - R_{measured}|^2_{time\ point\ 1}} \quad (4)$$

2.4.3 Vegetation-induced lag

In order to account for vegetation-lag, two scenarios were run: firstly (1) with no assumed lag between the fixation of carbon from the atmosphere and input into to the soil and (2) model run with a lag of fixation of the atmospheric carbon as inferred from the dominant vegetation (Von Arx et al., 2013; Etzold et al., 2014). In the case of full deciduous trees coverage a lag of two years was assumed, and for the case of 100% conifer-dominated coverage a lag of 8 years was incorporated (Table 1).

2.4.4 Turnover and size vulnerable pool based on two-pool model

As SOM is complex and composed of a continuum of pools with various ages (Schrumpf and Kaiser, 2015) and there is data available from two SOM pools, the ^{14}C time-series data can be leveraged to create a two-pool model. The following assumptions were made: First, both pools (slow & fast) make up the total carbon pool (Eq. 5). Secondly, the total turnover of the bulk soil is made up out of the "dynamic" fraction turnover multiplied by "dynamic" fraction pool size and the "slow" pool turnover multiplied by "slow" pool size (Eq. 6). Furthermore, we assume that the signature of the sample (the time-series bulk data) is determined by the rate of incorporation of the material (atmospheric signal) and the loss of carbon the two pools (Eq. 7). Lastly, we assume that the radiocarbon signal of the WEOC pool is representative for a dynamic pool, as it could be representative for a larger component of rapidly turning over carbon, even in the deep soil (Baisden and Parfitt, 2007; Koarashi et al., 2012). The turnover rate of the slow pool was set between 100 and 10.000 years, with a time-step of 10 years. The size of the dynamic pool was set to be between 0 and 0.5, with a size-step of 0.01.

$$1 = F_1 + F_2 \quad (5)$$

$$k_{total} = (F_1/k_1 + F_2/k_2)^{-1} \quad (6)$$

$$R_{sample,t} = k_{total} \times R_{atm,t} + F_1[(1 - k_1 - \lambda) \times R_{sample(t-1)}] + F_2(1 - k_2 - \lambda) \times R_{sample(t-1)} \quad (7)$$

Where F_1 is the relative size of the dynamic pool, and F_2 is the relative size of the (more) stable pool. The k_1 is the inverse of the turnover time of the dynamic or WEOC as determined using the numerical optimisation of Eq. 1-4. The k_2 is the inverse of the turnover time of the slow pool. The calculation of the error term becomes for complex because it needs to be recalculated for each unique combination of pool-size distribution (Eq. 5) and turnover time (inverse of k, Eq. 6). Therefore, the error space changes from column vector to a two-dimensional

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matrix of length of the step size increments (F_1) and width of the inverse of the turnover time of the slow pool (k_2).

$$Error_{k_2,F_1} = \sqrt[2]{|R_{calculated} - R_{measured}|^2_{time\ point\ 1} + |R_{calculated} - R_{measured}|^2_{time\ point\ 2}} \quad (8)$$

$$Error = Min(Error_{k_2,F_1}) \quad (9)$$

The numerical optimization finds the likeliest solution for the given dataset. This model constitutes a best fit, and more data would better constrain the results. Additional details can be found in the Supplementary Information (SI) text and SI Fig. 1. **All Matlab-based numerical optimization codes can be found in the SI**. For correlations (packages HMISC, corrgram, method = pearson), statistical software R version 1.0.153 was used.

3 Results

3.1 Changes of radiocarbon signatures over time

Overall, there is a pronounced decrease in radiocarbon signature with soil depth at all sites (Fig. 4). The time-series results show clear changes in radiocarbon signature over time from the initial sampling period (1995-1998) as compared to 2014, with the magnitude of change depending on site and soil depth. In the uppermost 5 cm of soils, the overarching trend in the bulk soil is a decrease in the ^{14}C bomb-spike signature in the warmer climates (Othmarsingen, Lausanne), whilst at higher elevation (colder) sites (Beatenberg, Nationalpark) the bomb-derived carbon appears to enter the top soil between 1995-8 and 2014.

Water-extractable OC (WEOC) has an atmospheric ^{14}C signature in the top soil at all sites in 2014. The deep soil in the 1990's still has a negative $\Delta^{14}C$ signature of WEOC at multiple sites. There are two distinguishable types of depth trends for WEOC in the 2014 dataset: (1) WEOC has the same approximate ^{14}C signature throughout depth (Othmarsingen, Beatenberg), (2) WEOC becomes increasingly ^{14}C depleted with depth (Alptal, Nationalpark), or an intermediate form where WEO^{14}C is modern throughout the top soil but becomes more depleted of ^{14}C in the deep soil (Lausanne) (Fig. 4). The isotopic trends of WEOC co-vary with grain size as inherited from the bedrock type (Walthert et al., 2003). Soils with a relatively modern WEO^{14}C signature in 2014 (down to 40 cm) are underlain by bedrock with large grained (SI Fig. 2, Table SI 3) components (the moraines and sandstone at Othmarsingen, Lausanne and Beatenberg respectively). Soils where WEO^{14}C signature decreases with depth are underlain by bedrock containing fine-grained components. For instance, the Flysch in Alptal (Schleppi et al., 1998) and intercalating layers of silt and coarse grained alluvial fan in Nationalpark (Walthert et al., 2003) respectively.

3.2 Carbon turnover patterns

Incorporation of a vegetation-induced time lag (Table 2, SI Table 2) has an effect on modelled carbon dynamics in the organic layer, but this effect is strongly attenuated in the 0-5 cm layer in the mineral soil and virtually absent for the deeper soil layers. The residual errors associated to the carbon turnover estimates converge to a single point (Figure 2) and are low (i.e. < 0.06 R, SI Tables 3 and 4). Turnover times show two modes of behavior for well-drained soils and hydromorphic soils, respectively. The non-hydromorphic soils

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have relatively similar values with decadal turnover times for the 0-5 cm layer, increasing to an order of centuries down to 20 cm depth, and to millenia in deeper soil layers (~980 to ~3940 years at 0.6 to 1 m depth) (Fig. 5). In contrast, the hydromorphic soils are marked by turnover times that are up to an order of magnitude larger, from centennial in top soil to (multi-) millennial in deeper soils. At the Beatenberg podsol, turnover time of the deepest layer (40-60 cm, ~1900 y) is faster than the shallow layer (20-40 cm, ~1300 y) (Figure 5, SI Table 5).

Carbon stocks also show distinct difference between drained and hydromorphic soils with greater stock in the hydromorphic soils (~15 kg C m^{-2} at Beatenberg and Alptal vs. ~ 6 - ~7 kg C m^{-2} at Othmarsingen, Lausanne and Nationalpark, Fig. 5, Table 3)).

The turnover times of the WEOC mimic the trends in the bulk soil but are up to an order of magnitude faster. Considering WEOC turnover in the non-hydromorphic soils only, there is a slight increase in WEOC turnover with decreasing site temperature, but the trend is not significant (SI Table 4). The modeled estimate for dynamic fraction is variable at the surface but decreases towards the lower top soil (from ~0.2 at 0-5 cm to ~0.01 at 10-20 cm in Othmarsingen). In the deep soil, the model indicates there could also be a non-negligible proportion of dynamic carbon (e.g. 0.10-0.23 at 20-40 cm). The residual errors associated to the error reduction of the two-pool model are also low (i.e. < 0.06 R). but do not converge as strongly as the single-pool model (SI Figure 1).

3.3 Pre-glacial carbon in deep soil profiles

The turnover times of deep soil carbon exceed 10,000 years in several profiles, indicating the presence of carbon that pre-dates the glacial retreat (Fig. 6). These profiles are located on carbon-containing bedrock and concern the deeper soil (80-100 cm) of the Gleysol (Alptal), as well as >100 cm in the Cambisol (Lausanne) (Fig. 6, SI Table 6).

3.4 Environmental drivers of carbon dynamics

Pearson correlation was used to assess potential relationships between carbon stocks and turnover and their potential controlling factors (climate, NPP, soil texture, soil moisture and physicochemical properties (Table 4, SI Table 7, 8)). For the averaged top soil (0-20 cm, n=5), carbon stocks were significantly positive correlated to Mean Annual Precipitation (MAP). Turnover time in the bulk top soil negatively correlated with silt content and positively with average grain size. Turnover time in the WEOC of the top soil did not correlate significantly with any parameter except a weak positive correlation with grain size. Deeper soil bulk stock and turnover time positively correlated with MAP and iron content.

4 Discussion

4.1 Dynamic deep soil carbon

4.1.1 Rapid shifts in ^{14}C abundance reflect dynamic deep carbon

The propagation of bomb-derived carbon into supposedly stable deep soil on the bulk level across the climatic gradient implies that SOM in deep soil contains a dynamic pool and could be less stable and potentially more vulnerable to change than previously thought. This possibility is further supported by the WEO^{14}C which is

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consistently more enriched in bomb-derived carbon than the bulk soil. Near-atmospheric signature WEO^{14}C pervades up to 40 or even 60 cm depth. Hagedorn et al., (2004) also found WEOC to be a highly dynamic pool using ^{13}C tracer experiments in forest soils.

We consider our ^{14}C comparison over time to be robust because the grid-based sampling and averaging was repeated on the same plots which excludes the effect plot-scale variability (Van der Voort et al., 2016). Our ^{14}C time-series data in the deep soil corroborate pronounced changes in ^{14}C (hence substantial SOM turnover) in subsoils of an area with pine afforestation (Richter and Markewitz, 2001). The findings are also in agreement with results from an incubation study by Fontaine et al., (2007) which showed that the deep soil can have a significant dynamic component. Baisden et al., (2007) also found indications of a deep dynamic pool using modeling on ^{14}C time-series on the bulk level on a New Zealand soil under stable pastoral management.

4.1.2 Carbon dynamics reflect soil-specific characteristics at depth

Bulk carbon turnover for the top and deeper soil fall in the range of prior observations and models, although the data for the latter category is sparse (Scharpenseel and Becker-Heidelmann, 1989; Paul et al., 1997; Schmidt et al., 2011; Mills et al., 2013; Braakhekke et al., 2014). The carbon turnover is related to soil-specific characteristics. The slower turnover of hydromorphic as compared to non-hydromorphic soils is likely due to increased waterlogging and limited aerobicity (Hagedorn et al., 2001) which is conducive to slow turnover and enhanced carbon accumulation. The WEOC turns over up to an order of magnitude faster than the bulk and mirrors these trends, indicating that it indeed is a more dynamic pool (Hagedorn et al., 2004; Lechleitner et al., 2016). Results also reflect known horizon-specific dynamics for certain soil types, particularly in the deep soil. The hydromorphic Podsol at Beatenberg shows specific pedogenetic features such as an illuviation layer with an enrichment in humus and iron in the deeper soil (Walthert et al., 2003) where turnover of bulk and WEOC is faster and stocks are higher than in the elluvial layer above (Fig. 5). This is likely due to the input of younger carbon via leaching of dissolved organic carbon. The non-hydromorphic Luvisols are marked by an enrichment of clay in the deeper soil, which can enhance carbon stabilization (Lutzow et al., 2006). This also reflected in the turnover time of the 60-80 cm layer in the Othmarsingen Luvisol – in this clay-enriched depth interval (Walthert et al., 2003), turnover is relatively slow as compared to the other (colder) non-hydromorphic soils (Fig. 5). These patterns are consistent with findings by Mathieu et al., (2015) that the important role of soil pedology on deep soil carbon dynamics.

4.1.3 Dynamic carbon at depth & implications for carbon transport

The analytical ^{14}C data as well as turnover time estimates indicate that there is likely a dynamic portion of carbon in the deep soil. The estimated size of the dynamic pool can be large, even at greater depth than it was observed by other ^{14}C time-series (Richter and Markewitz, 2001; Baisden and Parfitt, 2007; Koarashi et al., 2012). The two-pool modelling indicates that the size of dynamic pool in the deep soil can be upwards of ~10%. A deep dynamic pool is consistent with findings of a ^{13}C tracer experiment by Hagedorn et al., (2001) that shows with that relatively young (<4 years) carbon can be rapidly incorporated in the top soil (20% new C at 0-20 cm depth) but also in the deep soil (50 cm), and findings by Balesdent et al., (2018) which estimate that up to 21% of the carbon between 30-100 cm is younger than 50 years. Rumpel and Kögel-Knabner (2011) have highlighted the importance of the poorly understood deep soil carbon stocks and a significant dynamic pool in

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the deep soil could imply that carbon is more vulnerable than initially suspected. One major input pathway of younger C into deeper soils is the leaching of DOC (Kaiser and Kalbitz, 2012; Sanderman and Amundson, 2009). Here, we have measured WEOC – likely primarily composed of microbial metabolites (Hagedorn et al., 2004) – carrying a younger ^{14}C signature than bulk SOM and thus, representing a translocator of fresh carbon to the deep soil. The WEOC turnover time is in the order of decades, implying that it is not directly derived from decaying vegetation, but rather composed of microbial material feeding on the labile portion of the bulk soil. In addition to WEOC, roots and associated mycorrhizal communities may also provide a substantial input of new C into soils in deeper soils (Rasse et al., 2005). Additional modelling such as in CENTURY and RotC could provide additional insights into the soil carbon dynamics and fluxes (Manzoni et al., 2009)

4.2 Contribution of petrogenic carbon

Our results on deep soil carbon suggest the presence of pre-aged or ^{14}C-dead (fossil), pre-interglacial carbon in the Alptal (Gleysol) and Lausanne (Cambisol) profiles, implying that a component of soil carbon is not necessarily linked to recent (< millenial) terrestrial productivity and instead constitutes part of the long-term (geological) carbon cycle (> millions of years). In the case of the Gleysol in Alptal, the ^{14}C-depleted material could be derived from the poorly consolidated sedimentary rocks (Flysch) in the region (Hagedorn et al., 2001a; Schleppi et al., 1998; Smith et al., 2013), whereas carbon present in glacial deposits and molasse may contribute in deeper soils at the Lausanne (Cambisol) site. The potential contribution of fossil carbon was estimated using a mixing model using the signature of a soil without fossil carbon, the signature of fossil carbon and the measured values (SI Table 4). Fossil carbon contribution in the Alptal profile between 80-100 cm (Fig. 6, SI Table 4) is estimated at ~40 %. Below one meter at Lausanne site the petrogenic percentage ranges from ~20% at 145 cm up to ~80 % at 310 cm depth (Fig. 6, SI Table 4).

Other studies analyzing soils have observed the significant presence of petrogenic (geogenic in soil science terminology) in loess-based soils (Helfrich et al., 2007; Paul et al., 2001). Our results suggest that pre-glacial carbon may comprise a dominant component of deep soil organic matter in several cases, resulting in an apparent increase in the average age (and decrease in turnover) of carbon in these soils. Hemingway et al., (2018) have highlighted that fossil carbon oxidized in soils can lead to significant additional CO_2 emissions. Therefore, the potential of soils to 'activate' fossil petrogenic carbon should be considered when evaluating the soil carbon sequestration potential.

4.3 Controls on carbon dynamics and cycling

In order to examine the effects of potential drivers on soil C turnover and stocks we explore correlations between a number of available factors which have previously been proposed, such as texture, geology, precipitation, temperature and soil moisture (Doetterl et al., 2015; McFarlane et al., 2013; Nussbaum et al., 2014; Seneviratne et al., 2010; van der Voort et al., 2016).

From examination of data for all samples it emerges that C turnover does not exhibit a consistent correlation with any specific climatological or physico-chemical factor. This implies that no single mechanism predominates and/or that there is a combined impact of geology and precipitation as these soil-forming factors affect grain size distribution, water regime and mass transport in soils. Exploring potential relationships in greater detail, we see that carbon stocks in the top soil and deep soil as well as turnover time is positively related

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to MAP, which could be linked to waterlogging and anaerobic conditions even in upland soils leading to a lower decomposition and thus to a higher build-up of organic material (Keiluweit et al., 2015). Our results are supported by the findings based on >1000 forest sites that precipitation exerts a strong effect on soil C stocks across Switzerland (Gosheva et al., 2017; Nussbaum et al., 2014). Furthermore, Balesdent et al., (2008) also highlighted the role of precipitation and evapotranspiration on deep soil organic carbon stabilisation. Nonetheless, it has to be noted that for these sites, the precipitation range does not include very dry soils (MAP 864-2126 mm/y). Turnover in both top and deep soil was most closely correlated with texture. The positive correlation of top soil turnover with grain size and negative correlation with the amount of silt-sized particles reflects lower stabilization in larger-grained soils as opposed to clay-rich soils with a higher and more reactive surface area (Rumpel and Kogel-Knabner, 2011). Mathieu et al., (2015) also stressed the decisive role of soil pedology on deep soil carbon storage. Overall, geology seems to impact the carbon cycling in three key ways. Firstly, when petrogenic carbon is present in the bedrock from shale or reworked shale (Schleppi et al., 1998; Walthert et al., 2003), fossil carbon contributes to soil carbon. Secondly, porosity of underlying bedrock either prevents or induces waterlogging which in turn affects turnover. Thirdly, the initial components of the bedrock (i.e. silt-sizes layers in an alluvial fan) influence the final grain size distribution and mineralogy (SI Fig. 2, Table 3), which is also reflected in the bulk and pool-specific turnover. Within the limited geographic and temporal scope of this paper, we hypothesize that for soil carbon stocks and their turnover, temperature is not the dominant driver, which has been concluded by some (Giardina and Ryan, 2000) but refuted by others (Davidson et al., 2000; Feng et al., 2008). The only climate-related driver which appears to be significant for the deep soil is precipitation.

4.4 Modular robust numerical optimization

The numerical approach used here builds on previous work concerning turnover modeling of bomb-radiocarbon dominated samples (Herold et al., 2014; Solly et al., 2013; Torn et al., 2009) and the approach used in numerous time-series analysis with box modeling using Excel (Schrumpf and Kaiser, 2015) or Excel solver (Baisden et al., 2013; Prior et al., 2007). However certain modifications were made in order to (i) provide objective repeatable estimates, (ii) incorporate longer time-series data, and (iii) identify samples impacted by petrogenic (also called geogenic) carbon. Identifying petrogenic carbon in the deep soil is important considering the large carbon stocks in deep soils (Rumpel and Kogel-Knabner, 2011) and the wider relevance of petrogenically-derived carbon in the global carbon cycle (Galy et al., 2008). This approach is modular and could be adapted in the future to identify the correct turnover for time-series ^{14}C data, which is becoming increasingly important with the falling bomb-peak (Graven, 2015). For the single and time-series data, the results from the numerical solution were benchmarked to the Excel-based model, and it was found that the results agree.

Other studies (e.g. Baisden and Canessa, 2012; Prior et al., 2007) also use time-series data to estimate the value for two unknowns simultaneously (size of the pool size and turnover time). The error does not always converge to single low point, but can have multiple minima (SI Fig. 1). This potential issue should be considered when interpreting the data. More time-series data is required to eliminate this problem.

5 Conclusion

Deleted: The modeled size of the dynamic pool is mostly related to precipitation and texture. It correlates positively with MAP and clay content and negatively with sand content. This correlation could be because sandier soils offer less reactive surfaces for SOM stabilization as opposed to clay-rich soils (Lutzow et al., 2006). Additionally, wetter conditions inhibit SOM breakdown.

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Time-series radiocarbon (^{14}C) analyses of soil carbon across a climatic range reveals recent bomb-derived radiocarbon in both upper and deeper bulk soil, implying the presence of a rapidly turning over pool at depth. Pool-specific time-series measurements of the WEOC indicate this is a more dynamic pool which is consistently more enriched in radiocarbon than the bulk. Furthermore, the estimated modeled size of the dynamic fraction is non-negligible even in the deep soil (~0.1-0.2). This could imply that a component of the deep soil carbon could be more dynamic than previously thought.

The interaction between precipitation and geology appears to be the main control on carbon dynamics rather than site temperature. Carbon turnover in non-hydromorphic soils is relatively similar (decades to centuries) despite dissimilar climatological conditions. Hydromorphic soils have turnover times which are up to an order of magnitude slower. These trends are mirrored in the dynamic WEOC pool, suggesting that in sandy, non-waterlogged (aerobic) soils the transport of relatively modern (bomb-derived) carbon into the deep soil and/or the microbial processing is enhanced as compared to fine-grained waterlogged (anaerobic) soils.

Model results indicate certain soils contain significant quantities of pre-glacial or petrogenic (bedrock-derived) carbon in the deeper part of their profiles. This implies that soils not only sequester “modern” but can rather also mobilize and potentially metabolize “fossil” or geogenic carbon.

Overall, these time-series ^{14}C bulk and pool-specific data provide novel constraints on soil carbon dynamics in surface and deeper soils for a range of ecosystems.

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Author contributions

T.S. van der Voort planned, coordinated and executed the sampling strategy and sample collection, performed the analyses, conceptualized and optimized the model and processed resulting data. U. Mannu led the model development. F. Hagedorn lent his expertise on soil carbon cycling and soil properties. C. McIntyre facilitated and coordinated the radiocarbon measurements and associated data corrections. L. Walthert and P. Schleppi lent their expertise on the legacy sampling and provided data for the compositional analysis. N. Haghipour performed in isotopic and compositional measurements. T. Eglinton provided the conceptual framework and aided in the paper structure set-up. T.S. van der Voort prepared the manuscript with help of all co-authors.

Tables

Table 1 Overview sampling locations and climatic and ecological parameters.

Location	Soil type	Geology	Latitude(N)/ Longitude (E)	Soil depth (m)	Depth Upper limit waterlogging (m)[1]	Altitude Elevation (m a.s.l.)	MAT °C	MAP mm y^{-1}	NPP g C $m^{-2}y^{-1}$
Othmarsingen[1, 2, 3]	Luvisol	Calcareous moraine	47°24'/8°14'	>1.9	2.5	467-500	9.2	1024	845
Lausanne[1, 2, 3]	Cambisol	Calcarous and shaly moraine	46°34'/6°39'	>3.2	2.5	800-814	7.6	1134	824
Alptal[1, 2, 3, 4]	Gleysol	Flysch (carbon-holding sedimentary rock)	47°02'/8°43'	>1.0	0.1	1200	5.3	2126	347
Beatenberg[1, 2, 3]	Podzol	Sandstone	46°42'/7°46'	0.65	0.5	1178-1191	4.7	1163	302
Nationalpark[1, 2, 3]	Fluvisol	Calcareous alluvial fan	46°40'/10°14'	>1.1	2.5	1890-1907	1.3	864	111

[1] Walthert et al. (2003) [2]Etzold et al., (2014) [3]Von Arx et al., (2013) [4]Krause et al., (2013) for Alptal data

Table 2 Vegetation and soil data of the study sites. Soil water potential (hPa) are for 15 cm depth.

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Location[1]	Deciduous tree species (%)[3]	Dominant tree species[3]	Inferred lag carbon fixation (y)	Organic layer Type[1]	Soil water potential (hPa) percentiles[3]		
					5%	50%	95%
Othmarsingen	100	*Fagus sylvatica*	2	Mull	-577	-39	-9
Lausanne	80	*Fagus sylvatica*	3	Mull	-547	-49	-8
Alptal[4]	15	*Picea abies*	7	Mor to anmoor	-38	-13	+1
Beatenberg	0	*Picea abies*	8	Mor	-50	-14	+1
Nationalpark	0	*Pinus montana*	8	Moder	-388	-65	-13

[1] Walthert et al. (2003) [2]Etzold et al., (2014), [3]Von Arx et al. (2013),[4]Krause et al., (2013)

Table 3 Soil properties as well as carbon stocks and fluxes in 0-20, 20-60, and 60-100 cm depth of the study sites for the bulk and water-extractable organic carbon (WEOC).

Location	Depth interval (m)	pH[1]	CEC[1](mmolc/kg)	$Fe_{exchangeable}$ (mmolc/kg)	$Al_{exchangeable}$ (mmolc/kg)	Sand content (%)	Silt content (%)	Clay content (%)	Carbon stock kgC/m^2	Average turnover bulk (y)	Average turnover WEOC (y)
Othmarsingen[1]	0.0-0.2	4.4	62.2	0.15	42	46.8	35.5	17.6	4.84	173	35
	0.2-0.6	4.4	62.8	0.10	49	44.3	33.3	22.4	1.69	868	518
	0.6-0.8	4.9	99.5	0.06	41	46.7	28.4	25.0	0.28	3938	-
Lausanne[1]	0.0-0.2	4.5	60.8	0.13	43	49.2	32.6	18.2	3.24	353	77
	0.2-0.6	4.6	43.9	0	34	50.2	32.0	17.8	2.12	1239	588
	0.6-1.0	4.8	49.7	0	35	50.5	31.5	18.1	0.69	2246	1502[5]
Alptal[2,3,4]	0.0-0.2	4.5	417	-	19	19.3	39.4	41.3	7.73	437	166
	0.2-0.6	4.7	340	-	14	4.90	47.0	48.1	7.24	3314	893[6]
	0.6-1.0	4.7	340	-	-	-	-	-	6.54	5165	-
Beatenberg[1]	Organic layer	3.1	260.2	2.8	33	-	-	-	7.05	53	-
	0.0-0.2	4.0	35.6	1.7	18	84.9	12.4	2.7	3.65	1224	293
	0.2-0.6	4.1	23.1	0.40	17	83.2	12.3	4.6	4.10	1607	677
Nationalpark[1]	0.0-0.2	8.3	171.8	0.1	0.0	47.5	34.8	17.7	3.23	180	92
	0.2-0.6	8.8	106.3	0.0	0.0	61.9	32.5	5.7	0.36	612	214
	0.6-0.8	-	-	0.0	0.0	60.6	33.6	5.9	0.08	983	-

[1]Walthert et al., 2002, Walthert et al., 2003., Fe and Al content (mmolc/kg) determined by NH_4Cl extraction.
For the 0.2-0.6 depth interval the CEC determined for 0.2-0.4 m was taken, and similarly for the depth interval 0.6-1.0 m the values for 0.6-0.8 m were taken in the case of Othmarsingen, Lausanne Beatenberg and Nationalpark.
[2]Krause et al., 2013
[3]Diserens et al,1992, CEC determined (mmeq/kg), hydrogen lead and zinc ions were not included, Aluminium content determined by Lakanen method. CEC values for 0.2-0.4 m were extrapolated to 1 m. [4]Xu et al., 2009 [5]Depth to 0.8 m [6]Depth to 0.4 m

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Table 4 Pearson correlations for averaged depth intervals for the top soil (0-20 cm, n=5) and deep soil (20-60 cm, n=5). Significance denoted with ˙, *, ** or *** for respectively p-values smaller than 0.1 (marginally significant) 0.05, 0.005 and 0.0005 (significant). Non-significant correlations are indicated by the superscript **ns**. SWP or soil water potential used are the median values at 15 cm for each of these 5 sites (Von Arx et al., 2013). Water-extractable carbon is abbreviated to WEOC. Results indicate that no single climatic or textural factor consistently co-varies with carbon stocks, or turnover time.

Explaining variable	**Stock$_{0-20\ cm}$**	**Turnover time bulk$_{0-20\ cm}$**	**Turnover time WEOC $_{0-20\ cm}$**	**Stock$_{20-60\ cm}$**	**Turnover time$_{20-60\ cm}$**
MAT	0.17^{ns}	-0.14^{ns}	-0.36^{ns}	0.02^{ns}	0.02 ^{ns}
MAP	**0.96****	0.11^{ns}	0.30^{ns}	**0.93****	**0.98******
NPP	0.2^{ns}	0.65^{ns}	0.38^{ns}	0.03^{ns}	-0.10 ^{ns}
Sand	-0.66^{ns}	0.72^{ns}	0.53^{ns}	-0.56^{ns}	-0.70 ^{ns}
Silt	0.38^{ns}	**-0.91***	-0.78^{ns}	0.29^{ns}	-0.47 ^{ns}
Clay	**0.81˙**	-0.51^{ns}	-0.29 ^{ns}	0.71^{ns}	0.80 ^{ns}
CEC	-0.67^{ns}	-0.24^{ns}	0.05^{ns}	0.74^{ns}	**0.82˙**
pH	-0.74^{ns}	-0.47^{ns}	-0.3^{ns}	-0.51 ^{ns}	-0.46^{ns}
Fe	0.24^{ns}	0.98*	0.97*	0.98*	-0.78^{ns}
Al	0.18^{ns}	-0.16^{ns}	-0.41^{ns}	-0.17 ^{ns}	-0.17 ^{ns}
SWP	0.70^{ns}	0.68^{ns}	0.71^{ns}	-	-
Average Grain size	-0.25^{ns}	**0.97***	**0.88˙**	0.05 ^{ns}	-0.16 ^{ns}

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Figures

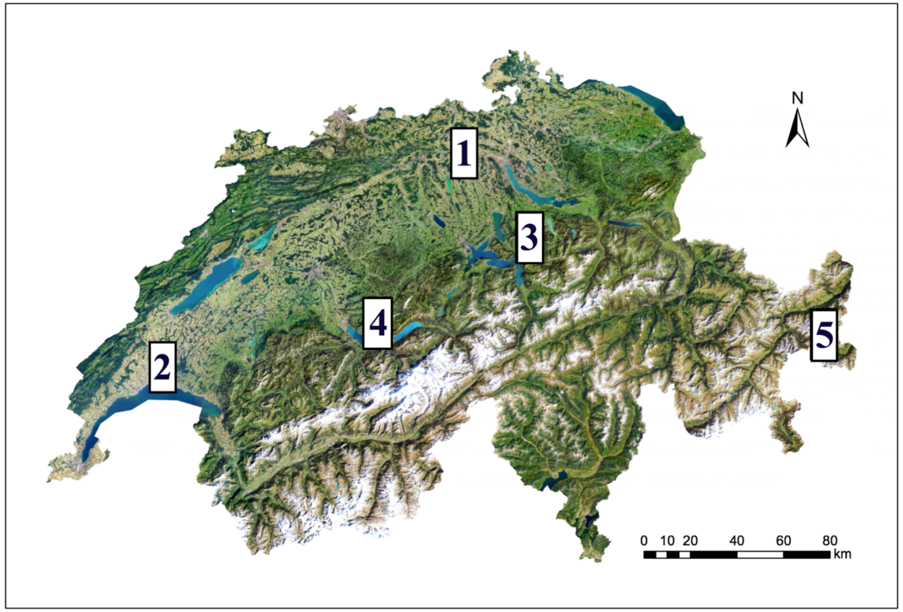

Figure 1 Sample locations, all of which are part of the Long-term ecosystem research program (LWF) of the Swiss Federal Institute WSL, 1) Othmarsingen, 2) Lausanne, 3) Alptal, 4) Beatenberg and 5) Nationalpark Image made using 2016 swisstopo (JD100042).

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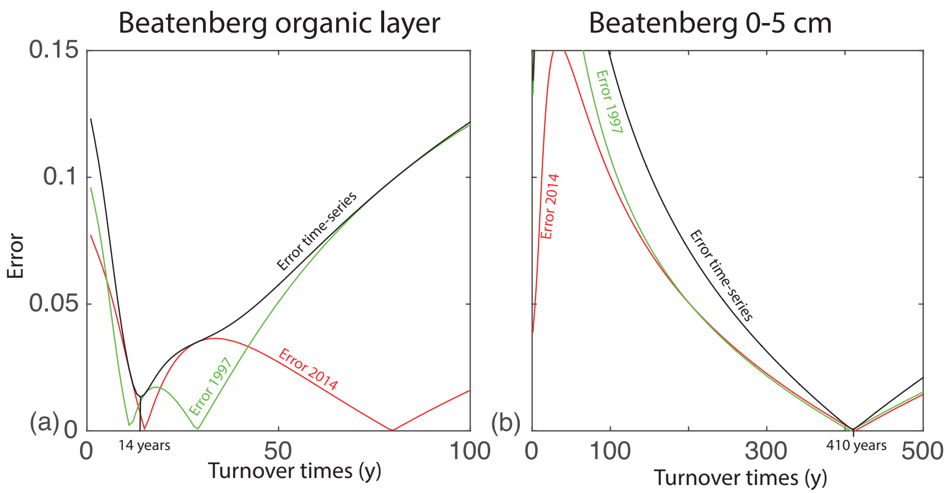

Figure 2 Numerical optimization of least mean-square error reduction, showing and the reduction of error spread for two soil depths. For the Beatenberg organic layer (a) the individual ^{14}C time-points for both 1997 and 2014 both yield two solutions are almost equally likely (i.e. the error nears zero). The combined optimization using both the time-points reveal the likeliest option. For the (b) 0-5 cm layer the single time points only have a single likely solution.

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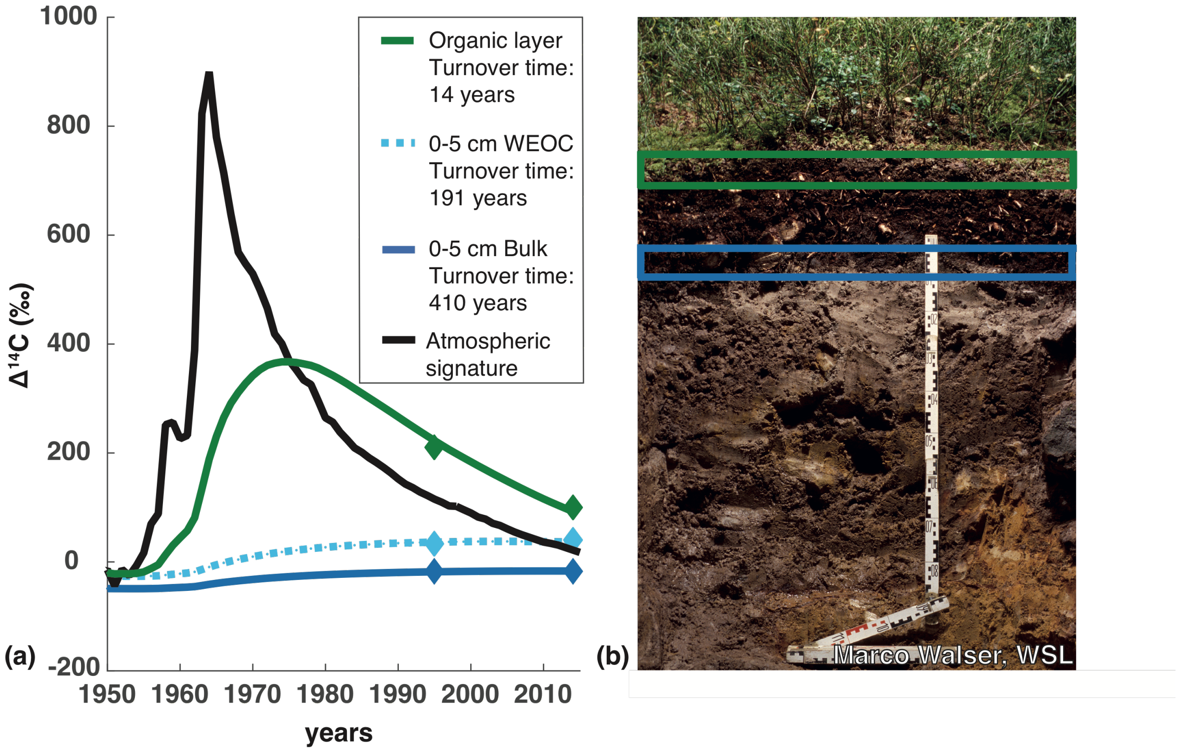
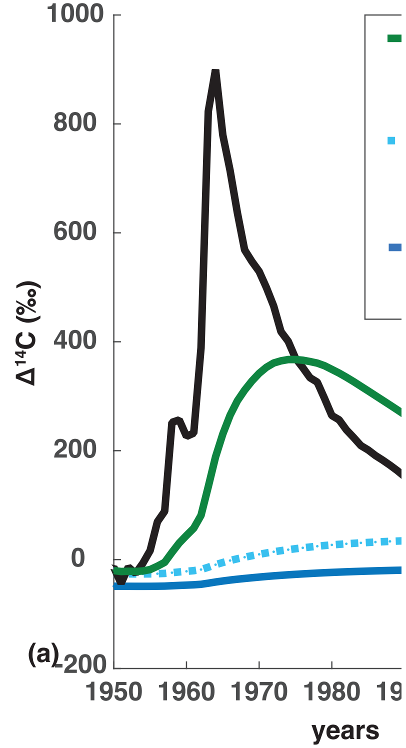


Figure 3 (a) Time-series soil carbon turnover time in years (y) as determined by numerical modelling for (b) sub-alpine site Beatenberg. The bulk turnover in the organic layer is rapid (14 years), followed by the turnover of the water-extractable organic carbon (WEOC) (191 years) and the bulk turnover of the soil (410 years). Photo soil profile courtesy of Marco Walser, WSL.



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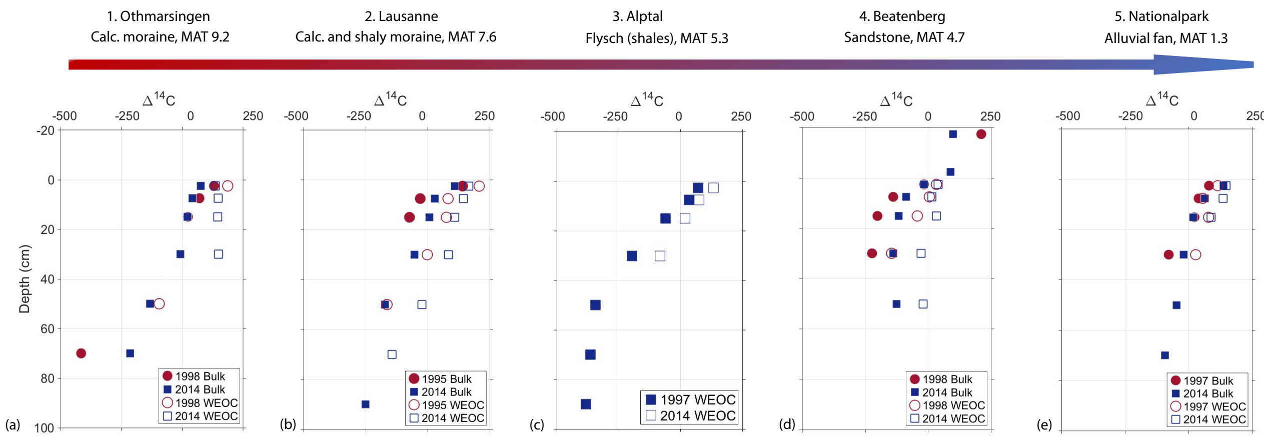

Figure 4 (a-e) Changes in radiocarbon signature of both bulk soil and WEOC over two decades at four sites on a climatic gradient. For Alptal (c) only the 2014 time-point was available. For the warmer locations (Luvisol, Cambisol MAT 9.2-7.6 °C), depletion in bomb-derived radiocarbon occurs in the first five centimeters soil in 2014 as compared to 1995-8. The colder Beatenberg site (Podzol, MAT 4.7 °C) is marked by a clear enrichment of ^{14}C in the mineral soil in 2014 w.r.t. 1997. At the coldest site Nationalpark (Fluvisol, MAT 1.3 °C) almost all samples taken two decades after the initial sampling show an enrichment in radiocarbon signature. WEOC contains bomb-derived carbon in the topsoil in 2014 at all sites.

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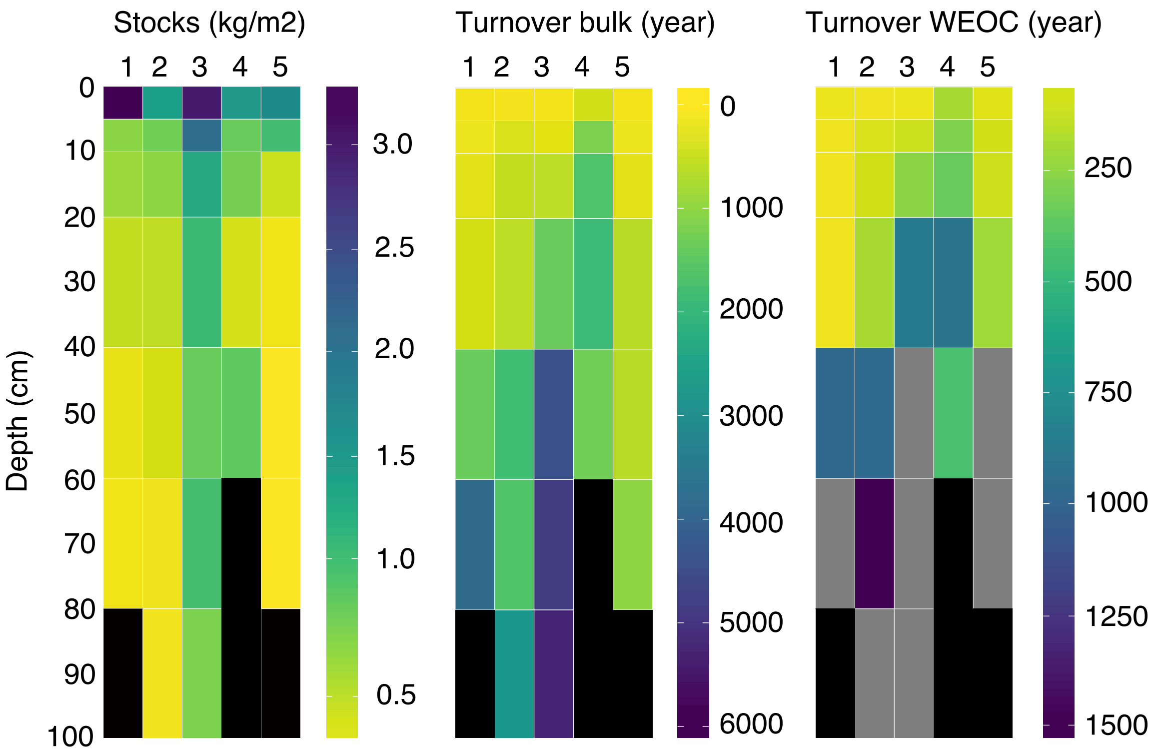
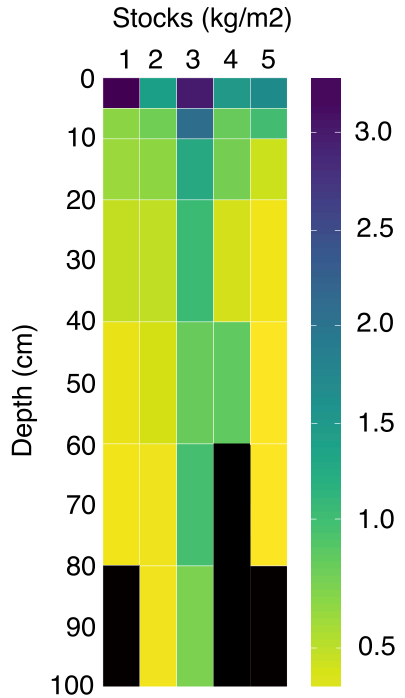


Figure 5 Carbon (a) stocks in the mineral soil kgC/m^2, (b) turnover time bulk soil in years and (c) turnover time water extractable organic carbon soil in years. Locations are ordered from the warmest to coldest sites i.e. (1) Othmarsingen, (2) Lausanne, (3) Alptal, (4) Beatenberg and (5) Nationalpark. Grey boxes indicate absence of material, black boxes indicate the occurence of the C-horizon (poorly consolidated bedrock-derived stony material or bedrock itself).



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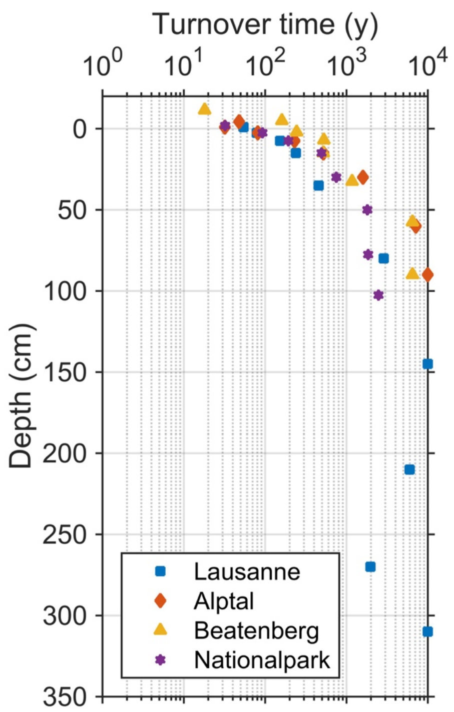

Figure 6 Modeled turnover times (y) of single profiles sampled down to the bedrock between 1995 and 1998. $\Delta^{14}C$ published in Van der Voort et al. (2016). Results indicate presence of petrogenic (bedrock-derived) carbon as modeled turnover time exceeds soil formation since the end of last ice age (10,000 years) in Lausanne (>100 cm, Cambisol) and Alptal (80-100 cm, Gleysol).

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For the turnover estimation, we assumed the system to be in steady state over the modeled period ($\sim 1\times10^4$ years, indicating soil formation since the last glacial retreat (Ivy-Ochs et al., 2009)), hence accounting both for radioactive decay and incorporation of the bomb-testing derived material produced in the 1950's and 1960's (Eq. 1.) (Herold et al., 2014; Torn et al., 2009).

2.4.1 Time-series based determination of likeliest turnover time

In order to optimally constrain carbon turnover estimates for the ^{14}C time-series data, a numerical model was constructed in MATLAB version 2015a (The MathWorks, Inc., Natick, Massachusetts, United States). For the turnover estimation, we assumed the system to be in steady state over the modeled period ($\sim 1\times10^4$ years, indicating soil formation since the last glacial retreat (Ivy-Ochs et al., 2009)), hence accounting both for radioactive decay and incorporation of the bomb-testing derived material produced in the 1950's and 1960's (Eq. 1.) (Herold et al., 2014; Torn et al., 2009).

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Turnover times determined with the numerical optimization match the manually optimized turnover modeling published previously (Herold et al., 2014; Solly et al., 2013).

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(Fig. 3[TSvdV1]). WEOC constitutes only a small portion of the total carbon (<1%), but could be representative for a larger component of rapidly turning over carbon, even in the deep soil (Baisden and Parfitt, 2007; Koarashi et al., 2012). Using the data from the bulk soil and WEOC time-series, the turnover of the slow pool and the relative size of the dynamic pool can be determined.

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Where F_1 is the relative size of the dynamic pool, and F_2 is the relative size of the (more) stable pool. The $k_1$is [TSvdV2]the inverse of the turnover time of the WEOC as determined using the numerical optimisation of Eq. (1) and (2), and k_1 is determined by numerical optimisation. The k_2 is the inverse of the turnover of the slow pool.

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