

1 We thank the reviewers for their thoughtful comments. The original comments are shown in  
2 italic below, and our response and manuscript changes in normal typeface. Page numbers refer to  
3 pages in the final manuscript (provided in the separate file). The annotated manuscript is  
4 provided below the response to comments.

5

6 **Review by Troy Baisden:**

7

8 *This manuscript presents a valuable new C-14 dataset that develops well-constrained soil C*  
9 *turnover estimates for a tropical forest site in Puerto Rico across slope positions, and integrates*  
10 *these with understanding of soil redox conditions. The C-14 dataset is made particularly*  
11 *valuable by the combination of replicated samples at both 1988 and 2012, which make it*  
12 *possible to follow the 'bomb C-14 spike' through the soil C fractions isolated in the study. As the*  
13 *authors point out, the resulting soil C turnover estimates have considerable value globally for*  
14 *understanding the nature of soil C cycling (with quantitative uncertainty estimates) and its*  
15 *relationship to redox. Overall, this is a well-written, topical and useful paper. I suggest only*  
16 *minor revisions.*

17

18 We appreciate your interest in and support for our work.

19

20 *The main thrust of the paper where I would suggest revisions is the way in which the*  
21 *two pool model is emphasised. I've been a strong advocate that single pool models*  
22 *yield misinterpretation of residence times and should be used with great caution. I'm*  
23 *pleased therefore to see Baisden et al (2013) used to emphasise this conclusion. I*  
24 *think this point deserves emphasis because there are still ongoing attempts to publish new single-*  
25 *pool residence times and to restate rather than reinterpret the potentially erroneous results and*  
26 *conclusions involving single-pool residence times already in the literature. But I think the the*  
27 *amount of space used for this emphasis itself can be reduced, and the emphasis better*  
28 *underscored by incorporating some discussion of other studies that have good soil C-14 data*  
29 *and modelling. Given that at least some studies have been undertaken in the temperate and*  
30 *boreal zones, this work now enables some integration of robust turnover estimates across the*  
31 *global range of soils and climate. Such a discussion would allow less emphasis on previous work*  
32 *that is essentially being criticised as 'single-pool' residence times, and more on what we learn*  
33 *from improved data and methods.*

34

35 Agreed. Given the substantial number of recent papers (cited in the manuscript) that continue to  
36 interpret mineral-associated C turnover in terms of a single-pool model, we think that it is  
37 important to emphasize the potential pitfalls of that approach. However, we have condensed our  
38 critique of single-pool models and incorporated additional discussion of insights from more  
39 robust multiple-pool models.

1  
2 We have substantially shortened the discussion of one vs. two pool models on page 24 (lines 10  
3 – 22) and have condensed two paragraphs into one. We added a synthesis paragraph with results  
4 from two-pool models of mineral-associated C from p. 24 line 23 to p. 25 line 11.

5  
6 *Here, I provide some suggestions what a paragraph or two bringing the soil C turnover*  
7 *estimates into a global context might include. (I note two pieces of my own recent work intended*  
8 *to highlight or demonstrate how C-14 can be better connected to rates of C cycling.) First, if the*  
9 *author's narrative wants to begin by outlining assumptions that undermine some single-pool*  
10 *single-time-point C-14 turnover calculations, then Baisden and Canessa (2013) may also be*  
11 *useful to cite.*

12  
13 Yes, Baisden and Canessa (2013) is relevant here and we have included this ref.

14  
15 We have incorporated insights from this paper in several instances on pages 23 and 24.

16  
17 *Methodologically, it is useful to point out that relatively few studies of time-series C-14 have*  
18 *been carried out in forests on colluvium – because forests soils and soils on colluvium are*  
19 *inherently more variable than grassland (or cropland) soils and soils on alluvium or loess, etc.*

20  
21 Agreed, we have now included this point; this is important in light of the small proportion of  
22 samples with very low  $\Delta^{14}\text{C}$ .

23  
24 We have added text on p. 23, lines 1-5.

25  
26 *Schrumpf and Kaiser (2015) had a useful recent paper that illustrates the problem, but confirms*  
27 *that resampling periods of at least 10 years and considerable replication are likely necessary.*  
28 *But the long resampling interval (24 years) in this work is not entirely required.*

29  
30 Agreed; the resampling interval used here was opportunistic based on the availability of paired  
31 archived samples. However, this interval is useful in that it corresponds to a period of greater  
32 atmospheric  $^{14}\text{C}$ , and thus greater sensitivity, than if a narrower time interval were used.

33  
34 We have acknowledged findings from Schrumpf and Kaiser (2015) in the Methods (p. 9 lines 20  
35 - 24) and in the Discussion (p. 23 line 12)

36  
37 *Regarding the use of NPP to make sense of the flow of C through the soil pools in*

1 the C-14 model, Baisden and Keller (2013) introduced the term “synthetic constraint”  
2 for this approach, which is now well handled by SoilR 1.1 (Sierra et al 2014).

3  
4 Here, we did the converse and used soil C turnover times to make inference about the  
5 partitioning of NPP to the mineral-associated C pools at steady state. Despite its value, we do not  
6 have the high temporal resolution NPP data that would be necessary for duplicating the approach  
7 of Baisden and Keller (2013). Please see added text on p. 23, lines 26 to p. 24 line 9.

8  
9 It may be worth commenting more solidly on the consequences of considering a third fast  
10 cycling pool as in Baisden and Keller (2013)?

11  
12 Agreed; there is good evidence that a small portion of mineral-associated C is even more highly  
13 dynamic (i.e., turnover of months – several years), and thus a three-pool model would be  
14 interesting to test. However, we could not fit parameters for a third pool using the two available  
15  $^{14}\text{C}$  timepoints. Also, density separation removed most of the material ( $< 1.8 \text{ g ml}^{-1}$ ) that would  
16 presumably comprise the fastest-cycling pool. Finally, Baisden and Keller (2013) note the poor  
17 sensitivity of  $^{14}\text{C}$  for resolving pools with turnover times  $< 2$  years.

18  
19 We have added text discussing a third pool from p. 23, lines 26 to p. 24 line 9.

20  
21 *Moving to understanding of global patterns where robust two-pool models have been applied to  
22 fractions or bulk soil C-14 time series, it may be useful to recognise that Baisden et al 2002 is  
23 carried out in a xeric grassland chronosequence, but one where the oldest sites are so highly  
24 weathered and poorly drained that they may provide a useful comparison to the tropical sites  
25 such as those reported by the authors. I would also encourage the authors to include C turnover  
26 from Harvard forest (Gaudinski et al 2000; Sierra et al 2012; Sierra et al 2014) if possible  
27 (given the different but still robust method). Froberg et al (2011) provides a good example of a  
28 boreal organic soil in which mineral associated organic matter is not present and a single pool  
29 model works across time-series C-14 measurements. An obvious discussion point is that there  
30 seems to be relatively little variation in mineral-associated organic matter turnover globally, but  
31 that non-mineral-associated organic matter follows the expected temperature response from  
32 Arrhenius relationships.*

33  
34 Agreed; we have incorporated discussion of some of these studies and emerging large-scale  
35 patterns in the revised manuscript. Please see the paragraph on p. 24 line 23 to p. 25 line 11.

36  
37 However, it is important to note that the Gaudinski et al. study also used a one-pool model for  
38 mineral-associated C, and thus likely overestimated turnover times of the mineral associated  
39 “slow pool”. Sierra et al. 2012 built off of this same model; we have included this reference (p.  
40 23, line 12) as another example of paired  $^{14}\text{C}$  studies, despite the fact that they did not use the

1 two timepoints to constrain multiple pools of mineral-associated C. Also, the Froberg et al. study  
2 used a two-pool model (litter and humus) to model O horizon turnover; we have not included the  
3 latter study given that it did not seem directly relevant to the questions at hand.

4

5 *The only other major issue I see is with the clarity of how the modelling and uncertainty analysis  
6 was carried out. I cannot verify exactly what was done, but if I've interpreted it correctly, I think  
7 I agree with all the steps. The main problem with lack of clarity appears to be mixing the  
8 description of the modelling method with the method for estimating uncertainties/errors. I  
9 recommend rewriting the 2 paragraphs beginning at L5 on P903 and considering restructuring  
10 to describe the modelling for a best estimate first, and then the uncertainty second.*

11

12 We have further clarified our modeling approach in the revision. In this paper, we have actually  
13 combined model fitting with the sensitivity analysis in order to avoid assuming a single passive  
14 pool turnover time and 1988  $^{14}\text{C}$  value for each modeled sample. This is a useful approach  
15 because the mean of a non-linear function evaluated across a range of probable values (i.e.,  
16 passive pool turnover times varying from 100 – 1000 years) does not equal the value of a  
17 function evaluated at the mean of those probable values (i.e. 550 years). For example, rather than  
18 simply assuming a single value for passive pool turnover, as has been done in previous studies,  
19 here we assumed that passive pool turnover varied randomly between 100 and 1000 years, and fit  
20 the model for 1000 different randomly chosen turnover times. Then, from the distribution of  
21 modeled parameters, we were able to calculate mean parameter estimates and their uncertainty  
22 (standard deviation) simultaneously. These mean parameter values are subtly distinct from what  
23 we would have estimated using a single assumed passive turnover time (i.e., 550 years).

24

25 Please see added text from p. 11 line 22 to p. 12 line 3 summarizing the modeling approach. We  
26 have also clarified the overall model text with revisions on pages 9, 10, and 12

27

28 *This issue is highlighted by the other reviewer(Jon Sanderman), but my interpretation is  
29 somewhat different in that I think the 1988 data was used appropriately to estimate a mean for  
30 each landscape position, and the more extensive 2012 data used to estimate the uncertainty  
31 around this mean and ultimately estimate uncertainty in calculated turnover times. If this is the  
32 case, I think that's appropriate, but the methods should be further clarified.*

33

34 This is an important point that we have clarified in the revision. We now clearly and  
35 explicitly state that we assume no significant trends in mineral associated C turnover times with  
36 topography in 1988 based on the lack of a pattern in the 2012 samples. As stated in the Methods  
37 in the initial manuscript, we have only four samples from 1988, two each from valleys and  
38 slopes, respectively. Given the substantial variability among  $^{14}\text{C}$  values from the 1988 samples,  
39 simply averaging these two values to derive a mean value for these positions could bias the  
40 model results, and would not allow us to model ridge samples, because we were unable to  
41 analyze samples for this position.

1  
2 The results from 2012 samples provide some useful insight here. Given that  $\Delta^{14}\text{C}$  did not differ  
3 among topographic positions in 2012, it is parsimonious to assume that it did not differ in 1988  
4 either. Thus, we combined the four 1988  $\Delta^{14}\text{C}$  values, and sampled from a normal distribution  
5 defined by the observed mean and standard deviation to serve as a constraint in each model  
6 realization. This approach allows us to assess the impact of varying the assumed 1988  $\Delta^{14}\text{C}$   
7 values on the parameters of interest (slow pool turnover time and fraction slow pool). As we  
8 mention in the original text, there was relatively little impact of varying assumed 1988  $\Delta^{14}\text{C}$  on  
9 our modeled parameters; varying 1988  $\Delta^{14}\text{C}$  across this distribution generated a typical standard  
10 deviation of 3 years in the modeled slow pool turnover time. Thus, we feel justified in this  
11 approach to dealing with uncertainty in 1988  $\Delta^{14}\text{C}$ .

12  
13 We have added or clarified relevant text on p. 9, lines 13 – 24, and p. 12, lines 4 – 12.

14  
15  
16 *There's no need for italicized emphasis about 'absolute' individual comparisons when sites can't  
17 be exactly located 1/4 of a century later. It seems that the uncertainty method is  
18 acceptable/appropriate, although it is by no means the only appropriate approach.*

19  
20 Agreed; we included this section in response to one of the preliminary reviewers for  
21 Biogeosciences, who was skeptical of the validity of a paired  $^{14}\text{C}$  approach given the inherent  
22 spatial heterogeneity of soil C. We eliminated the italic text.

23  
24 *The decision I take issue with is choosing not to model points with  $\Delta^{14}\text{C}$  below the 2012  
25 atmosphere. This may introduce some bias - which should be avoided. I point out that these  
26 results are not unexpected given acceptance of a two pool model where the bulk  $\Delta^{14}\text{C}$  is dragged  
27 down by the presumed millennial pool. They mostly occur in the deeper horizon, which is to be  
28 expected. Keep in mind that an alternative presented in Baisden (2013) is to mathematically  
29 recombine the horizons before modelling the C turnover.*

30  
31 We clarify that we did in fact model the points with  $\Delta^{14}\text{C}$  less than the 2012 atmosphere (5 of the  
32 30 samples from 2012), and these samples are included in the overall mean estimates of slow  
33 pool content; they serve to decrease the overall content of slow pool C relative to passive C  
34 across the landscape. However, we could only estimate one of the two parameters that we had  
35 modeled for the other samples as opposed to modeling both parameters at the same time. Both of  
36 these parameters could not be realistically estimated simultaneously using the observed mean  
37 and variation of  $\Delta^{14}\text{C}$  from the four 1988 samples, given that they yielded turnover times (and  
38 corresponding C inputs) that were ecologically unreasonable. It is unsurprising that the four  
39 samples from 1988 did not exhibit the same degree of heterogeneity of turnover times as did the  
40 30 samples from 2012, due to an increased probability of detecting extreme values in the larger

1 dataset. In 2012, a small portion of the landscape (which was not represented in the 1988  
2 samples ) appears to have less slow-cycling C.

3  
4 Thus, we are left with a situation where we needed to estimate two parameters (slow pool  
5 turnover and fraction slow pool) using only one constraint, 2012  $\Delta^{14}\text{C}$ . One approach would be to  
6 assume that slow pool turnover is similar to the other samples, and that the fraction of the slow  
7 pool declined (i.e.,  $^{14}\text{C}$  is being dragged down by the passive pool as stated above by Baisden).  
8 This is the approach we used in the original manuscript, as this seems most ecologically  
9 reasonable.

10  
11 We agree that mathematically combining the horizons is a useful strategy in many situations, but  
12 this would not avoid the fact that the measured 1988 values for  $^{14}\text{C}$  are still not apparently  
13 representative of these particular 2012 samples. Combining the horizons in this case would still  
14 unrealistically decrease slow pool turnover times if we were to assume that the same 1988 values  
15 hold.

16  
17 We have added relevant text on p. 12 lines 15 – 26 and p. 9 lines 13 – 24.

18  
19 *More detailed and technical comments:*  
20 *P893 L26 It might be useful to include a sentence relating density fractions such as*  
21 *Sollins et al 2009 to the issue of global soils and Al/Fe and C interrelationships.*

22  
23 See added text on p. 3, lines 2 – 5.

24  
25 *P896 L4 The use of ‘historical’ in this sentence is strictly speaking not quite correct. It would be if*  
26 *the C-14 measurements were reported in the historic literature, but should be replaced with*  
27 *‘archived sample’ or a similar term in this case.*

28  
29 Replaced with “archived sample”

30  
31 *P898 L13 ‘associate’-> associated*  
32  
33 The present tense of the verb is required in this context: “various Al (oxy)hydroxide species and  
34 monomeric Al also associate with organic compounds”

35  
36 *P901 L17 Lags are first mentioned much later, but would be appropriate to mention*  
37 *here given the introduction of the model and related discussion on assumptions.*

1  
2 Text added on p. 10, lines 10 – 19.  
3  
4 P904 L15 I've already mentioned the need to better clarify methods in these para-  
5 graphs. Here, I can perhaps guess what Taylor series expansion you refer to, but this should be  
6 specified if it matters. If it doesn't matter quantitatively at the scale of the uncertainties reported, this  
7 use of approximations can be removed to avoid unnecessary confusing detail.  
8  
9 Removed for clarity  
10  
11 P912 L18 I'm not sure why this is 'intriguing'. I don't see why this is surprising given acceptance of  
12 a two pool model where the bulk  $^{14}\text{C}$  is dragged down by the presumed millennial pool. If they occur  
13 in low %C samples, this implies the more stable millennial pool is more dominant in these samples. If  
14 they are not modelled, discussion should be introduced of how their exclusion in the modelling  
15 method introduces potential bias.  
16  
17 "Intriguing" has been removed, and replaced with text discussing dilution of the slow pool by the  
18 passive pool (p. 20, lines 12 – 14). Please see text above for details on modeling of these samples  
19 (they were in fact modeled and included in the site-level means).  
20  
21 **Review by Jon Sanderman:**  
22  
23 This well written manuscript presents a study demonstrating that in humid tropical forest soils  
24 with varying texture and reactive metal concentrations, O<sub>2</sub> availability was  
25 the only factor that could explain variations in soil carbon turnover time. This is an  
26 important finding in that it suggests a hierarchy of controls on decomposition and that  
27 factors which directly limit heterotrophic microbial activity are more important than fac-  
28 tors which just retard organic matter availability.  
29  
30 Thanks for the interest in our work.  
31  
32 While this finding is important, I have a major reservation about the methodological ap-  
33 proach – the authors use time series radiocarbon measurements interpreted using a  
34 steady-state two-pool model which is a very powerful way of assessing decadal scale  
35 turnover time. However, instead of matching the 1988 values with the samples col-  
36 lected in 2012 along the different topographic positions (which represent 3 different soil

1 orders) they have chosen to average the 1988 values in the modeling. The reason for  
2 this averaging was never explained and it seems to invalidate the importance of looking  
3 for differences in carbon cycling along the toposequence.

4

5 This is an important point that we have clarified in the revision. We now clearly and  
6 explicitly state that we assume no significant trends in mineral associated C turnover times with  
7 topography in 1988 based on the lack of a pattern in the 2012 samples. Irrespective of the 1988  
8 data, our 2012 data show that there were no differences in  $\Delta^{14}\text{C}$  values across the topographic  
9 positions, implying that there was no systematic topographic variation in mineral-associated C  
10 turnover times. Thus, parsimony would suggest that there was similarly little topographic  
11 variation in  $^{14}\text{C}$  values in the 1988 samples. As stated in the original manuscript Methods (p 899  
12 line 26-27), we have only four samples from 1988, two each from valleys and slopes,  
13 respectively. Previous paired  $^{14}\text{C}$  studies have relied on similar or even sparser replication for  
14 comparing sites over time (e.g. Trumbore et al. 1996, Science). This is due in part to the  
15 substantial costs associated with accelerator mass spectrometer analyses. Here, we use the 1988  
16 samples to constrain realistic turnover times of 2012 samples in a two-pool model,  
17 acknowledging the uncertainty inherent in pairing  $^{14}\text{C}$  measurements over time given spatial  
18 heterogeneity. Given the observed variability among  $^{14}\text{C}$  values from the 1988 samples  
19 (indicative of spatial heterogeneity but not necessarily related to topography), simply averaging  
20 the two values for each position to derive a mean value by position could bias the model results.  
21 Furthermore, this would not allow us to model ridge samples, given that archived samples for  
22 this position could not be analyzed for this study.

23

24 Thus, for our modeling, we combined the four 1988  $\Delta^{14}\text{C}$  values and sampled from a normal  
25 distribution defined by their observed mean and standard deviation to serve as a constraint in  
26 each of 1000 model realizations for each 2012 sample. This approach allows us to empirically  
27 assess the impact of varying the assumed 1988  $\Delta^{14}\text{C}$  values on the modeled parameters of interest  
28 (slow pool turnover time and fraction slow pool). As we mention in the text (p 908 1:7), there  
29 was relatively little impact of varying assumed 1988  $\Delta^{14}\text{C}$  on our modeled parameters; varying  
30 1988  $\Delta^{14}\text{C}$  across this distribution generated a typical standard deviation of 3 years in the  
31 modeled slow pool turnover time (relative to a mean value of 18 years). Thus, we feel justified in  
32 this approach to dealing with uncertainty in 1988  $\Delta^{14}\text{C}$ .

33

34 Please see added text on page 9, lines 13 – 24, and page 12, lines 4 – 12.

35

36 A lag time in the model should be considered for the 10-20 cm samples as it is highly  
37 unlikely that the current year atmospheric  $^{14}\text{CO}_2$  value is being directly transferred  
38 into the C in this soil layer. This may perhaps help constrain the model for the samples  
39 where the model struggled to find a solution.

1 This is a relevant point that we have clarified with additional text and modeling in the revision.  
2 We have now incorporated a conservative lag of three years for both 0 – 10 and 10 – 20 samples.  
3 Fine roots and DOC are likely to be a major if not primary source of C inputs to the 10 – 20 cm  
4 depth increments, as opposed to particulate leaf litter (e.g. Rasse et al. 2005, Plant and Soil,  
5 269:341–356). Fine roots are abundant in this depth increment (Table 2), and fine roots and DOC  
6 should have a predominantly recent  $^{14}\text{C}$  signature (within several years of atmospheric values;  
7 e.g. Vargas et al. 2009, New Phytologist), especially given the extremely rapid root turnover and  
8 decomposition of buried roots and surface leaf litter in this ecosystem (mean residence times of  
9 0.9 and 0.8 years, respectively; Cusack et al. 2009, Global Change Biology, 15:1339–1355).  
10 Worms contribute actively to bioturbation in this ecosystem as mentioned in the Discussion (p.  
11 20, lines 6 – 9).

12

13 Please see new text on page 10, lines 10 – 19 describing the model lag.

14

15 Unfortunately, adding a lag to the  $^{14}\text{C}$  model further exacerbates the problem of fitting model  
16 solutions to samples with  $\Delta^{14}\text{C}$  less than the 2012 atmosphere. Adding a lag time of five years,  
17 for example, further decreases the modeled slow pool turnover times to unrealistic levels (e.g.  
18 slow pool turnover times of 6 years for a sample where 2012  $\Delta^{14}\text{C} = 30\text{‰}$ ). These short turnover  
19 times are incompatible with measured NPP. As discussed in the response to Troy Baisden's  
20 review, the major problem with modeling the samples with low  $\Delta^{14}\text{C}$  (5 of the 30 2012 samples)  
21 appears to be idiosyncratic spatial variation. This heterogeneity is represented to a minor extent  
22 in the 2012 samples, but not in the 1988 samples, likely due to the differences in sample size. It  
23 is unsurprising that the four samples from 1988 did not exhibit the same degree of heterogeneity  
24 of turnover times as did the 30 samples from 2012, due to a greater capacity for detecting  
25 extreme observations in a larger sample size. For example, the same samples that had low  $^{14}\text{C}$   
26 values in 2012 likely also had low  $^{14}\text{C}$  in 1988, yet we do not have an appropriate match in the  
27 1988 data.

28

29 Thus, we are left with a situation where we need to estimate two parameters (slow pool turnover  
30 and fraction slow pool) without the constraint of a reasonable 1988  $^{14}\text{C}$  value. One approach  
31 would be to assume that slow pool turnover is similar to the other samples, and that the fraction  
32 of the slow pool declined (i.e.,  $^{14}\text{C}$  was dragged down by the passive pool as noted in the review  
33 by Baisden). This is the approach we used in the original manuscript, as this seems most  
34 ecologically reasonable. The fact that these samples have more passive C is reflected in the mean  
35 estimates of slow and passive pool abundance at the site scale.

36

37 Please see added text on page 12, 13 – 26.

38

39 *The authors make a point to say that it is important to focus on multiple pools within  
40 measured fractions versus just the bulk sample and they have focused on the mineral-  
41 associate pool in this study. However, is a fraction that contains nearly 90% of the*

1 organic matter really a distinct fraction from the bulk OM? Is there perhaps a more  
2 meaningful fractionation method for these soils?

3

4 We have further justified our approach in the revision. We contend that separating mineral vs.  
5 particulate soil fractions by density can be quite useful and meaningful in the case of  $^{14}\text{C}$   
6 modeling, because one can remove particulate organic matter with predominantly rapid turnover  
7 times (i.e., less than a year) which has a dramatically different isotopic signature than the  
8 remaining mineral-associated C. Because of the impact of this recently-fixed C, mineral-  
9 associated  $^{14}\text{C}$  values are likely to be distinct from bulk soil  $^{14}\text{C}$  even in soils with little  
10 particulate C, which is important for subsequent  $^{14}\text{C}$  modeling, given the difficulties of modeling  
11 this rapidly-cycling C with an additional  $^{14}\text{C}$  pool given the coarse time resolution of atmospheric  
12  $^{14}\text{C}$  values.

13

14 Others have shown that further separating mineral soil ( $> 1.8 \text{ g ml}^{-1}$ ) by density can yield  
15 fractions with differing  $^{14}\text{C}$  values (e.g. Sollins et al. 2009 Biogeochemistry, 96:209–231,  
16 Giardina et al. 2014, Nature Climate Change, 4:822–827). However, these fractions can still be  
17 difficult to interpret in terms of modeled C pools, as they are unlikely to represent pools with  
18 uniform turnover times. This point is echoed by Baisden and Canessa (2013), who state that: “the  
19 separation of a light density fraction at typical densities of  $1.6\text{--}1.7 \text{ g cm}^{-3}$  or particulate organic  
20 matter by sieving appears highly useful, but that larger ranges of density fractions consume many  
21  $^{14}\text{C}$  AMS measurements with less benefit.” For example, Telles et al. (2003, Global  
22 Biogeochemical Cycles 17, 1040) found significant bomb  $^{14}\text{C}$  in acid hydrolysis residue of heavy  
23 soil fractions, showing that even C that is highly stable to chemical degradation can have a  
24 significant fast-cycling component, and thus cannot be treated as a uniform pool in C cycling  
25 models.

26

27 An alternative approach to physically separating multiple pools of mineral-associated organic  
28 matter is to model these pools mathematically using multiple timepoints, as we did in the present  
29 study. The long time-series studies of  $^{14}\text{C}$  presented by Baisden et al. (2013) suggest that a two-  
30 pool model can describe long-term C dynamics quite well.

31

32 Please see text added on page 23, line 15 to page 24, line 9.

33

34 *I would suggest that the entire section on comparing one-pool versus two-pool model*  
35 *results be dropped. This point has been made in numerous papers and it seems to*  
36 *detract from the main focus of this one.*

37

38 Point well taken that other studies have commented on one-pool vs. two-pool models (e.g.  
39 Trumbore 2000, Ecological Applications 10:399-411), but few have done so in the context of  
40 interpreting mineral-associated C turnover, where one pool models have predominated in the

1 recent literature. We think the one pool vs. two pool distinction for mineral-associated C is quite  
2 important in that numerous recent studies (cited in the manuscript), including some in the  
3 Biogeosciences journal, have continued to interpret mineral-associated C as a single pool. As we  
4 show, this has potentially serious consequences for interpreting C turnover times. The other  
5 reviewer (Troy Baisden) was also supportive of including this section, stating that "I think this  
6 point deserves emphasis because there are still ongoing attempts to publish new single-pool  
7 residence times and to restate rather than reinterpret the potentially erroneous results and  
8 conclusions involving single-pool residence times already in the literature."

9

10 In light of the comments from both reviewers, however, we have condensed our critique of  
11 single-pool models and incorporated additional discussion of insights from more robust multiple-  
12 pool models.

13

14 Please see revised and new text from page 23, line 8 to page 25, line 11.

15

16

17

18

19

20

21

22

23

24

25

26

steven hall 4/4/2015 6:59 PM  
Deleted:

1 **Large fluxes and rapid turnover of mineral-associated carbon**  
2 **across topographic gradients in a humid tropical forest:**  
3 **Insights from paired  $^{14}\text{C}$  analysis**

4 **Steven J. Hall<sup>1\*</sup>, Gavin McNicol<sup>1</sup>, Taichi Natake<sup>1</sup>, Whendee L. Silver<sup>1</sup>**

5 [1] {Department of Environmental Science, Policy, and Management,  
6 University of California-Berkeley}

7 [\*]{now at: **Global Change and Sustainability Center**, University of Utah}

8 Correspondence to: S. J. Hall (steven.j.hall@utah.edu)

9 **Abstract**

10 It has been proposed that the large soil carbon (C) stocks of humid tropical forests result  
11 predominantly from C stabilization by reactive minerals, whereas oxygen ( $\text{O}_2$ ) limitation of  
12 decomposition has received much less attention. We examined the importance of these factors in  
13 explaining patterns of C stocks and turnover in the Luquillo Experimental Forest, Puerto Rico,  
14 using radiocarbon ( $^{14}\text{C}$ ) measurements of contemporary and archived samples. Samples from  
15 ridge, slope, and valley positions spanned three soil orders (Ultisol, Oxisol, Inceptisol)  
16 representative of humid tropical forests, and differed in texture, reactive metal content,  $\text{O}_2$   
17 availability, and root biomass. Mineral-associated C comprised the large majority ( $87 \pm 2\%$ , n =  
18 30) of total soil C. Turnover of most mineral-associated C ( $66 \pm 2\%$ ) was rapid (11 to 26 y,  
19 mean and SE  $18 \pm 3$  y) in 25 of 30 soil samples across surface horizons (0 – 10 and 10 – 20 cm  
20 depths) and all topographic positions, independent of variation in reactive metal concentrations  
21 and clay content. Passive C with centennial – millennial turnover was typically much less  
22 abundant ( $34 \pm 3\%$ ), even at 10 – 20 cm depths. Carbon turnover times and concentrations  
23 significantly increased with concentrations of reduced iron (Fe(II)) across all samples, suggesting  
24 that  $\text{O}_2$  availability may have limited the decomposition of mineral-associated C over decadal  
25 scales. Steady-state inputs of mineral-associated C were statistically similar among the three  
26 topographic positions, and could represent 10 – 25 % of annual litter production. Observed  
27 trends in mineral-associated  $\Delta^{14}\text{C}$  over time could not be fit using the single pool model used in

steven hall 3/17/2015 7:23 PM
<b>Deleted:</b> 71
steven hall 3/17/2015 7:23 PM
<b>Deleted:</b> 3
steven hall 3/17/2015 7:22 PM
<b>Deleted:</b> 9
steven hall 3/17/2015 7:22 PM
<b>Deleted:</b> 9
steven hall 3/17/2015 7:23 PM
<b>Deleted:</b> 20
steven hall 3/17/2015 7:23 PM
<b>Deleted:</b> 2
steven hall 3/15/2015 6:59 PM
<b>Deleted:</b> 26
steven hall 3/15/2015 6:59 PM
<b>Deleted:</b>
steven hall 3/18/2015 9:20 AM
<b>Deleted:</b> 30
steven hall 4/1/2015 12:40 PM
<b>Deleted:</b> litterfall
steven hall 4/1/2015 12:40 PM
<b>Deleted:</b> (estimated by doubling aboveground litterfall)

1 many other studies, which generated contradictory relationships between turnover and  $\Delta^{14}\text{C}$  as  
2 compared with a more realistic two-pool model. The large C fluxes in surface and near-surface  
3 soils documented here are supported by findings from paired  $^{14}\text{C}$  studies in other types of  
4 ecosystems, and suggest that most mineral-associated C cycles relatively rapidly (decadal scales)  
5 across ecosystems that span a broad range of state factors.

6

## 7 1 Introduction

8 Humid tropical forest soils represent a large terrestrial C reservoir ( $\sim 500 \text{ Pg}$ ; Jobbagy and  
9 Jackson, 2000) with the potential to exert important feedbacks on global climate change, yet  
10 much remains unknown about the biogeochemical mechanisms underlying their C dynamics.  
11 Patterns and controls on plant litter decomposition in tropical ecosystems have been well  
12 documented in recent years (e.g. Cusack et al., 2009; Powers et al., 2009), but the turnover  
13 dynamics of tropical soil organic C have received less attention (Trumbore et al., 1995; Torn et  
14 al., 1997; Telles et al., 2003; Marin-Spiotta et al., 2008; Giardina et al., 2014). The majority of  
15 organic matter (56 – 95 %) in humid tropical forests spanning a broad range of soil types has  
16 been shown to be associated with mineral particles (Trumbore et al., 1995; de Camargo et al.,  
17 1999; Telles et al., 2003; Marin-Spiotta et al., 2009; Cusack et al., 2011; Giardina et al., 2014).  
18 Previous work used radiocarbon ( $^{14}\text{C}$ ) modeling to demonstrate that turnover times of mineral-  
19 associated C pools can vary by several orders of magnitude—from decades to millennia—within  
20 and among humid tropical soils (Trumbore et al., 1995; Torn et al., 1997; Telles et al., 2003).  
21 Nevertheless, relatively few studies have described the dynamics of mineral-associated C  
22 turnover in natural humid tropical forests, particularly in relation to proposed biogeochemical  
23 mechanisms of C stabilization. Constraining the turnover times of surface soil C pools and their  
24 biogeochemical drivers in humid tropical soils remains an important research challenge given  
25 their intimate couplings with plant productivity and potentially rapid responses to climate  
26 change.

27 Much of the recent work on soil organic matter stabilization has focused on the importance of  
28 reactive metals and short range-order minerals in protecting C via sorption and precipitation.

29 | Concentrations of iron (Fe) and aluminum (Al) in various soil extractions often correlate strongly

steven hall 3/17/2015 8:07 PM  
**Deleted:** constrained  
steven hall 3/17/2015 7:30 PM  
**Deleted:** implied by our data  
steven hall 3/17/2015 7:24 PM  
**Deleted:** suggest  
steven hall 3/17/2015 7:33 PM  
**Deleted:** that other studies using single-pool  $\Delta^{14}\text{C}$  models of mineral-associated C dynamics, unconstrained by multiple time points, may have systematically underestimated C turnover.

1 with spatial variation in soil C stocks (Torn et al., 1997; Baldock and Skjemstad, 2000; Powers  
2 and Schlesinger, 2002; Kleber et al., 2005; Kramer et al., 2012). Although positive relationships  
3 between reactive Fe and Al and C stocks of mineral soils occur across a broad range of  
4 ecosystems, there have been fewer tests of their relationships with C turnover rates (Sollins et al.  
5 2009). This distinction is important, because positive relationships between reactive metals and  
6 C stocks do not provide information regarding the dynamics of these C pools or the temporal  
7 scale of stabilization. Carbon turnover rates increased strongly with short range-order mineral  
8 content across gradients of precipitation and soil age in allophane-rich tropical Andisols (Torn et  
9 al., 1997; Giardina et al., 2014), and increased with reactive Al content across a sequence of  
10 temperate Mollisols, Alfisols, and Inceptisols (Masiello et al., 2004). However, the relative  
11 impact of reactive Fe and Al on mineral-associated C turnover has received less attention in  
12 highly weathered soils (Ultisols, Oxisols) that are prevalent across humid tropical forests at a  
13 global scale. The content and composition of silicate clay minerals are also likely to contribute to  
14 soil C stabilization in many tropical ecosystems (Feller and Beare, 1997), although their impact  
15 on C turnover times in these soils remains unclear.

16 In addition to reactive metals and minerals, oxygen ( $O_2$ ) availability and redox dynamics can also  
17 influence C dynamics in humid ecosystems where anaerobic microsites are especially prevalent  
18 in surface soils. All else being equal, rates of organic matter decomposition typically decline  
19 under sustained anaerobic conditions (Ponnamperuma, 1972). Recent work has demonstrated  
20 high spatial and temporal variation in  $O_2$  availability in surface soils of humid tropical forests,  
21 and the concomitant importance of anaerobic metabolic processes such as dissimilatory Fe  
22 reduction in maintaining high rates of soil respiration (Silver et al., 1999; Schuur et al., 2001;  
23 Dubinsky et al., 2010; Liptzin et al., 2011; Hall et al., 2013). Soil C stocks increased with  
24 decreasing redox potential across a Hawaiian rainfall gradient even as net primary productivity  
25 declined, suggesting that reducing conditions constrained decomposition and promoted organic  
26 matter accumulation at the landscape scale (Schuur et al., 2001). At smaller spatial scales  
27 ranging from topographic catenas to soil microsites, variation in  $O_2$  availability and reducing  
28 conditions could also have important impacts on rates of soil C cycling.

29 Radiocarbon analysis is a powerful method for modeling the turnover times of slow-cycling C  
30 pools, such as those associated with mineral surfaces, which are generally thought to cycle over

steven hall 4/1/2015 12:44 PM

**Deleted:** P

steven hall 4/1/2015 12:44 PM

**Deleted:** although

steven hall 3/15/2015 6:59 PM

**Deleted:** commonly

steven hall 3/17/2015 8:08 PM

**Deleted:** in

steven hall 3/15/2015 6:59 PM

**Deleted:** However, recent

steven hall 3/15/2015 6:59 PM

**Deleted:** potentially

steven hall 3/15/2015 6:59 PM

**Deleted:** similar

steven hall 3/15/2015 6:59 PM

**Deleted:** exhibit turnover times

1 scales of many decades to millennia. However, a paucity of replicated  $^{14}\text{C}$  measurements in many  
2 studies has often prevented statistical examination of relationships between turnover times and  
3 proposed C stabilization mechanisms. Furthermore, many recent  $^{14}\text{C}$  studies have used turnover  
4 models assuming that operationally-defined organic matter pools (separated by physical or  
5 chemical fractionation) had homogeneous turnover rates. Several studies have demonstrated that  
6 operationally-defined C fractions (e.g. free light, aggregate-occluded light, and mineral-  
7 associated) seldom represent pools with uniform turnover, and the inclusion of multiple pools  
8 with distinct turnover times within measured fractions is often necessary to generate realistic  
9 model results (Trumbore et al., 1995; von Lützow et al., 2007; Baisden et al., 2013; Torn et al.,  
10 2013). Using multiple  $^{14}\text{C}$  measurements over time provides a valuable method for constraining  
11 the turnover of multiple pools within a single measured fraction (Trumbore et al., 1996; Koarashi  
12 et al., 2012; Baisden et al., 2013), particularly in humid tropical ecosystems that exhibit  
13 relatively rapid C turnover (Telles et al., 2003).

14 Here, we assessed the relative importance of several proposed C stabilization mechanisms  
15 (reactive metal content, soil texture, and proxies for reducing conditions) in explaining patterns  
16 of mineral-associated C concentrations, stocks, and turnover, using  $^{14}\text{C}$  measurements and  
17 modeling of density-fractionated samples. We intensively sampled soils across a topographic  
18 catena spanning three soil orders (an Ultisol, Oxisol, and Inceptisol) typical of humid tropical  
19 forest ecosystems (McDowell et al., 2012). Samples represented a range of soil geochemical  
20 characteristics and O<sub>2</sub> dynamics in close spatial proximity (tens of m) (Silver et al., 1999). Our  
21 sampling strategy captured relatively large differences in soil biogeochemical characteristics  
22 while controlling for temperature, a potentially important influence on the turnover of decadal-  
23 cycling C (Townsend et al., 1995). To provide a more rigorous interpretation of our  
24 contemporary soil  $^{14}\text{C}$  data, we constrained models of C turnover with  $^{14}\text{C}$  measurements from  
25 archived (1988) samples from the same site. Finally, to illustrate the importance of considering  
26 multiple pools and  $^{14}\text{C}$  timepoints for fitting C turnover models to data, we compared modeled  
27 turnover times of mineral-associated C between a simple one-pool model commonly used in the  
28 current literature, and a two-pool model constrained by data from archived samples.

29

steven hall 3/15/2015 6:59 PM

**Deleted:** typically

steven hall 3/15/2015 6:59 PM

**Deleted:** previous

steven hall 3/15/2015 6:59 PM

**Deleted:** have

steven hall 3/15/2015 6:59 PM

**Deleted:** Yet, it has been

steven hall 3/15/2015 6:59 PM

**Deleted:** critical

steven hall 3/17/2015 1:31 PM

**Deleted:** historical data

1    2    **Methods**

2    **2.1 Site description**

3    We sampled ridge, slope, and valley topographic positions along a hillslope catena in the Bisley  
4    Watershed of the Luquillo Experimental Forest, Puerto Rico, an NSF-funded Long Term  
5    Ecological Research and Critical Zone Observatory site. Our sites (250 masl, 18.3157 °N,  
6    65.7487 °W) support a lower montane tabonuco (*Dacryodes excelsa*) forest with a mean annual  
7    temperature of 23 °C. Annual precipitation averaged 3800 mm yr<sup>-1</sup> and varied between 2600 –  
8    5800 mm yr<sup>-1</sup> from 1989 - 2011 (Scatena et al., unpublished data); see McDowell et al. (2012)  
9    for a detailed site description. Soils formed from volcaniclastic sedimentary rocks derived from  
10   andesitic to basaltic material, and catena positions represent three different orders in the USDA  
11   taxonomy (Soil Survey Staff, 2002). Ridges are dominantly Ultisols (Typic Haplohumults),  
12   slopes are Oxisols (Inceptic and Aquic Hapludox), and valleys are Inceptisols (Typic  
13   Epiaquepts). Catena positions differed in surface soil O<sub>2</sub> concentrations measured at 10 cm depth  
14   over several years (Silver et al., 1999). Mean O<sub>2</sub> concentrations decreased from ridges to slopes  
15   to riparian valleys (19, 16, and 10 % O<sub>2</sub>, respectively), with increasing temporal variability such  
16   that valley soils frequently experienced low O<sub>2</sub> ( $\leq 3\%$ ). Anaerobic microsites are present in all  
17   topographic positions, indicated by periodic methane emissions (Silver et al., 1999). Given that  
18   Fe oxides represent the most abundant anaerobic terminal electron acceptor in these soils (Hall et  
19   al., 2013) we used measurements of Fe(II) to provide an index of reducing conditions at the scale  
20   of individual soil samples. We acknowledge that soil Fe(II) concentrations constitute a one-time  
21   measurement in a dynamic redox environment, but indices of O<sub>2</sub> availability and other redox  
22   reactions tended to vary consistently with Fe(II) concentrations among plots and over time (Hall  
23   et al., 2013).

steven hall 3/15/2015 6:59 PM

**Deleted:**

24    **2.2 Soil sampling and analysis**

25    We dug a soil pit in each catena position to establish relationships between soil depth and  
26   horizons. Mineral soil A horizons spanned depths of 0 – 10, 0 – 9, and 0 – 10 cm in the ridge,  
27   slope, and valley, whereas B1 horizons were at depths of 10 – 22, 9 – 25, and 10 – 20 cm,  
28   respectively. Visual inspection of soil cores showed that A horizon depths were reasonably  
29   consistent among samples, but total soil depth varied among plots and occasionally did not

1 exceed 20 cm in the riparian valley (due to the sporadic presence of buried boulders). We thus  
2 sampled soils at depths of 0 – 10 and 10 – 20 cm, which contain the large majority of roots and  
3 organic matter in this ecosystem (Silver and Vogt, 1993).

4 We established five 0.25-m<sup>2</sup> plots in each topographic position (ridge, slope, and valley)  
5 for a total of 15 plots in the same field sites previously examined by Silver et al. (1999). Plots  
6 were randomly placed within 5 – 10-m intervals along a 50 m linear transect. On the slope  
7 position, the slopes of individual plots varied between 25 and 40°. Surface organic horizons were  
8 usually sparse, and any O horizon material was removed prior to coring. In each plot and depth  
9 increment (total n = 30), we collected a total of four replicate 6-cm diameter soil cores. Two  
10 cores from each plot were sampled in July 2011 to determine bulk density and fine root biomass,  
11 and two additional cores were sampled in February 2012 for C density fractionation and  
12 chemical analysis.

13 The cores for chemical analyses were immediately homogenized in the field, and separate  
14 subsamples extracted in 0.5 M hydrochloric acid (HCl) and 0.2 M sodium citrate/0.05M ascorbic  
15 acid solutions within 1 - 2 min of sampling. The low pH of the HCl extraction inhibits oxidation  
16 of Fe(II). Soil subsamples (3 g dry mass equivalent) were immersed in a 1:10 ratio with HCl,  
17 vortexed, shaken for 1 hour, and filtered to 0.22 µm. Concentrations of Fe(II) were measured  
18 using a colorimetric ferrozine assay and corrected for Fe(III) interference (Viollier et al., 2000).  
19 We used Fe(II) concentrations as an index of reducing conditions at the scale of soil samples,  
20 given that Fe reduction represents the dominant anaerobic respiratory process in this system  
21 (Dubinsky et al., 2010), and that Fe(II) readily oxidizes in the presence of O<sub>2</sub>. Separate  
22 subsamples were extracted in the field with sodium citrate/ascorbate solution (Reyes and  
23 Torrent, 1997) to provide an estimate of reactive Fe oxides (Fe<sub>ca</sub>) and associated Al (Al<sub>ca</sub>) in short  
24 range-order minerals and organic fractions. Short range-order Fe and organo-Fe complexes are  
25 analytically indistinguishable in chemical extractions, with nano-scale Fe (oxy)hydroxides often  
26 dominating (Thompson et al., 2011); thus we subsequently refer to Fe<sub>ca</sub> as “reactive.” Aluminum  
27 is not redox active in soils but frequently substitutes in Fe minerals, and various Al  
28 (oxy)hydroxide species and monomeric Al also associate with organic compounds. In the  
29 absence of detailed Al speciation data, we similarly refer to Al<sub>ca</sub> as “reactive.” Soil subsamples  
30 (1.5 g dry mass equivalent) were immersed in a 1:30 ratio with citrate/ascorbate solution,

1 vortexed, shaken for 18 hours, and centrifuged for 10 min at 1500 rcf.

2 Field extractions likely yield the most representative patterns of reactive metal abundance  
3 due to rapid crystallization of short range-order minerals upon drying, but we also extracted  
4 oven-dried (105° C) and ground heavy density fractions (described below) with acid ammonium  
5 oxalate solution in the dark at pH 3 to allow comparison with previous studies (e.g. Kleber et al.,  
6 2005). Drying soils leads to mineral crystallization, thus decreasing extractable metal  
7 concentrations. However, oxalate extraction of moist samples with high Fe(II) is undesirable and  
8 can promote catalytic extraction of crystalline Fe oxides (Phillips et al., 1993). Subsamples (0.5  
9 g) were extracted for two hours in 30 ml of ammonium oxalate solution. For all of the above  
10 extractions, concentrations of Fe and Al were analyzed in triplicate using an inductively coupled  
11 plasma optical emission spectrometer (ICP-OES; Perkin Elmer Optima 5300 DV, Waltham,  
12 Massachusetts). Total Fe measured colorimetrically in HCl extractions agreed within 1 % of  
13 ICP-OES measurements. Soil pH was measured in 1:2 slurries of soil and deionized water.  
14 Additional subsamples of field-moist soil from each plot were analyzed for particle size by the  
15 hydrometer method (Gee and Bauder, 1986). Samples (50 g dry mass equivalent) were passed  
16 through a 2 mm sieve and immersed for 16 h in sodium hexametaphosphate solution (50 g l<sup>-1</sup>) to  
17 chemically disperse aggregates, which were then physically dispersed in an electric mixer. We  
18 measured changes in soil suspension density over 24 hours to calculate clay, sand, and silt  
19 fractions.

20 We assayed the two remaining replicate 6 cm diameter cores from each plot and depth increment  
21 for fine root biomass and bulk density, respectively. Fine roots (< 2 mm diameter) were  
22 separated from soil by wet sieving and separated into live and dead fractions based on visual  
23 observations of turgor and tensile strength. Roots were thoroughly washed in deionized water  
24 and dried at 65°C. To determine bulk density from the intact cores, we carefully removed any  
25 coarse roots and rocks (which were rare) after cleaning them to retain all soil. We estimated the  
26 volume of coarse roots > 5 mm using a cylindrical approximation, and measured the volumetric  
27 water displacement of rocks with diameter > 2 mm; these corrections minimally affected our  
28 bulk density measurements (mean relative change of 2 %). Soils were dried at 105° C to constant  
29 mass, and bulk density was calculated as dry soil mass divided by coarse root and rock-corrected  
30 sample volume.

1    **2.3 Soil density fractionation**

2    We separated soil organic matter from each plot/depth increment ( $n = 30$ ) into three fractions  
3    based on density and occlusion: (1) a free light fraction consisting of low-density ( $< 1.85 \text{ g cm}^{-3}$ )  
4    organic matter not contained within aggregates, (2) an occluded light fraction, comprising low-  
5    density organic matter released from aggregates following sonication, and (3) a heavy fraction  
6    with density  $> 1.85 \text{ g cm}^{-3}$  associated with soil minerals. Soil cores for density fractionation were  
7    stored at field moisture in sealed polyethylene bags at  $4^\circ \text{ C}$  and analyzed within 6 months of  
8    collection. We used the same protocol to fractionate four air-dried archived samples (0 – 10 cm  
9    increment) that were collected in 1988 from nearby slope and riparian valley plots representative  
10   of the plots sampled in 2012 (Silver et al., 1994). The samples from 1988 served as a benchmark  
11   for a two-pool  $^{14}\text{C}$  model, and were not intended to describe ecosystem-scale spatial or temporal  
12   patterns in the C content of density fractions. The fractionation assay followed Swanston et al.  
13   (2005) as modified for Fe-rich soils (Marin-Spiotta et al., 2008). We passed samples (20 g dry  
14   mass equivalent) through a 4.75 mm sieve to remove coarse litter fragments while maintaining  
15   aggregate structure. The free light fraction was separated by flotation after immersing soils in  
16   sodium polytungstate at a density of  $1.85 \text{ g cm}^{-3}$ . The occluded light fraction was similarly  
17   obtained after mixing and sonicating soils to disrupt aggregates, with a total energy input of 200  
18    $\text{J ml}^{-1}$ . The heavy fraction consisted of the remaining mineral-associated organic matter. Mass  
19   recovery of density fractions from 2012 samples measured  $100 \pm 1\%$  (mean  $\pm$  SE) of the initial  
20   soil mass; mass recovery greater than 100 % may reflect residual sorption of a small amount of  
21   sodium polytungstate or heterogeneous soil moisture content of the moist soil samples. Recovery  
22   measured  $95.0 \pm 0.3\%$  for the air-dried 1988 soils. Masses of free and occluded light fractions  
23   may differ between the air-dried 1988 samples and field-moist 2012 samples due to the notable  
24   effects of air-drying on aggregate structure, but variation in sample moisture during fractionation  
25   was less likely to impact the partitioning of C between particulate C and mineral-associated  
26   fractions. Oven-dried ( $105^\circ \text{ C}$ ) density fractions were analyzed in duplicate for C concentrations  
27   and  $\delta^{13}\text{C}$  isotopic ratios relative to V-PDB on a Vario Micro elemental analyzer in-line with an  
28   Isoprime 100 isotope ratio mass spectrometer (Elementar, Hanau, Germany).

steven hall 3/15/2015 6:59 PM  
**Deleted:** is

1    **2.4 Radiocarbon measurements and modelling**

2    We measured radiocarbon content of the 30 heavy fraction samples from 2012 and the four  
3    samples from 1988 on the Van de Graaff FN accelerator mass spectrometer at the Center for  
4    Accelerator Mass Spectrometry at Lawrence Livermore National Laboratory, Livermore CA.  
5    Heavy C fractions were subsampled into quartz tubes which were evacuated, flame-sealed, and  
6    combusted in the presence of copper oxide and silver. Resulting CO<sub>2</sub> was reduced to graphite on  
7    iron powder in the presence of H<sub>2</sub> at 570 °C. Corrections were applied for mass-dependent  
8    fractionation using measured δ<sup>13</sup>C, for sample preparation background using <sup>14</sup>C-free coal, and  
9    for <sup>14</sup>C decay since 1950. We report final radiocarbon values in Δ<sup>14</sup>C notation with an average  
10   precision of 3 ‰ (Stuiver and Polach, 1977).

11   To infer temporal trends in Δ<sup>14</sup>C, previous studies have used representative samples collected at  
12   locations within 100 m of one another (Trumbore et al., 1996), and over even greater distances  
13   when similarities in ecosystem state factors could be maintained (Baisden et al., 2013). However,  
14   soils display fine-scale spatial heterogeneity over scales of cm – m that can potentially  
15   complicate the assessment of temporal trends. This heterogeneity precluded absolute  
16   comparisons between individual samples from 1988 and 2012, as 1988 sampling locations could  
17   not be located to that degree of resolution. Furthermore, we could only analyze four 1988  
18   samples (two slope and two valley soils) for this study. Nevertheless, as shown in the Results  
19   (3.2) the Δ<sup>14</sup>C values from our 2012 samples (n = 30) showed no systematic spatial variation  
20   across the catena, supporting the idea that the 1988 samples were broadly representative of the  
21   site as a whole. The 24-year period between sample collections allowed for sensitive detection of  
22   changes in Δ<sup>14</sup>C over time (Schrumpf and Kaiser, 2015).

steven hall 3/15/2015 6:59 PM  
Moved (insertion) [1]

23   We modeled the turnover time of mineral-associated C using a time-dependent steady-state  
24   difference equation model of soil Δ<sup>14</sup>C dynamics in conjunction with Δ<sup>14</sup>C of atmospheric CO<sub>2</sub>  
25   (Trumbore 1993; Torn et al., 2009). We assumed that Δ<sup>14</sup>C of a soil organic matter pool in a  
26   given year is a function of soil Δ<sup>14</sup>C from the previous year minus losses from decomposition and  
27   radioactive decay, plus additions of recently-fixed CO<sub>2</sub> with atmospheric Δ<sup>14</sup>C, represented by  
28   the following equation:

29   
$$F'_{\text{soil pool},t} = k F'_{\text{atm},t} + F'_{\text{soil pool},t-1} (1 - k_{\text{soil pool}} - \lambda) \quad (1)$$

1 Here,  $F'$  equals  $\Delta^{14}\text{C}/1000 + 1$  at time  $t$ ,  $k$  is the decomposition rate constant, and  $\lambda$  is the  
2 radioactive decay constant; the subscripts atm and soil pool indicate the atmosphere and an  
3 arbitrary soil C pool, respectively. We used a time series of atmospheric  $\Delta^{14}\text{CO}_2$  measurements  
4 from 1511 – 1950 (Stuiver et al., 1998) and 1950 – 2009 (Hua et al., 2013) for atmospheric zone  
5 2, which includes Puerto Rico, and assumed a 5 ‰ annual decline in atmospheric  $\Delta^{14}\text{C}$  from  
6 2010 – 2012 (Fig. 1). This model implies first-order decay such that the inverse of the modeled  
7 decomposition rate constant represents the mean turnover time of a soil C pool. We fit models  
8 with a lag of three years between atmospheric  $^{14}\text{C}$  values and the corresponding  $^{14}\text{C}$  value of soil  
9 C inputs to provide a conservative estimate of C residence in plant biomass, and also present  
10 modeling results with no time lag. In both cases, turnover times represent the combined plant/soil  
11 system. Fine roots are increasingly thought to represent a dominant source of soil C (Rasse et al.,  
12 2005). Here, a three-year lag accounted for the fact that the age of C in fine root tissue often  
13 varies over several years (Vargas et al., 2009), and that dead roots decompose in < 1 year at this  
14 site (Cusack et al., 2009). We used the same time lag for the 0 – 10 and 10 – 20 cm depths given  
15 that fine roots were abundant in both increments, and bioturbation by earthworms was extremely  
16 common across both depth increments at this site. Finally, the model assumed†that mineral-  
17 associated C pools were at steady state, an assumption supported by annual soil C measurements  
18 in nearby plots over a decade that included severe storm events (Teh et al., 2009), and  
19 similarities in forest above-ground biomass since the late 1980's (Heartsill Scalley et al., 2010).

20 A single-pool model for the mineral-associated (heavy) C fraction could not reproduce the  
21 observed temporal trend of  $\Delta^{14}\text{C}$  between 1988 and 2012. Therefore, we assumed that the  
22 mineral-associated C fraction was the sum of two pools that cycled over different timescales: a  
23 slow pool with decadal turnover, and a passive pool with centennial to millennial turnover  
24 (Trumbore et al., 1995; Telles et al., 2003; Baisden et al., 2013).

$$25 F'_{\text{heavy},t} = P_{\text{slow}} F'_{\text{slow},t} + (1 - P_{\text{slow}}) F'_{\text{passive},t} \quad (2)$$

26 Here,  $P_{\text{slow}}$  represents the proportion of mineral-associated C in the slow pool. Incorporating two  
27 separate pools of organic matter in the mineral-associated fraction was necessary to fit  
28 decomposition rate constants to the observed  $\Delta^{14}\text{C}$  values in 2012 and 1988, and a two-pool  
29 model is consistent with long-term observations of soil organic matter dynamics constrained by

steven hall 3/15/2015 6:59 PM

**Deleted:** per mil

steven hall 3/17/2015 6:56 PM

**Deleted:** The modeled turnover time of a soil C pool represents a maximum estimate given that plant C does not always enter the soil during the same year that it is fixed via photosynthesis; thus, turnover times represent the residence time of C in the plant/soil system. The model also

steven hall 3/17/2015 8:16 PM

**Deleted:** s

steven hall 3/17/2015 8:16 PM

**Deleted:** are

steven hall 3/15/2015 6:59 PM

**Deleted:** dense

steven hall 4/1/2015 3:37 PM

**Deleted:** is

1 frequent  $\Delta^{14}\text{C}$  measurements (Baisden et al., 2013). The passive C fraction contains negligible  
2 modern  $^{14}\text{C}$  and acts to dilute the modern  $^{14}\text{C}$  signal of the faster-cycling slow C pool.  
3 Thus, our model had three free parameters: turnover times of the slow and passive pools, and the  
4 proportion of mineral-associated C in the slow pool. We estimated two of the three parameters,  
5 the proportion of slow pool C and its turnover time, with our data (soil  $\Delta^{14}\text{C}$  from 1988 and  
6 2012). For the third parameter, turnover time of the passive pool, we assumed a liberal range of  
7 values (100 – 1000 y) during the model fitting process (described below). Previous studies have  
8 assumed passive C turnover times of several hundred to 100,000 y (Trumbore et al., 1995; Telles  
9 et al., 2003; Baisden et al., 2013). However, empirically determining passive C ages is difficult.  
10 Radiocarbon analysis of C remaining after acid hydrolysis has been used to define passive C age,  
11 yet even hydrolysis residue may contain bomb  $^{14}\text{C}$ , indicative of faster turnover times (Telles et  
12 al., 2003). Here, we assumed a relatively faster distribution of passive turnover times because  
13 these allow for realistic increases in model uncertainty. For example, allowing passive C  
14 turnover to increase from 1000 to 100,000 y in our models had very little effect on mean slow  
15 pool turnover (< 0.2 y), but substantially *decreased* the variance by decreasing the overall  
16 proportion of model runs that had shorter turnover times of the passive pool, given that shorter  
17 turnover times have more leverage on model results.

18 Here, we modeled the slow-pool turnover time and slow pool fraction for each 2012 sample  
19 using an approach that combined parameter estimation with sensitivity analysis, thus avoiding  
20 the assumption of a single passive pool turnover time and 1988  $\Delta^{14}\text{C}$  value for each 2012 sample.  
21 For example, rather than simply assuming a single value for passive pool turnover as has been  
22 done in previous studies, here we assumed that passive pool turnover varied randomly between  
23 100 and 1000 years, and fit model parameters for each of 1000 different randomly chosen  
24 turnover times. We used a similar approach to vary the 1988  $\Delta^{14}\text{C}$  value assumed in each model  
25 iteration. Then, using the distribution of parameter values calculated from the 1000 model  
26 iterations for each sample, we were able to estimate mean values and their uncertainty (standard  
27 deviation). The resulting mean parameter values are subtly distinct from what we would have  
28 estimated by applying a single “best estimate” of mean 1988  $\Delta^{14}\text{C}$ , due to the non-linear trend in  
29 atmospheric  $\Delta^{14}\text{C}$  values.

steven hall 3/15/2015 6:59 PM

**Deleted:** ), and assessed impacts on the other parameters by randomly selecting values from a uniform distribution.

1 We varied the assumed value of 1988  $\Delta^{14}\text{C}$  in each model realization by sampling from a normal  
2 distribution defined by the observed values (mean and standard deviation of  $\Delta^{14}\text{C} = 186 \pm 10\%$ ),  
3 and excluding the extreme 5% of the distribution to ensure model convergence. We acknowledge  
4 that fitting each of the 2012 samples to the same distribution of  $\Delta^{14}\text{C}$  in 1988 is imperfect, as  
5  $\Delta^{14}\text{C}$  values in 1988 and 2012 would likely be correlated if they could be estimated on precisely  
6 the same sample. However, the Monte Carlo approach employed here allowed us to assess the  
7 impacts of variation in assumed 1988  $\Delta^{14}\text{C}$  values on modeled turnover times. Varying 1988  $\Delta^{14}\text{C}$   
8 across the observed distribution had relatively minor impacts on modeled slow pool turnover,  
9 affecting turnover times of each sample by an average of three years (Supplemental Table 1).

10 For each 2012 soil sample and randomly generated parameter set, we calculated the turnover  
11 time and proportion of slow-cycling C with Equations 1 and 2, adjusting  $k_{slow}$  and  $P_{slow}$  until  
12 modeled  $\Delta^{14}\text{C}$  matched the measured 2012 value and assumed 1988 value. Five 2012 samples  
13 had  $\Delta^{14}\text{C}$  less than the 2012 atmosphere, and slow pool turnover times for these samples could  
14 not be realistically estimated using the available 1988 samples (Fig. 1). It is unsurprising that the  
15 four samples from 1988 did not exhibit the same degree of heterogeneity in turnover times as did  
16 the 30 samples from 2012, due to an increased probability of detecting extreme values in the  
17 larger dataset. For the five samples with smaller  $\Delta^{14}\text{C}$  values, we made the simplifying  
18 assumption that these values were primarily caused by an increasing abundance of the passive  
19 pool, and that slow pool turnover times were of similar magnitude as the other samples. To  
20 estimate proportions of slow vs. passive C in these five samples, we randomly selected 1000  
21 slow pool turnover times from the previously modeled distribution ( $18 \pm 3$  y) in addition to the  
22 other randomly selected parameters described above. Then, we solved for  $P_{slow}$  without  
23 constraining the model to 1988 data. All modeling was conducted with R version 3.0.2, and free  
24 parameters were fit using the “optim” function with the Nelder-Mead method. After estimating  
25  $k_{slow}$  and  $P_{slow}$ , we calculated annual C inputs to the slow pool under steady state by dividing  
26 slow pool C stocks by turnover times. Standard errors reported below include the sum of  
modeling variation and spatial variation.

steven hall 3/15/2015 6:59 PM

**Moved up [1]:** To infer temporal trends in  $\Delta^{14}\text{C}$ , previous studies have used representative samples collected at locations within 100 m of one another (Trumbore et al., 1996), and over even greater distances when similarities in ecosystem state factors could be maintained (Baisden et al., 2013).

steven hall 3/15/2015 6:59 PM

**Deleted:** However, soils display fine-scale spatial heterogeneity that can potentially complicate the assessment of temporal trends, even when samples are collected in close spatial proximity. The differences in  $\Delta^{14}\text{C}$  ... [1]

steven hall 3/17/2015 12:34 PM

**Deleted:** Thus, we generated 1000 di ... [2]

steven hall 3/15/2015 6:59 PM

**Deleted:** are

steven hall 3/15/2015 6:59 PM

**Deleted:** .

steven hall 3/15/2015 6:59 PM

**Deleted:** importance

steven hall 3/15/2015 6:59 PM

**Deleted:** the assumed

steven hall 3/17/2015 7:41 PM

**Deleted:** four

steven hall 3/15/2015 6:59 PM

**Deleted:**  $\Delta^{14}\text{C}$  in

steven hall 3/15/2015 6:59 PM

**Deleted:** and 2012

steven hall 3/15/2015 6:59 PM

**Deleted:** ; therefore,

steven hall 3/17/2015 7:42 PM

**Deleted:** constrained

steven hall 3/15/2015 6:59 PM

**Deleted:** the bomb- $^{14}\text{C}$  enriched

steven hall 3/15/2015 6:59 PM

**Deleted:** We

steven hall 3/15/2015 6:59 PM

**Deleted:** variation in the proportion o ... [3]

steven hall 3/15/2015 6:59 PM

**Deleted:** C was primarily responsible ... [4]

steven hall 3/15/2015 6:59 PM

**Deleted:** cycling

steven hall 3/17/2015 7:43 PM

**Deleted:** 9

steven hall 3/17/2015 7:43 PM

**Deleted:** .9

steven hall 3/15/2015 6:59 PM

**Deleted:** This calculation involves a ... [5]

1    **2.5 Statistical analysis**

2    We assessed relationships between biogeochemical variables and C cycling among individual  
3    samples and catena positions using linear mixed effects models fit with the lme function in R  
4    (Pinheiro et al., 2014). Interactions among topographic positions and depths were assessed by  
5    assigning a distinct factor level to each position/depth combination, with post-hoc comparisons  
6    using the Tukey method. To assess topographic variation in C from 0 – 20 cm, C was summed  
7    by depth increment for each plot. In addition, we fit linear mixed effects models for mineral-  
8    associated C concentrations and stocks in individual samples. Potential predictor variables  
9    included  $\Delta^{14}\text{C}$  and the other measured biogeochemical variables described above. We normalized  
10   predictor variables by mean and standard deviation to allow comparison of their relative  
11   importance (analogous to Pearson's r). Models included plots as potential random effects to  
12   account for any correlation between depth increments in a given plot. We selected the optimal  
13   random effect structure by comparing the Akaike Information Criterion (AIC) of saturated  
14   models fit using restricted maximum likelihood. Including random effects did not improve fit, so  
15   we proceeded with multiple regression. We selected fixed effects on models fit using maximum  
16   likelihood using backwards selection and AIC with a correction for small sample size, and  
17   reported models with similar goodness of fit.

18

19    **3 Results**

20    **3.1 Topographic patterns in soil C and biogeochemical variables**

21    Mineral-associated C comprised the dominant C fraction across all topographic positions,  
22    representing  $90 \pm 1$ ,  $89 \pm 1$ , and  $80 \pm 4$  % of total soil C in the ridge, slope, and valley soils,  
23    respectively (Table 1, Fig. 2, Supplemental Table 1). Mineral-associated C content (soil mass  
24    basis) in ridge 0 – 10 cm soils was 1.45 times greater ( $p < 0.01$ ) than slopes and valleys (Fig. 2,  
25    Table 1). Mineral associated C content was also greater in surface (0 – 10 cm) than subsurface  
26    (10 – 20 cm) soils on ridges and slopes, but did not differ by depth in valleys. Free light C  
27    content was similar among topographic positions. Valleys had significantly more occluded light  
28    C when considering both depth increments together, measuring 3.6 times the occluded light C on

1 ridge and slope soils (Fig. 2;  $p < 0.05$ ). Carbon concentrations of the individual fractions  
2 generally showed similar trends with topography and depth as soil mass-based C content (Table  
3 1). Heavy fraction C concentrations were greatest in ridge surface soils, and occluded light C  
4 concentrations were greatest in valley soils. Heavy fraction C content was lower in the four 1988  
5 samples than in many of the 2012 samples (Supplemental Table 2), but replication was  
6 insufficient to assess any temporal changes.

7 Paralleling the topographic patterns in mineral-associated C, most measured biogeochemical  
8 indicators varied strongly and significantly across the catena (Table 2, Supplemental Table 3).  
9 Ridges supported the highest fine root biomass, which declined in the slopes and valleys. Soil pH  
10 was significantly more acidic ( $4.6 \pm 0.0$ ) in the ridge and slope samples than in the valleys ( $5.2 \pm$   
11 0.1;  $p < 0.0001$ ), regardless of depth increment. Clay content was consistently high in ridge and  
12 slope samples ( $41 \pm 3\%$ ) and significantly lower in the valley soils ( $23 \pm 2\%$ ;  $p < 0.0001$ ).  
13 Conversely, sand content was low in ridge and slope samples but was 2 – 3 times greater in the  
14 valley soils, and silt content was similar across topographic positions. Clay content increased and  
15 silt declined with depth in the ridge soils, while texture did not vary significantly with depth in  
16 the other topographic positions.

17 Reactive Fe and Al showed distinct patterns among topographic positions and depths that varied  
18 by chemical extraction (Table 2). Concentrations of Fe and Al extracted in the field by citrate-  
19 ascorbate solution ( $\text{Fe}_{\text{ca}}$  and  $\text{Al}_{\text{ca}}$ ) were significantly (more than two-fold) greater in ridge than in  
20 slope or valley surface soils. Oxalate extractions of dried soil heavy fractions yielded  
21 significantly greater Al in ridge surface soils than most other positions/depths, but  $\text{Fe}_{\text{ox}}$  varied  
22 little among samples. Concentrations of  $\text{Fe}_{\text{ox}}$  and Fe(II) were the only measured biogeochemical  
23 variables that did not vary by topographic position. Fe(II) concentrations were variable and  
24 consistently measurable, indicative of the ubiquitous presence of reducing microsites across the  
25 plots.

26 Mineral-associated C stocks varied almost three-fold among samples, between 1230 - 3030 g C  
27  $\text{m}^{-2}$  (mean  $2150 \pm 100$ ) in each 10-cm depth increment (Table 1, Supplemental Table 1).  
28 Summing the two 10-cm depth increments in each plot yielded mineral-associated C stocks of  
29 3310 – 6630 g C  $\text{m}^{-2}$  (mean  $5010 \pm 290$ ) to 20 cm depth. Despite the large and significant

1 differences in C concentrations, mass based C content, and reactive Fe and Al across the catena,  
2 mineral-associated C stocks were statistically similar among the topographic positions (Fig. 2)  
3 due to co-variation in bulk density. Bulk density was significantly lower in ridge 0 – 10 cm soils  
4 than all other positions, whereas mineral-associated C concentrations and mass based C content  
5 were greatest in these samples ( $p < 0.01$ ; Fig. 2, Table 1). Mass-based mineral-associated C  
6 content and bulk density also negatively co-varied at the scale of individual samples ( $R^2 = 0.49$ ,  
7  $p < 0.0001$ ).

8 Of the density fractions measured, only the occluded light fraction C stocks differed by  
9 topographic position; these were significantly greater in valleys when considering both depth  
10 increments together ( $p < 0.05$ ). Total soil C stocks (sum of all three fractions in each depth  
11 increment) did not significantly differ among positions in either depth increment. Similarly, total  
12 C stocks summed to 20 cm in each plot ( $\text{g C m}^{-2}$ ) did not differ significantly, although valleys  
13 tended to be greatest ( $5558 \pm 511$ ), ridges intermediate ( $5112 \pm 337$ ), and slopes lowest ( $4370 \pm$   
14 579; Table 1).

15

### 16 **3.2 Patterns in $\Delta^{14}\text{C}$ , turnover and C inputs**

17 The four soil samples from 1988 were enriched in  $^{14}\text{C}$  relative to the 2012 samples, reflecting a  
18 dominance of C with decadal turnover in the 0 – 10 cm depth increment (Fig. 1a and b, Table 3,  
19 Supplemental Table 2). Figure 1 shows temporal trends in modeled  $\Delta^{14}\text{C}$  for three representative  
20 2012 samples. Radiocarbon content in the 2012 mineral-associated C fractions exceeded the  
21 2012 atmosphere for most samples (25 of 30; Fig. 4, Supplemental Table 2), reflecting the  
22 dominance of bomb C inputs over the preceding decades. Radiocarbon content tended to be  
23 greatest in valley 10 – 20 cm and lowest in slope 10 – 20 cm soils, but differences were not  
24 significant across catena positions and depths, with the exception of slope and valley 10 – 20 cm  
25 samples (Table 3). Five slope and ridge samples had lower  $\Delta^{14}\text{C}$  than the 2012 atmosphere (< 30  
26 ‰), indicating dominance of slower cycling C pools (centennial – millennial; Fig. 1c). Figure 1  
27 shows that in samples dominated by decadal-cycling C, larger  $\Delta^{14}\text{C}$  values in 2012 correspond  
28 with longer turnover times, whereas for samples dominated by centennial-cycling C, lower  $\Delta^{14}\text{C}$   
29 values in 2012 imply longer turnover times.

steven hall 3/15/2015 6:59 PM

**Deleted:** ) where turnover could not be constrained using the bomb  $^{14}\text{C}$ -enriched surface soil samples from 1988 (Fig. 1c).

1 Mean modeled turnover times of the slow pool of mineral-associated C varied between 11 and  
2 26 y, with an overall mean and SE of  $18 \pm 3$  y ( $n = 25$ ) among samples where  $\Delta^{14}\text{C}$  exceeded the  
3 2012 atmosphere (Fig. 1; Supplemental Table 2). Omitting a three-year lag between plant C  
4 fixation and inputs to the mineral-associated pools subtly increased the mean modeled slow pool  
5 turnover time to  $20 \pm 2$  y. Varying turnover times of the passive mineral-associated C pool and  
6 sample  $\Delta^{14}\text{C}$  in 1988 had relatively little impact on slow pool turnover times and the amount of  
7 slow pool C. The combined impact of varying passive C turnover and 1988  $\Delta^{14}\text{C}$  over 1000  
8 model runs generated standard deviations in slow pool turnover times between 2 and 4 y (mean  
9  $2.8 \pm 0.1$ ) for individual samples (Supplemental Table 1). Slow pool turnover times and the  
10 amount of slow pool C did not differ significantly among topographic positions or depth,  
11 although the modeled fraction of slow pool C was smallest in the slope 10 – 20 cm soils,  
12 corresponding with lowest sample  $\Delta^{14}\text{C}$  (Table 3).

steven hall 3/17/2015 7:03 PM

Deleted: 9...and 269...y, with an ove ... [6]

13 Slow pool C greatly exceeded passive C in most (25 of 30) samples, whereas passive C was  
14 dominant in the other five samples (Supplemental Table 1). Across all 30 samples, slow pool C  
15 comprised a mean of  $66 \pm 2$  % of the mineral-associated fraction (Table 3). Omitting a three-year  
16 lag between C fixation and input increased the percentage of slow pool C to  $71 \pm 3$  % of the  
17 mineral-associated fraction. Variation in other model parameters had relatively little impact on  
18 the size of the mineral-associated slow C pool; standard deviations of the percentage of slow  
19 pool C varied between 3 and 10 % for individual soil samples (mean  $5.0 \pm 0.2$  %). However,  
20 inputs of C to the slow pool required to maintain steady-state C stocks in each 10 cm depth  
21 increment varied more than five-fold among samples (between 27 and 126 g C m<sup>2</sup> y<sup>-1</sup>) with an  
22 overall mean and SE of  $80 \pm 5$  g C m<sup>2</sup> y<sup>-1</sup> (Table 3; Supplemental Table 1). When summed over  
23 both depths, slow pool C inputs tended to be greatest on ridges, intermediate in valleys, and  
24 lowest on slopes, measuring  $183 \pm 18$ ,  $166 \pm 19$ ,  $132 \pm 17$  g C m<sup>2</sup> y<sup>-1</sup>; these differences were not  
25 statistically significant.

steven hall 3/15/2015 6:59 PM

Deleted: accounting for  $74 \pm 4$ ...here ... [7]

26 The slow pool of our constrained two-pool model and the one-pool model of mineral-associated  
27 C turnover implied contradictory relationships (of opposite sign) between C turnover times and  
28  $\Delta^{14}\text{C}$  (Fig. 3). Slow pool turnover time increased with  $\Delta^{14}\text{C}$ , whereas the overall turnover times of  
29 a single pool model decreased with  $\Delta^{14}\text{C}$ .

steven hall 3/18/2015 9:11 AM

Deleted: simple ...ne-pool model of r ... [8]

1    **3.3 Statistical models of C concentrations, stocks, and turnover**

2    Mineral-associated C concentrations (mass basis) decreased with depth, increased with Al<sub>ca</sub>,  
3    Fe(II), and Al<sub>ox</sub> concentrations, and showed no relationship with Δ<sup>14</sup>C or modeled slow pool  
4    turnover times (Table 4). Excluding depth as a potential variable in model selection yielded a  
5    model with similar explanatory power that included live fine root biomass. Both models  
6    explained the majority of spatial variation in mineral-associated C concentrations, with R<sup>2</sup> values  
7    of 0.88 and 0.73, respectively. Two similar models of mineral-associated C stocks included a)  
8    Δ<sup>14</sup>C and Al<sub>ox</sub>, or b) Fe(II) and depth, and explained less variation than models of C  
9    concentrations (R<sup>2</sup> = 0.46 and 0.43 respectively, Table 4). The optimal model for Δ<sup>14</sup>C and the  
10   slow pool turnover time of the mineral-associated C fraction included only one variable: Δ<sup>14</sup>C  
11   increased with log-transformed Fe(II) concentrations (R<sup>2</sup> = 0.35, p < 0.001; Table 4, Fig. 4), as  
12   did slow pool turnover times (R<sup>2</sup> = 0.25, p < 0.05; Table 4).

13

14    **4 Discussion**

15    We examined relationships between soil C content, turnover of mineral-associated C, and a suite  
16   of soil biogeochemical variables thought to affect C storage across topographic zones  
17   representing three different soil orders characteristic of humid tropical forests. Samples differed  
18   greatly in their concentrations of reactive Fe and Al, reducing conditions (as measured by Fe(II)  
19   concentrations), and live fine root biomass—drivers that have been proposed to control soil C  
20   dynamics within and among ecosystems (Torn et al., 1997; Schuur et al., 2001; Powers and  
21   Schlesinger, 2002; Kleber et al., 2005). While we found strong relationships between several of  
22   these biogeochemical indices and mineral-associated C concentrations, they explained less  
23   variation in mineral-associated C stocks and turnover, which were surprisingly consistent across  
24   these disparate soils. Thus, despite the demonstrable differences in biogeochemical  
25   characteristics among topographic positions, our results implied that these soils received similar  
26   annual inputs of C to a slow-cycling mineral-associated C pool.

27

steven hall 3/17/2015 7:46 PM  
Deleted: s

1    **4.1 Patterns of mineral-associated C concentrations and stocks**

2    The mineral-associated “heavy” fraction comprised the vast majority of soil C in this ecosystem,  
3    similar to findings from other humid tropical forests (Trumbore, 1993; Marin-Spiotta et al.,  
4    2009; Cusack et al., 2011), and contrasting with many temperate forests where low-density C  
5    fractions are often significant in mineral soils (von Lützow et al., 2007). We could explain a  
6    large majority of the variation in mineral-associated C concentrations across the catena with a  
7    small suite of biogeochemical drivers: concentrations of reactive Al, Fe(II), and depth, a factor  
8    that could largely be explained by live fine root biomass. These models of C concentrations are  
9    consistent with previous work documenting strong and widespread relationships between  
10   reactive metals, short range-order minerals and C content across disparate soils (e.g. Baldock and  
11   Skjemstad, 2000; Kleber et al., 2005; Kramer et al., 2012). However, models also reflected the  
12   likely importance of reducing microsites, as indicated by Fe(II). Reduced iron accumulation is  
13   reflective of O<sub>2</sub> limitation, which can decrease decomposition rates by inhibiting oxidative  
14   enzymes and decreasing the ATP yield of respiration. Iron reduction can also release DOC into  
15   the soil solution and increase its bioavailability (Thompson et al., 2006), although the positive  
16   relationship shown here between Fe(II) and mineral-associated C content and turnover times  
17   suggests that any inhibitory effects of anaerobiosis on organic matter decomposition may  
18   predominate in these soils over decadal timescales. Reduced Fe is potentially vulnerable to  
19   leaching, although O<sub>2</sub> heterogeneity promotes Fe oxidation and precipitation in close proximity  
20   to anaerobic microsites (Hall et al., 2013), leading to the maintenance of large pools of short  
21   range-order Fe in this system as indicated by the soil extraction data. Depth was another effective  
22   surrogate for mineral-associated C, which could largely be explained by increased live fine root  
23   biomass in 0 – 10 cm samples. Substituting live fine roots for depth only slightly decreased  
24   model predictive power, suggesting the influence of root C inputs or other covariate(s) related to  
25   live fine root biomass in increasing mineral-associated C content. Standardized regression  
26   coefficients indicated that each of these variables (reactive Al, Fe(II), and live fine roots) was  
27   similarly important in explaining spatial variation in C concentrations. Clay content showed no  
28   relationship with soil C, suggesting that hydrous oxides and metals were more important in  
29   stabilizing C than silicates in these soils, which are dominated by relatively less-reactive  
30   kaolinite.

1 The measured biogeochemical variables were less effective in explaining patterns of surface soil  
2 C stocks than C concentrations among these samples. This may have been partially due to a  
3 strong inverse relationship between soil mass-based C concentrations and bulk density that has  
4 been widely documented in other soils (Saini, 1966). Translocation of surface C to deeper (and  
5 denser) un-sampled horizons could also contribute to this discrepancy.

6

7 **4.2 Patterns of mineral-associated C turnover**

8 **4.2.1 Comparisons with previous studies in tropical forests**

9 Despite the importance of tropical forest soils to the global C cycle, few studies have rigorously  
10 constrained the turnover rates of organic matter pools associated with mineral surfaces, which  
11 represent the bulk of soil C in these ecosystems. Numerous studies have exploited tropical land-  
12 use conversions characterized by shifts from C3 to C4 vegetation and concomitant changes in C  
13 isotopes to examine C turnover, yet these investigations cannot separate effects of disturbance  
14 and species change from background C dynamics (Ehleringer et al., 2000). Our results  
15 demonstrated consistently rapid turnover of the mineral-associated slow C pool (11 – 26 y, mean  
16 and SE 18 ± 3 y), which represented a majority (66 ± 2 %) of mineral-associated C in most  
17 samples. These estimates are remarkably consistent with <sup>14</sup>C-derived turnover estimates by  
18 Trumbore et al. (1995) from Oxisols in seasonally dry Amazonian forests, where the slow pool  
19 of dense organic matter turned over on timescales of 10 – 30 y. Our estimates of slow pool C  
20 turnover are also similar to several Hawaiian ecosystems (Townsend et al., 1995; Torn et al.,  
21 2005), but faster than those modeled in an Amazonian Oxisol in Manaus, where most C (70%)  
22 turned over on timescales of 70 y at 10 cm depth (Telles et al., 2003).

23

24 **4.2.2 Trends in C turnover with depth**

25 Notably, we generally found similar  $\Delta^{14}\text{C}$  and turnover rates in surface (0 – 10 cm) and  
26 subsurface (10 – 20 cm) depths (corresponding with A and B horizons, respectively) across all  
27 catena positions. This finding contrasts with steep declines in decomposition rates frequently  
28 observed in relatively shallow B horizons in other ecosystems (Trumbore et al., 1995; Gaudinski  
29 et al., 2000; Telles et al., 2003; Torn et al., 2013). Inputs of surface litter and root biomass are

steven hall 3/17/2015 7:48 PM  
**Deleted:** 9  
steven hall 3/17/2015 7:48 PM  
**Deleted:** 9  
steven hall 3/17/2015 7:48 PM  
**Deleted:** large

1 typically thought to be lower in B horizons, but C redistribution by anecic earthworms and other  
2 soil fauna likely provides an important source of C inputs to the subsoil, as these organisms are  
3 extremely abundant at this site and in many other humid tropical forests (Gonzalez et al., 2006).  
4 Dissolved organic C (DOC) represents another likely source of C to subsurface soils, and  
5 concentrations are very low (~2 mg l<sup>-1</sup>) at A/B horizon boundaries at this site (Hall et al., 2013),  
6 indicative of high sorption capacity. ~~Five of the 30 samples had Δ<sup>14</sup>C lower than the 2012~~  
7 atmosphere, and also tended to have lower C concentrations, likely reflecting “dilution” of the  
8 slow pool by passive C with pre-modern Δ<sup>14</sup>C values. These samples may reflect recent fine-  
9 scale redistribution of surface soil and exposure of deeper soil horizons following disturbances  
10 such as small treefall gaps (scale of m<sup>2</sup>), which are common in this forest (Scatena and Lugo,  
11 1995). Overall, our findings of high Δ<sup>14</sup>C and relatively fast turnover across the two depths  
12 sampled here suggests that both surface and shallow subsurface mineral-associated organic  
13 matter may respond more rapidly to environmental change than was previously thought (de  
14 Camargo et al., 1999).

15

#### 16 **4.2.3 Impacts of biogeochemical drivers on C turnover**

17 The overall similarities in turnover times across topographic positions despite large differences  
18 in concentrations of reactive Al, Fe, and clay content suggest that spatial variation in mineral-  
19 organic interactions may have less impact on C turnover rates in these highly weathered soils  
20 (Ultisols, Oxisols, and Inceptisols) than in allophane-rich Andisols, where short range-order  
21 minerals were a dominant predictor of Δ<sup>14</sup>C (Torn et al., 1997). Herold et al. (2014) similarly  
22 found a positive relationship between reactive metals and C content, but not Δ<sup>14</sup>C, in German  
23 Luvisols and Stagnosols. Thus, strong positive relationships between soil C concentrations and  
24 the reactive Al content measured in soil extractions may reflect the importance of metal-organic  
25 associations in transient C accumulation rather than long-term C stabilization in this ecosystem.  
26 Concentrations of Fe(II) indicative of microbial Fe reduction were the only variable significantly  
27 correlated with Δ<sup>14</sup>C and turnover times. Although Fe(II) concentrations vary over time in this  
28 ecosystem in response to microsite-scale biogeochemical processes, nearby plots tended to  
29 maintain differences and rank order in reducing conditions (Hall et al., 2013), likely as a function  
30 of microtopography (ie concave vs. convex surfaces over scales of tens of cm). Thus, the nearly

steven hall 3/17/2015 1:34 PM

**Deleted:** Intriguingly,

steven hall 3/17/2015 1:34 PM

**Deleted:** f

1 four orders of magnitude of variation in Fe(II) concentrations that we observed among plots  
2 likely reflects, to at least some extent, constitutive differences in reducing conditions. The  
3 positive relationship between Fe(II) and  $\Delta^{14}\text{C}$  may suggest that microsite O<sub>2</sub> limitation affects  
4 slow-pool C turnover, possibly due to the inhibition of oxidative enzymes and/or decreased  
5 energy yield of decomposition, as discussed above. The trend towards greater  $\Delta^{14}\text{C}$  (implying  
6 longer turnover) in 10 – 20 cm horizons of the valley soils, which experience the lowest bulk soil  
7 O<sub>2</sub> availability, is also supportive of this hypothesis. We found no significant differences in  
8 mineral-associated C turnover and stocks across ridge, slope, and valley positions despite  
9 variation in bulk soil O<sub>2</sub> measured previously (Silver et al., 1999). Differences between bulk soil  
10 O<sub>2</sub> and Fe(II) concentrations are by no means surprising, as they reflect different spatial scales of  
11 redox heterogeneity (macropore O<sub>2</sub> content vs. soil microsites) that have long been known to  
12 exist in soils (Sexstone et al., 1985). The pattern of significantly greater occluded light C content  
13 in valley soils, however, may imply that consistently lower bulk soil O<sub>2</sub> availability might  
14 promote the accumulation of this fraction. Berhe et al. (2012) found a similar pattern of light-  
15 fraction C accumulation in a poorly-drained valley in a Mediterranean shrub/grassland.

16

#### 17 **4.2.4 Steady-state mineral-associated C inputs**

18 Another implication of the relatively rapid turnover of the slow cycling C pool and the uniform C  
19 stocks across the catena is the substantial C input to the mineral-associated fraction required to  
20 maintain steady state. Aboveground litterfall NPP averages approximately 900 g biomass m<sup>-2</sup> y<sup>-1</sup>  
21 in this forest (Scatena et al., 1996), implying that root NPP is likely of similar magnitude (Malhi  
22 et al., 2011). Assuming a C concentration of 50 % in above- and belowground litter inputs, this  
23 implies that roughly 900 g C m<sup>-2</sup> y<sup>-1</sup> are delivered to the soil via leaf and root litter. Modeled  
24 turnover times of the mineral-associated slow pool imply C inputs of 76–228 g C m<sup>-2</sup> y<sup>-1</sup> (mean  
25 = 160 g C m<sup>-2</sup> y<sup>-1</sup>) from 0 – 20 cm in each plot. Inputs of C to the mineral-associated slow pool  
26 thus represent a substantial C flux of approximately 10 – 25% of annual litter inputs in this  
27 forest. These are conservative estimates given that inputs to the mineral-associated slow fraction  
28 C below 20 cm depth are also likely to be important (Koarashi et al., 2012), especially in ridges  
29 and valleys where the 0 – 10 and 10 – 20 cm horizons did not differ in  $\Delta^{14}\text{C}$  or C stocks.

30

steven hall 3/18/2015 9:19 AM

**Deleted:** 8

steven hall 3/18/2015 9:19 AM

**Deleted:** 49

steven hall 3/18/2015 9:19 AM

**Deleted:** 5

steven hall 3/18/2015 9:20 AM

**Deleted:** 30

steven hall 3/18/2015 9:21 AM

**Deleted:** likely

steven hall 3/18/2015 9:21 AM

**Deleted:** may also be

1    **4.2.5 Erosion / deposition impacts on  $\Delta^{14}\text{C}$**

2    Erosion and deposition also represent potentially important fluxes of C in these soils over  
3    pedogenic timescales given their steep slopes and high precipitation, but erosion is unlikely to  
4    have an overriding impact on interpretations of decadal C turnover times in our sites. Other  
5    studies have shown  $\Delta^{14}\text{C}$  enrichment in toeslope positions corresponding with recent C inputs  
6    from upslope erosion (Berhe et al., 2012; Berhe and Kleber, 2013). In these studies, relatively  
7    small inputs of modern C had a large impact on soil  $\Delta^{14}\text{C}$  and inferred turnover times due to the  
8    predominance of pre-modern  $^{14}\text{C}$  in these soils. In our study ecosystem, however, rapid cycling  
9    of the mineral-associated slow C pool led to significant bomb  $^{14}\text{C}$  enrichment in all samples, ▶  
10   regardless of year sampled or topographic position. In this context, realistic inputs of C from  
11   erosion would have negligible impact on mineral-associated  $\Delta^{14}\text{C}$ , given that erosive transport in  
12   this site is relatively minor in comparison with biological C fluxes over annual to decadal  
13   timescales. Lateral surface fluxes of fine litter and soil on slopes averaged  $5 \pm 4$  and  $9 \pm 6 \text{ g m}^{-2}$   
14    $\text{y}^{-1}$  (masses of litter and soil, respectively) in sites located near our plots (Larsen et al., 1999).  
15   Even if erosion were ten-fold greater at our site, the C fluxes would be negligible in comparison  
16   with litter inputs to the mineral-associated slow pool. We acknowledge that large episodic  
17   landslides impacting  $> 100 \text{ m}^2$  can occur in this forest (Scatena and Lugo, 1995), although there  
18   is no evidence of major landslides at this site over the preceding decades. The low clay content  
19   of the riparian valley soils relative to ridges and slopes suggests that clay removal by irregular  
20   flood events exceeds clay deposition over pedogenic timescales. Thus, large-scale geomorphic  
21   processes such as landslides and floods shape this forest landscape over scales of centuries to  
22   millennia, but their impact on C dynamics of the mineral-associated slow pool has likely been  
23   minor at this particular site over at least the last several decades.

steven hall 3/16/2015 12:16 PM

Deleted: ,

24   Nevertheless, fine-scale variation in geomorphic processes such as soil creep or tree tip-up  
25   mounds may be important in explaining variation in soil  $\Delta^{14}\text{C}$  among samples, particularly for  
26   the five samples with relatively lower  $\Delta^{14}\text{C}$  values. Relatively fewer  $^{14}\text{C}$  studies have been  
27   conducted along hillslope catenas, which are likely to show greater spatial variability than soils  
28   formed on level terrain.

29

1    **4.3 Implications for models and global trends in mineral-associated C turnover**

2    The presence of archived soil samples was critical for constraining our models of mineral-  
3    associated C turnover, yet relatively few historical datasets have been analyzed for  $^{14}\text{C}$ . We are  
4    aware of only one such study conducted in humid tropical soils (Telles et al., 2003), although  
5    this approach has been applied more frequently in temperate soils (Trumbore et al., 1996;  
6    [Baisden et al., 2002](#); Koarashi et al., 2012; [Sierra et al., 2012](#); Baisden et al., 2013; [Schrumpf and](#)  
7    [Kaiser 2015](#)). Radiocarbon analysis of archived 1988 samples revealed that we could not  
8    usefully model the mineral-associated C fraction as a uniform pool with a single turnover time.  
9    Although the heterogeneous nature of mineral-associated C and the implications for modeling  
10   were noted almost two decades ago (Trumbore et al., 1995), many recent studies modeled the  
11   turnover of physically-separated mineral-associated C fractions as uniform pools (Crow et al.,  
12   2007; Marin-Spiotta et al., 2008; Leifeld and Fuhrer, 2009; Meyer et al., 2012; McFarlane et al.,  
13   2013; Herold et al., 2014). Other studies have further separated mineral-associated C by  
14   additional density fractions or chemical treatment (Telles et al., 2003; Sollins et al., 2009;  
15   Giardina et al., 2014), but these fractions also likely represent mixtures of faster and slower-  
16   cycling C pools that complicate interpretation of a single mean residence time. For example,  
17   [Telles et al. \(2003\) found significant bomb  \$^{14}\text{C}\$  in acid hydrolysis residue of a dense soil fraction.](#)  
18   This finding illustrates that even C that is highly stable to chemical degradation can have a  
19   significant fast-cycling component, and thus cannot be usefully modeled as a uniform pool.  
20   Using multiple  $^{14}\text{C}$  measurements over time to mathematically model multiple C pools within a  
21   single density fraction provides a useful alternative to physically separating each pool of interest.  
22   Using an 11-point time series from a New Zealand grassland soil, Baisden et al. (2013) showed  
23   that a two-pool model could closely approximate bulk soil  $^{14}\text{C}$  dynamics. Additional work  
24   showed that including a third pool of rapidly-cycling (months – years) C was also useful  
25   (Baisden and Keller 2013). However, in the present study we did not have access to additional  
26   data necessary to parameterize a third pool. Rather, to minimize the impact of very rapidly-  
27   cycling C we removed particulate organic matter by density fractionation (Baisden and Canessa  
28   2013). Particulate organic matter typically decomposes over timescales of months in this  
29   ecosystem (Cusack et al. 2009), and thus has a significantly different isotope composition from  
30   mineral-associated C which would affect bulk soil  $\Delta^{14}\text{C}$  values even at low abundance. It is likely

steven hall 3/17/2015 12:14 PM

**Deleted:** soil  $^{14}\text{C}$  turnover modeling

steven hall 3/16/2015 12:43 PM

**Formatted:** Normal, Line spacing: single

steven hall 3/18/2015 9:24 AM

**Deleted:** have continued to model

steven hall 3/16/2015 11:23 AM

**Deleted:** it appears likely that

steven hall 3/16/2015 11:23 AM

**Deleted:** contain

steven hall 3/16/2015 7:33 PM

**Deleted:** the

steven hall 3/16/2015 7:33 PM

**Deleted:** s

steven hall 3/16/2015 7:33 PM

**Deleted:** provided in these studies.

1 that mineral-associated C also contains a portion of very rapidly-cycling C, although this pool is  
2 inherently difficult to estimate with  $^{14}\text{C}$  measurements (Baisden and Keller 2013) and merits  
3 future study using other methods such as stable isotope labeling.

4 Our results confirm that single-pool turnover models of mineral-associated C can lead to  
5 misleading interpretations of turnover times and their relationship to  $\Delta^{14}\text{C}$ , as recently  
6 summarized by Baisden and Canessa (2013). In our study, a one-pool model could not capture  
7 observed changes in  $\Delta^{14}\text{C}$  between 1988 and 2012, and one and two-pool models of mineral-  
8 associated C yielded contradictory relationships between  $\Delta^{14}\text{C}$  and modeled mean residence  
9 times. Importantly, the one pool model reversed the sign of the relationship between  $\Delta^{14}\text{C}$  and  
10 turnover time, because a slow pool with decadal turnover times represented the majority of  
11 mineral-associated C in the two-pool model. Thus, relationships between C turnover and  
12 environmental drivers assessed using one-pool models of mineral-associated C can be  
13 qualitatively incorrect. Furthermore, the centennial turnover times of mineral-associated organic  
14 matter implied by many one-pool models (Leifeld and Fuhrer, 2009; Meyer et al., 2012;  
15 McFarlane et al., 2013; Herold et al., 2014) obscure the finding from two-pool models that most  
16 mineral-associated C appears to cycle over decadal scales.

17 Accumulating evidence from studies employing paired  $^{14}\text{C}$  analysis across a broad range of  
18 ecosystems and soil types suggests that most mineral-associated C in surface horizons cycles  
19 over scales of years to several decades and thus may respond more rapidly to ecosystem  
20 perturbations than previously thought. For example, residence times of a slow pool comprising >  
21 70 % of mineral-associated C varied between 10 and 40 years among samples spanning a 3  
22 million-year grassland chronosequence in California (Baisden et al., 2002). Similarly, 78 – 85 %  
23 of total soil C (including a large portion of mineral-associated C) cycled over timescales of 9 –  
24 17 years across New Zealand grassland soils varying in mineralogy (Baisden et al., 2013).  
25 Hydrolyzable C comprising the majority of mineral-associated C also cycled over decadal scales  
26 in surface soils across an elevation gradient in the Sierra Nevada Mountains of California  
27 (Trumbore et al., 2006). Subsoil mineral-associated organic matter in this region showed similar  
28 dynamics, where 28 – 73 % of C had turnover times of 10 – 95 years (Koarashi et al., 2012). The  
29 slow pool turnover times and abundances modeled in our study fall within this range, as do

steven hall 3/16/2015 11:25 AM

**Deleted:** For example

steven hall 3/16/2015 11:25 AM

**Deleted:** we found that modeling the

steven hall 3/16/2015 11:25 AM

**Deleted:** fraction as a pool with uniform turnover time vs. slow and passive pools with turnover times of decades to centuries – millennia, respectively,

steven hall 3/16/2015 11:27 AM

**Deleted:** .

steven hall 3/16/2015 11:28 AM

**Deleted:** The two-pool model was more ecologically meaningful and corresponded well with previous  $^{14}\text{C}$  soil C turnover models constrained by multiple data points (Baisden et al., 2013), whereas the single pool model could not capture observed changes in  $\Delta^{14}\text{C}$  between 1988 and 2012.

steven hall 3/16/2015 7:48 PM

**Deleted:** single

steven hall 3/16/2015 11:28 AM

**Formatted:** Font:Not Italic

steven hall 3/16/2015 7:49 PM

**Deleted:** the

steven hall 3/16/2015 7:49 PM

**Deleted:** constituted

steven hall 3/17/2015 11:03 AM

**Moved (insertion) [3]**

steven hall 3/17/2015 11:08 AM

**Deleted:** For example, single-pool models of dense fraction  $^{14}\text{C}$  in temperate soils often imply mean turnover times of hundreds of years (Leifeld and Fuhrer, 2009; Meyer et al., 2012; McFarlane et al., 2013; Herold et al., 2014).

1 reported values from other humid tropical forests (Trumbore et al., 1995). In summary, these  
2 studies suggest that most mineral-associated C across a broad range ecosystems cycles over  
3 decadal timescales, with little apparent relationship to differences in climate. Additional  
4 geographically distributed measurements of soil  $^{14}\text{C}$  over time could provide insight into the  
5 impacts of ecosystem state factors on mineral-associated C turnover at a global scale.▼  
6

## 7 5 Conclusions

8 Overall, we found that dynamics of mineral-associated soil C pools were remarkably similar  
9 across biogeochemically distinct soils spanning three taxonomic orders and two surface soil  
10 depths in a humid tropical forest, with 66% of this C cycling over timescales of approximately  
11 18 years. Our results highlight the importance of dynamics of the mineral-associated slow pool  
12 in mediating surface soil responses to global change. Our data indicate that soil depths of at least  
13 20 cm can be dominated by mineral-association C with decadal rather than centennial/millennial  
14 turnover times, and that large differences in soil biogeochemical properties such as texture, pH,  
15 reactive metal content, root biomass, and bulk soil O<sub>2</sub> do not necessarily have discernable  
16 impacts on decadal turnover rates. However, an index of reducing conditions (Fe(II)  
17 concentrations) at the scale of individual soil samples provided the best single predictor of  $\Delta^{14}\text{C}$ ,  
18 slow-pool turnover, and C stocks, suggesting the influence of microsite redox conditions on C  
19 dynamics in these upland soils.  
20

## 21 Acknowledgements

22 All data summarized in this manuscript is available in the associated Supplemental Material. We  
23 thank H. Dang, J. Treffkorn, J. Cosgrove, R. Ryals, O. Gutierrez, A. McDowell, B. Ryals, and C.  
24 Torrens for crucial help in the field and lab. M. Firestone, R. Rhew, and the reviewers T. Baisden  
25 and J. Sanderman provided extremely valuable comments and discussion. SJH was funded by  
26 the DOE Office of Science Graduate Fellowship Program supported by the American Recovery  
27 and Reinvestment Act of 2009, administered by ORISE-ORAU under contract no. DE-AC05-  
28 06OR23100. Funding was also provided by NSF grant EAR-08199072 to WLS, the NSF  
29 Luquillo Critical Zone Observatory (EAR-0722476) with additional support provided by the

steven hall 3/17/2015 11:55 AM

**Deleted:** In a single pool model, Equation (1) yielded turnover times of either 67 – 150 years or 6 – 15 years. Mass balance constraints on C inputs imply that only the longer turnover times are plausible, a conclusion supported by weighted mean turnover times from the two-pool model. However, the single-pool model would imply a negative relationship between  $\Delta^{14}\text{C}$  and overall turnover time. This interpretation contradicts the constrained two-pool model, which showed a positive relationship between soil  $\Delta^{14}\text{C}$  and turnover time of the slow fraction, the dominant C pool, despite the fact that these two models had similar overall turnover times. Phrased differently, for a single-pool model we would expect shorter overall turnover times to correspond with greater bomb  $^{14}\text{C}$  in 2012, whereas for a two-pool model, shorter turnover times for the slow pool correspond with less bomb  $^{14}\text{C}$  in 2012. Thus, single pool turnover models based on mineral-associated  $\Delta^{14}\text{C}$  can lead to misleading interpretations of turnover times and their relationship to  $\Delta^{14}\text{C}$ .

steven hall 3/16/2015 7:39 PM

**Deleted:** For example, Leifeld and Fuhrer (2009) argued that differences in the  $\Delta^{14}\text{C}$  of silt and clay-associated organic matter among two alpine grasslands could not be attributed to differences in management over decadal scales because of the long mean turnover times of these fractions (200 – 800) implied by single-pool models. Acknowledging the presen... [9]

steven hall 3/17/2015 11:03 AM

**Moved up [3]:** For example, single-pool models of dense fraction  $\Delta^{14}\text{C}$  in temperate soils often imply mean turnover times of hundreds of years (Leifeld and Fuhrer ... [10])

steven hall 3/16/2015 7:35 PM

**Moved up [2]:** For example, Leifeld and Fuhrer (2009) argued that differences in the  $\Delta^{14}\text{C}$  of silt and clay-associated organic matter among two alpine grasslands could no ... [11]

steven hall 3/16/2015 7:35 PM

**Moved (insertion) [2]**

steven hall 3/18/2015 10:18 AM

**Deleted:** 74

steven hall 3/18/2015 10:19 AM

**Deleted:** decomposing

steven hall 3/18/2015 10:19 AM

**Deleted:** after

steven hall 3/18/2015 10:19 AM

**Deleted:** 20

steven hall 4/4/2015 6:54 PM

**Deleted:** several anonymous reviewers

1 USGS Luquillo WEBB program, and grant DEB 0620910 from NSF to the Institute for Tropical  
2 Ecosystem Studies, University of Puerto Rico, and to the International Institute of Tropical  
3 Forestry USDA Forest Service, as part of the Luquillo Long-Term Ecological Research Program.  
4 This work was supported by the USDA National Institute of Food and Agriculture, McIntire  
5 Stennis project CA-B-ECO-7673-MS.

Whendee Silver 3/31/2015 7:56 PM  
**Formatted:** Font:Not Bold

6  
7

## 8 References

- 9 Baisden, W. T., Amundson, R., Cook, A. C. and Brenner, D. L.: Turnover and storage of C and  
10 N in five density fractions from California annual grassland surface soils, Global  
11 Biogeochemical Cycles, 16, doi:10.1029/2001GB001822, 2002.
- 12 Baisden, W. T. and Canessa, S.: Using 50. and Brenner, D. L.: Turnover and storage of C and N  
13 in five density fractions from California annual grassland surface soils, Glosmt Act of 2009,  
14 administered by ORISE-ORAU under ons with Materials and Atoms, 294, 588–592, 2013.
- 15 Baisden, W. T. and Keller, E. D.: Synthetic constraint of soil C dynamics using 50 years of <sup>14</sup>C  
16 and net primary production (NPP) in a New Zealand grassland site, Radiocarbon, 55, 1071 of  
17 approach.
- 18 Baisden, W. T., Parfitt, R. L., Ross, C., Schipper, L. A. and Canessa, S.: Evaluating 50 years of  
19 time-series soil radiocarbon data: towards routine calculation of robust C residence times,  
20 Biogeochemistry, 112, 129–137, 2013.
- 21 Baldock, J. A. and Skjemstad, J. O.: Role of the soil matrix and minerals in protecting natural  
22 organic materials against biological attack, Org. Geochem., 31, 697–710, 2000.
- 23 Berhe, A. A., Harden, J. W., Torn, M. S., Kleber, M., Burton, S. D. and Harte, J.: Persistence of  
24 soil organic matter in eroding versus depositional landform positions, J. Geophys. Res., 117,  
25 doi:10.1029/2011JG001790, 2012.
- 26 Berhe, A. A. and Kleber, M.: Erosion, deposition, and the persistence of soil organic matter:  
27 mechanistic considerations and problems with terminology, Earth Surf. Process. Landforms, 38,  
28 908–912, 2013.
- 29 de Camargo, P. B., Trumbore, S. E., Martinelli, L. A., Davidson, E. A., Nepstad, D. C. and  
30 Victoria, R. L.: Soil carbon dynamics in regrowing forest of eastern Amazonia, Global Change  
31 Biol., 5, 693–702, 1999.
- 32 Crow, S. E., Swanston, C. W., Lajtha, K., Brooks, J. R. and Keirstead, H.: Density fractionation  
33 of forest soils: methodological questions and interpretation of incubation results and turnover  
34 time in an ecosystem context, Biogeochemistry, 85, 69–90, 2007.

steven hall 4/1/2015 2:11 PM  
**Formatted:** Font:Times

steven hall 3/18/2015 11:06 AM  
**Formatted:** Normal

steven hall 4/1/2015 2:11 PM  
**Formatted:** Font:Times

steven hall 4/1/2015 2:11 PM  
**Formatted:** Font:Times

steven hall 4/1/2015 2:11 PM  
**Formatted:** Font color: Auto

- 1 Cusack, D. F., Chou, W. W., Yang, W. H., Harmon, M. E. and Silver, W. L.: Controls on long-  
2 term root and leaf litter decomposition in neotropical forests, *Global Change Biol.*, 15, 1339–  
3 1355, 2009.
- 4 Cusack, D. F., Silver, W. L., Torn, M. S. and McDowell, W. H.: Effects of nitrogen additions on  
5 above- and belowground carbon dynamics in two tropical forests, *Biogeochemistry*, 104, 203–  
6 225, 2011.
- 7 Dubinsky, E. A., Silver, W. L. and Firestone, M. K.: Tropical forest soil microbial communities  
8 couple iron and carbon biogeochemistry, *Ecology*, 91, 2604–2612, 2010.
- 9 Ehleringer, J. R., Buchmann, N. and Flanagan, L. B.: Carbon isotope ratios in belowground  
10 carbon cycle processes, *Ecol. Appl.*, 10, 412–422, 2000.
- 11 Feller, C. and Beare, M. H.: Physical control of soil organic matter dynamics in the tropics,  
12 *Geoderma*, 79, 69–116, 1997.
- 13 Gaudinski, J. B., Trumbore, S. E., Davidson, E. A. and Zheng, S.: Soil carbon cycling in a  
14 temperate forest: radiocarbon-based estimates of residence times, sequestration rates and  
15 partitioning of fluxes, *Biogeochemistry*, 51, 33–69, 2000.
- 16 Gee, G. and Bauder, J.: Particle size analysis, in *Methods of Soil Analysis, Part 1, Physical and  
17 Mineralogical Methods*, edited by A. Klute, pp. 383–411, American Society of Agronomy,  
18 Madison, WI, USA., 1986.
- 19 Giardina, C. P., Litton, C. M., Crow, S. E. and Asner, G. P.: Warming-related increases in soil  
20 CO<sub>2</sub> efflux are explained by increased below-ground carbon flux, *Nature Climate Change*, 4,  
21 822–827, 2014.
- 22 Gonzalez, G., Huang, C. Y., Zou, X. and Rodríguez, C.: Earthworm invasions in the tropics,  
23 *Biol. Invasions*, 8, 1247–1256, 2006.
- 24 Hall, S. J., McDowell, W. H. and Silver, W. L.: When wet gets wetter: Decoupling of moisture,  
25 redox biogeochemistry, and greenhouse gas fluxes in a humid tropical forest soil, *Ecosystems*,  
26 16, 576–589, 2013.
- 27 Heartsill Scalley, T., Scatena, F. N., Lugo, A. E., Moya, S. and Estrada Ruiz, C. R.: Changes in  
28 structure, composition, and nutrients during 15 yr of hurricane-induced succession in a  
29 subtropical wet forest in Puerto Rico, *Biotropica*, 42, 455–463, 2010.
- 30 Herold, N., Schöning, I., Michalzik, B., Trumbore, S. and Schrumpf, M.: Controls on soil carbon  
31 storage and turnover in German landscapes, *Biogeochemistry*, 119, 435–451, 2014.
- 32 Hua, Q., Barbetti, M. and Rakowski, A. Z.: Atmospheric radiocarbon for the period 1950–2010,  
33 *Radiocarbon*, 55, 2059–2072, 2013.
- 34 Jobbagy, E. G. and Jackson, R. B.: The vertical distribution of soil organic carbon and its relation  
35 to climate and vegetation, *Ecol. Appl.*, 10, 423–436, 2000.

- 1 Kleber, M., Mikutta, R., Torn, M. S. and Jahn, R.: Poorly crystalline mineral phases protect  
2 organic matter in acid subsoil horizons, *Eur. J. Soil Sci.*, 56, 717–725, 2005.
- 3 Koarashi, J., Hockaday, W. C., Masiello, C. A. and Trumbore, S. E.: Dynamics of decadally  
4 cycling carbon in subsurface soils, *J. Geophys. Res.*, 117, G03033, doi:10.1029/2012JG002034,  
5 2012.
- 6 Kramer, M. G., Sanderman, J., Chadwick, O. A., Chorover, J. and Vitousek, P. M.: Long-term  
7 carbon storage through retention of dissolved aromatic acids by reactive particles in soil, *Global  
8 Change Biol.*, 18, 2594–2605, 2012.
- 9 Larsen, M. C., Torres-Sánchez, A. J. and Concepción, I. M.: Slopewash, surface runoff and fine-  
10 litter transport in forest and landslide scars in humid-tropical steppes, *Luquillo Experimental  
11 Forest*, Puerto Rico, *Earth Surf. Process. Landforms*, 24, 481–502, 1999.
- 12 Leifeld, J. and Fuhrer, J.: Long-term management effects on soil organic matter in two cold,  
13 high-elevation grasslands: clues from fractionation and radiocarbon dating, *Eur. J. Soil Sci.*, 60,  
14 230–239, 2009.
- 15 Liptzin, D., Silver, W. L. and Detto, M.: Temporal dynamics in soil oxygen and greenhouse  
16 gases in two humid tropical forests, *Ecosystems*, 14, 171–182, 2011.
- 17 Von Lützow, M., Kögel-Knabner, I., Ekschmitt, K., Flessa, H., Guggenberger, G., Matzner, E.  
18 and Marschner, B.: SOM fractionation methods: Relevance to functional pools and to  
19 stabilization mechanisms, *Soil Biol. Biochem.*, 39, 2183–2207, 2007.
- 20 Malhi, Y., Doughty, C. and Galbraith, D.: The allocation of ecosystem net primary productivity  
21 in tropical forests, *Phil. Trans. R. Soc. B*, 366, 3225–3245, 2011.
- 22 Marin-Spiotta, E., Silver, W. L., Swanston, C. W. and Ostertag, R.: Soil organic matter dynamics  
23 during 80 years of reforestation of tropical pastures, *Global Change Biol.*, 15, 1584–1597, 2009.
- 24 Marin-Spiotta, E., Swanston, C. W., Torn, M. S., Silver, W. L. and Burton, S. D.: Chemical and  
25 mineral control of soil carbon turnover in abandoned tropical pastures, *Geoderma*, 143, 49–62,  
26 2008.
- 27 Masiello, C. A., Chadwick, O. A., Southon, J., Torn, M. S. and Harden, J. W.: Weathering  
28 controls on mechanisms of carbon storage in grassland soils, *Global Biogeochem. Cycles*, 18,  
29 GB4023, doi:10.1029/2004GB002219, 2004.
- 30 McDowell, W. H., Scatena, F. N., Waide, R. B., Brokaw, N., Camilo, G., Covich, A., Crowl, T.,  
31 Gonzalez, G., Greathouse, E., Klawinski, P., Lodge, D., Lugo, A., Pringle, C., Richardson, B.,  
32 Richardson, M., Schaefer, D., Silver, W., Thompson, J., Vogt, D., Vogt, K., Willig, M.,  
33 Woolbright, L., Zou, X. and Zimmerman, J.: Geographic and ecological setting of the Luquillo  
34 Mountains, in *A Caribbean Forest Tapestry: The Multidimensional Nature of Disturbance and  
35 Response*, edited by N. Brokaw, T. Crowl, A. Lugo, W. H. McDowell, F. N. Scatena, R. B.  
36 Waide, and M. Willig, pp. 72–163, Oxford University Press, New York, USA., 2012.

- 1 McFarlane, K. J., Torn, M. S., Hanson, P. J., Porras, R. C., Swanston, C. W., Callaham, M. A.  
2 and Guilderson, T. P.: Comparison of soil organic matter dynamics at five temperate deciduous  
3 forests with physical fractionation and radiocarbon measurements, *Biogeochemistry*, 112, 457–  
4 476, 2013.
- 5 Meyer, S., Leifeld, J., Bahn, M. and Fuhrer, J.: Free and protected soil organic carbon dynamics  
6 respond differently to abandonment of mountain grassland, *Biogeosciences*, 9, 853–865, 2012.
- 7 Phillips, E. J. P., Lovley, D. R. and Roden, E. E.: Composition of non-microbially reducible  
8 Fe(III) in aquatic sediments, *Appl. Environ. Microbiol.*, 59, 2727–2729, 1993.
- 9 Pinheiro, J., Bates, D., DebRoy, S., Sarkar, D. and R Core Development Team: nlme: Linear and  
10 Nonlinear Mixed Effects Models. [online] Available from: <http://CRAN.R-project.org/package=nlme>, (last access: 1 June 2014), 2014.
- 11
- 12 Ponnamperuma, F. N.: The chemistry of submerged soils, *Adv. Agron.*, 24, 29–96, 1972.
- 13 Powers, J. S., Montgomery, R. A., Adair, E. C., Brearley, F. Q., DeWalt, S. J., Castanho, C. T.,  
14 Chave, J., Deinert, E., Ganzhorn, J. U., Gilbert, M. E., González-Iturbe, J. A., Bunyavejchewin,  
15 S., Grau, H. R., Harms, K. E., Hiremath, A., Iriarte-Vivar, S., Manzane, E., De Oliveira, A. A.,  
16 Poorter, L., Ramanamanjato, J.-B., Salk, C., Varela, A., Weiblen, G. D. and Lerdau, M. T.:  
17 Decomposition in tropical forests: a pan-tropical study of the effects of litter type, litter  
18 placement and mesofaunal exclusion across a precipitation gradient, *J. Ecol.*, 97, 801–811, 2009.
- 19 Powers, J. S. and Schlesinger, W. H.: Relationships among soil carbon distributions and  
20 biophysical factors at nested spatial scales in rain forests of northeastern Costa Rica, *Geoderma*,  
21 109, 165–190, 2002.
- 22 [Rasse, D. P., Rumpel, C. and Dignac, M.-F.: Is soil carbon mostly root carbon? Mechanisms for  
23 a specific stabilisation, \*Plant Soil\*, 269, 341a356, 2005.](#)
- 24 Reyes, I. and Torrent, J.: Citrate-ascorbate as a highly selective extractant for poorly crystalline  
25 iron oxides, *Soil Sci. Soc. Am. J.*, 61, 1647–1654, 1997.
- 26 [Saini, G. R.: Organic matter as a measure of bulk density of soil, \*Nature\*, 210, 1295–1296, 1966.](#)
- 27 Scatena, F. N. and Lugo, A. E.: Geomorphology, disturbance, and the soil and vegetation of two  
28 subtropical wet steepland watersheds of Puerto Rico, *Geomorphology*, 13, 199–213, 1995.
- 29 Scatena, F. N., Moya, S., Estrada, C. and Chinea, J. D.: The first five years in the reorganization  
30 of aboveground biomass and nutrient use following Hurricane Hugo in the Bisley Experimental  
31 Watersheds, Luquillo Experimental Forest, Puerto Rico, *Biotropica*, 28, 424–440, 1996.
- 32 [Schrumpf, M. and Kaiser, K.: Large differences in estimates of soil organic carbon turnover in  
33 density fractions by using single and repeated radiocarbon inventories, \*Geoderma\*,  
34 239xperimental Watershe](#)

steven hall 3/15/2015 6:59 PM

**Deleted:** 2014.

steven hall 4/1/2015 2:11 PM

**Formatted:** Font:Times

- 1 Schuur, E. A. G., Chadwick, O. A. and Matson, P. A.: Carbon cycling and soil carbon storage in  
2 mesic to wet Hawaiian montane forests, *Ecology*, 82, 3182–3196, 2001.
- 3 Sexstone, A., Revsbech, N., Parkin, T. and Tiedje, J.: Direct measurement of oxygen profiles and  
4 denitrification rates in soil aggregates, *Soil Sci. Soc. Am. J.*, 49, 645–651, 1985.
- 5 Sierra, C. A., Trumbore, S. E., Davidson, E. A., Frey, S. D., Savage, K. E. and Hopkins, F. M.:  
6 Predicting decadal trends and transient responses of radiocarbon storage and fluxes in a  
7 temperate forest soil, *Biogeosciences*, 9, 3013s3028, 2012.
- 8 Silver, W. L., Lugo, A. E. and Keller, M.: Soil oxygen availability and biogeochemistry along  
9 rainfall and topographic gradients in upland wet tropical forest soils, *Biogeochemistry*, 44, 301–  
10 328, 1999.
- 11 Silver, W. L., Scatena, F. N., Johnson, A. H., Siccamma, T. G. and Sanchez, M. J.: Nutrient  
12 availability in a montane wet tropical forest - Spatial patterns and methodological considerations,  
13 *Plant Soil*, 164, 129–145, 1994.
- 14 Silver, W. L. and Vogt, K. A.: Fine-root dynamics following single and multiple disturbances in  
15 a subtropical wet forest ecosystem, *J. Ecol.*, 81, 729–738, 1993.
- 16 Soil Survey Staff: Soil survey of Caribbean National Forest and Luquillo Experimental Forest,  
17 Commonwealth of Puerto Rico, United States Department of Agriculture, Natural Resources  
18 Conservation Service, Washington, DC, USA, 2002.
- 19 Sollins, P., Kramer, M. G., Swanston, C., Lajtha, K., Filley, T., Aufdenkampe, A. K., Wagai, R.  
20 and Bowden, R. D.: Sequential density fractionation across soils of contrasting mineralogy:  
21 evidence for both microbial- and mineral-controlled soil organic matter stabilization,  
22 *Biogeochemistry*, 96, 209–231, 2009.
- 23 Stuiver, M. and Polach, H. A.: Discussion: reporting of  $^{14}\text{C}$  data, *Radiocarbon*, 19, 355–63, 1977.
- 24 Stuiver, M., Reimer, P. J. and Braziunas, T. F.: High-precision radiocarbon age calibration for  
25 terrestrial and marine samples, *Radiocarbon*, 40, 1127–1151, 1998.
- 26 Swanston, C. W., Torn, M. S., Hanson, P. J., Southon, J. R., Garten, C. T., Hanlon, E. M. and  
27 Ganio, L.: Initial characterization of processes of soil carbon stabilization using forest stand-level  
28 radiocarbon enrichment, *Geoderma*, 128, 52–62, 2005.
- 29 Teh, Y. A., Silver, W. L. and Scatena, F. N.: A decade of belowground reorganization following  
30 multiple disturbances in a subtropical wet forest, *Plant Soil*, 323, 197–212, 2009.
- 31 Telles, E. de C. C., Camargo, P. B. de, Martinelli, L. A., Trumbore, S. E., Costa, E. S. da, Santos,  
32 J., Higuchi, N. and Oliveira, R. C.: Influence of soil texture on carbon dynamics and storage  
33 potential in tropical forest soils of Amazonia, *Global Biogeochem. Cycles*, 17, 1040,  
34 doi:10.1029/2002GB001953, 2003.

steven hall 4/1/2015 2:11 PM

**Formatted:** Font:Times

steven hall 3/15/2015 6:59 PM

**Deleted:** .

- 1 Thompson, A., Chadwick, O. A., Boman, S. and Chorover, J.: Colloid mobilization during soil  
2 iron redox oscillations, *Environ. Sci. Technol.*, 40, 5743–5749, 2006.
- 3 Thompson, A., Rancourt, D., Chadwick, O. and Chorover, J.: Iron solid-phase differentiation  
4 along a redox gradient in basaltic soils, *Geochim. Cosmochim. Ac.*, 75, 119–133, 2011.
- 5 Torn, M. S., Kleber, M., Zavaleta, E. S., Zhu, B., Field, C. B. and Trumbore, S. E.: A dual  
6 isotope approach to isolate soil carbon pools of different turnover times, *Biogeosciences*, 10,  
7 8067–8081, 2013.
- 8 Torn, M. S., Swanston, C. W., Castanha, C. and Trumbore, S. E.: Storage and turnover of  
9 organic matter in soil, in *Biophysico-Chemical Processes Involving Natural Nonliving Organic*  
10 *Matter in Environmental Systems*, edited by N. Senesi, B. Xing, and P. M. Huang, pp. 219–272,  
11 John Wiley & Sons, Inc., Hoboken, NJ, USA., 2009.
- 12 Torn, M. S., Trumbore, S. E., Chadwick, O. A., Vitousek, P. M. and Hendricks, D. M.: Mineral  
13 control of soil organic carbon storage and turnover, *Nature*, 389, 170–173, 1997.
- 14 Torn, M. S., Vitousek, P. M. and Trumbore, S. E.: The influence of nutrient availability on soil  
15 organic matter turnover estimated by incubations and radiocarbon modeling, *Ecosystems*, 8,  
16 352–372, 2005.
- 17 Townsend, A. R., Vitousek, P. M. and Trumbore, S. E.: Soil organic matter dynamics along  
18 gradients in temperature and land use on the island of Hawaii, *Ecology*, 76, 721–733, 1995.
- 19 Trumbore, S. E.: Comparison of carbon dynamics in tropical and temperate soils using  
20 radiocarbon measurements, *Global Biogeochem. Cycles*, 7, 275–290, 1993.
- 21 Trumbore, S. E., Chadwick, O. A. and Amundson, R.: Rapid exchange between soil carbon and  
22 atmospheric carbon dioxide driven by temperature change, *Science*, 272, 393–396, 1996.
- 23 Trumbore, S. E., Davidson, E. A., Camargo, P. B. de, Nepstad, D. C. and Martinelli, L. A.:  
24 Belowground cycling of carbon in forests and pastures of eastern Amazonia, *Global*  
25 *Biogeochem. Cycles*, 9, 515–528, 1995.
- 26 Vargas, R., Trumbore, S. E. and Allen, M. F.: Evidence of old carbon used to grow new fine  
27 roots in a tropical forest, *New Phytologist*, 182, 710 of eastern Viollier, E., Inglett, P., Hunter,  
28 K., Roychoudhury, A. and van Cappellen, P.: The ferrozine method revisited: Fe(II)/Fe(III)  
29 determination in natural waters, *Appl. Geochem.*, 15, 785–790, 2000.

steven hall 4/1/2015 2:11 PM

Formatted: Font:Times

steven hall 4/1/2015 2:11 PM

Formatted: Font:Times

1 Table 1: Mean (SE) concentrations and stocks of soil C by density fraction, topographic position, and depth; n = 5 for each  
 2 position/depth combination. Total C represents the sum of all density fractions. Means with different letters are significantly different  
 3 ( $p < 0.05$ ).

Position	Depth	Free light	Occluded	Heavy	Free	Occluded				Total C	
		fraction (mg C g <sup>-1</sup> )	light fraction (mg C g <sup>-1</sup> )	fraction (mg C g <sup>-1</sup> )	light C soil	light C soil	Mineral- associated C (mg g <sup>-1</sup> soil)	Bulk density (g cm <sup>-3</sup> )	Free light C (g m <sup>-2</sup> )	Occluded light C (g m <sup>-2</sup> )	Mineral- associated C (g m <sup>-2</sup> )
Ridge	0 - 10 cm	384 (10)	333 (13) bc	49 (2) a	3.7 (0.8)	2.7 (0.3)	47.7 (1.9) c	0.51 (0.02) a	195 (45)	137 (13)	2429 (157)
	10 - 20 cm	347 (34)	249 (24) ab	32 (3) b	1.3 (0.4)	1.3 (0.3)	31.6 (2.6) ab	0.68 (0.05) b	88 (22)	86 (19)	2177 (277)
Slope	0 - 10 cm	357 (22)	333 (29) bc	33 (3) b	2.4 (0.5)	2.6 (1.0)	32.9 (3.3) b	0.69 (0.06) b	142 (30)	161 (67)	2214 (154)
	10 - 20 cm	316 (41)	193 (40) a	22 (5) b	1.5 (0.9)	0.9 (0.3)	21.8 (5.1) a	0.79 (0.03) b	111 (59)	67 (24)	1675 (295)
Valley	0 - 10 cm	325 (28)	376 (10) c	34 (2) b	4.4 (2.0)	4.7 (1.8)	32.7 (2.4) ab	0.71 (0.03) b	311 (149)	341 (130)	2347 (231)
	10 - 20 cm	291 (28)	382 (20) c	27 (2) b	2.6 (0.7)	3.8 (0.8)	26.1 (2.1) ab	0.78 (0.05) b	212 (59)	310 (71)	2037 (214)
											2558 (313)

1 Table 2: Soil characteristics (mean  $\pm$  SE) by topographic position and horizon, n = 5 for each position/depth combination. Means with  
 2 different letters are significantly different.  $\text{Fe}_{\text{ox}}$  and  $\text{Al}_{\text{ox}}$  denote ammonium oxalate extractions on air-dried samples,  $\text{Fe}_{\text{ca}}$  and  $\text{Al}_{\text{ca}}$  were  
 3 measured in citrate/ascorbate extractions of field-moist samples immediately after sampling, and  $\text{Fe}(\text{II})_{\text{HCl}}$  was field-extracted in 0.5 M  
 4 HCl.

Position	Depth	pH	$\text{Fe}_{\text{ox}}$	$\text{Fe}_{\text{ca}}$	$\text{Fe}(\text{II})_{\text{HCl}}$	$\text{Al}_{\text{ox}}$	$\text{Al}_{\text{ca}}$	Live fine roots			
			(mg g <sup>-1</sup> )	(mg g <sup>-1</sup> )	(mg g <sup>-1</sup> )	(mg g <sup>-1</sup> )	(mg g <sup>-1</sup> )	Clay (%)	Silt (%)	Sand (%)	(g m <sup>-2</sup> )
Ridge	0 - 10 cm	4.51 (0.03) a	8.9 (0.8)	25.9 (1.7) c	0.31 (0.05)	2.4 (0.3) b	9.9 (0.8) b	38 (4) a	51 (4) a	11 (2) a	250 (60) b
	10 - 20 cm	4.64 (0.04) a	7.6 (0.6)	24.0 (1.7) bc	0.13 (0.01)	1.9 (0.2) ab	8.1 (0.7) b	49 (4) b	43 (5) b	8 (1) a	140 (30) ab
Slope	0 - 10 cm	4.68 (0.06) a	8.6 (1.0)	16.4 (2.2) ab	0.90 (0.07)	1.5 (0.1) a	4.1 (0.7) a	37 (6) abc	53 (5) ab	10 (2) a	150 (40) ab
	10 - 20 cm	4.75 (0.07) a	7.3 (2.0)	13.7 (2.8) a	2.18 (2.13)	1.6 (0.1) a	3.4 (0.8) a	39 (3) ab	49 (5) ab	13 (2) ab	80 (20) a
Valley	0 - 10 cm	5.23 (0.16) b	7.4 (0.4)	9.9 (1.2) a	0.44 (0.19)	1.6 (0.1) a	2.2 (0.4) a	23 (2) c	51 (5) ab	26 (6) bc	80 (20) a
	10 - 20 cm	5.23 (0.15) b	8.3 (0.3)	9.5 (1.4) a	0.83 (0.27)	1.8 (0.2) ab	2.0 (0.5) a	23 (3) c	45 (4) ab	31 (6) c	60 (20) a

1 Table 3: Mean (SE)  $\Delta^{14}\text{C}$ , modeled turnover of the mineral-associated slow C pools, and the percent contributions of the slow pool and  
 2 passive pools to mineral associated C by topographic position; n = 5 for each position/depth combination except where indicated (see  
 3 Section 2.4 for details on modeling). Slow pool turnover times and the percent slow and passive pools are presented using separate  
 4 models with a three-year lag between plant C fixation and soil C inputs, and with no lag. Slow pool C inputs were estimated by the  
 5 quotient of the slow pool C stock and its mean turnover time.

steven hall 3/15/2015 7:13 PM  
**Formatted:** Width: 11", Height: 8.5"  
 steven hall 3/18/2015 11:31 AM  
**Deleted:**

Position	Depth	Mineral-associated $\Delta^{14}\text{C}$ (‰)	Slow pool	Percent	Percent	C input to slow				
			turnover time (years) <u>with three year lag</u>	slow pool <u>with three year lag</u>	slow pool <u>with no lag</u>	passive pool <u>with three year lag</u>	passive pool <u>with no lag</u>			
Ridge	0 - 10 cm	81 (4) ab	<u>18 (2), n = 5</u>	19 (3), n = 5	<u>70 (3)</u>	76 (3)	<u>30</u>	24	<u>95 (16)</u>	94 (12)
	10 - 20 cm	64 (11) ab	<u>16 (3), n = 4</u>	17 (5), n = 4	<u>65 (7)</u>	70 (5)	<u>35</u>	30	<u>88 (24)</u>	97 (27)
Slope	0 - 10 cm	74 (14) ab	<u>20 (4), n = 4</u>	21 (5), n = 4	<u>67 (8)</u>	72 (5)	<u>33</u>	28	<u>78 (19)</u>	78 (12)
	10 - 20 cm	40 (14) a	<u>16 (4), n = 2</u>	16 (6), n = 2	<u>52 (11)</u>	57 (8)	<u>48</u>	43	<u>54 (25)</u>	60 (22)
Valley	0 - 10 cm	81 (5.0) ab	<u>18 (3), n = 4</u>	20 (1), n = 5	<u>71 (3)</u>	76 (4)	<u>29</u>	24	<u>91 (18)</u>	91 (12)
	10 - 20 cm	86 (3.0) b	<u>20 (2), n = 4</u>	21 (1), n = 5	<u>72 (3)</u>	77 (4)	<u>28</u>	23	<u>75 (15)</u>	75 (11)

6

1 | [Table 4: Linear models of C concentrations, stocks, and slow pool turnover. Normalized model](#)  
 2 | [coefficients \(SE\) are analogous to Pearson's r. See Table 2 for variable descriptions. Models A](#)  
 3 | [and B represent different models with similar AICc.](#)

C concentrations		
Model A	Depth	-0.90 (0.14)
	Al <sub>ca</sub>	0.51 (0.09)
	Fe(II)	0.35 (0.07)
	Al <sub>ox</sub>	0.22 (0.09)
	R <sup>2</sup>	0.88
Model B	Al <sub>ca</sub>	0.41 (0.16)
	Fe(II)	0.35 (0.11)
	Live fine roots	0.27 (0.14)
	Al <sub>ox</sub>	0.25 (0.13)
	R <sup>2</sup>	0.73
C stocks		
Model A	Δ <sup>14</sup> C	0.53 (0.14)
	Al <sub>ox</sub>	0.36 (0.14)
	R <sup>2</sup>	0.46
Model B	Fe(II)	0.58 (0.15)
	Depth	-0.53 (0.29)
	R <sup>2</sup>	0.43
Δ <sup>14</sup> C		
	Fe(II)	0.59 (0.15)
	R <sup>2</sup>	0.35
Slow pool turnover		
	Fe(II)	0.47 (0.18)

1 |  
R<sup>2</sup> 0.25

2

3

4

5

6

7

8

9

10

11

12

13

14

15

16

steven hall 3/17/2015 7:17 PM

**Moved up [4]:** Table 4: Linear models of C concentrations, stocks, and slow pool turnover. Normalized model coefficients (SE) are analogous to Pearson's r. See Table 2 for variable descriptions. Models A and B represent different models with similar AICc.

1    **Figure Captions**

2    Figure 1: Modeled soil  $\Delta^{14}\text{C}$  over time for the slow pool (light grey shading) and the slow +  
3    passive pools (dark grey shading) of mineral-associated C. The dashed line shows atmospheric  
4     $\Delta^{14}\text{C}$  for zone 2 of the Northern Hemisphere. Shaded regions represent 95% confidence intervals  
5    for  $\Delta^{14}\text{C}$  of a given soil sample calculated using Monte Carlo methods that varied 1988  $\Delta^{14}\text{C}$  and  
6    turnover time of passive C. Panels **a** and **b** show trends for individual samples where 2012  $\Delta^{14}\text{C}$   
7    = 109.4 and 51.0 ‰, respectively in 2012, along with four samples from 1988. Panel **c** shows  
8    trends for a sample where 2012  $\Delta^{14}\text{C}$  = 28.5 ‰; this and four other samples with  $\Delta^{14}\text{C}$  less than  
9    the 2012 atmosphere could not be realistically constrained by 1988 surface soil  $\Delta^{14}\text{C}$  due to a  
10    greater abundance of passive C (see Section 2.4 for details on how these samples were modeled).

steven hall 3/18/2015 11:27 AM  
**Deleted:**  $^{14}\text{C}$

11  
12    Figure 2: Carbon concentrations and stocks (mean  $\pm$  SE, n = 5 for each bar) by density fraction,  
13    catena position, and depth increment. Means with different letters are significantly different.  
14    Occluded light fractions were greatest in valleys when compared across positions irrespective of  
15    depth.

16

17    Figure 3: Relationships between  $\Delta^{14}\text{C}$  of the mineral-associated fraction and modeled turnover  
18    time. Circles represent the slow mineral-associated pool, which comprised most (66  $\pm$  2 %) of  
19    total mineral associated C. Squares represent a single-pool model of mineral associated C, which  
20    has commonly been employed in other studies but could not be fit to both our 1988 and 2012  
21    samples.

steven hall 3/18/2015 11:28 AM  
**Deleted:** 71  
steven hall 3/18/2015 11:28 AM  
**Deleted:** 3

1

2     Figure 4: Soil  $\Delta^{14}\text{C}$  vs. Fe(II) concentrations measured in 0.5 M HCl extractions conducted in the  
3     field ( $R^2 = 0.35$ ,  $p < 0.001$ ). Reduced Fe was the best single correlate of  $\Delta^{14}\text{C}$  and slow pool  
4     turnover times; Fe(II) correlated positively with turnover times. Circles indicate samples with a  
5     majority of C in the decadal-cycling slow pool, triangles indicate samples dominated by passive  
6     C of centennial to millennial age. The relationship between  $\Delta^{14}\text{C}$  and C turnover was qualitatively  
7     different between these groups of samples; increased  $\Delta^{14}\text{C}$  implied longer turnover times for the  
8     circles, and smaller  $\Delta^{14}\text{C}$  implied longer turnover for the triangles. The grey horizontal line  
9     indicates the approximate  $\Delta^{14}\text{C}$  of the 2012 atmosphere in the Northern Hemisphere zone 2.



