Dear Dr. Carolina Boix-Fayos,

We thank the editorial board of SOIL for the opportunity to submit a revised version of our manuscript, titled: “Identifying and quantifying geogenic organic carbon in soils – the case of graphite”. The revised manuscript was improved, thanks to the excellent reviews addressing points we had initially overlooked, making it more robust and increasing the transferability of our findings. We will provide again the point to point responses to both the reviews, here below, followed by the marked-up revised manuscript.

On behalf of all authors,

Jeroen H. T. Zethof
Reviewer 1:

I have read with interest the draft untitled "Identifying and quantifying geogenic organic carbon in soils – the case of graphite". Overall, I have found that the draft is very clear. To be published in SOIL, I consider that the authors should provide a proper description of the soils they used in the study. I have also several (rather minor) concerns and questions that have to be answered before I can recommend the publication of this draft in SOIL.

Dear Reviewer,

First of all, we want to thank you for your time reviewing our article, your kind comments and your effort helping us improving the work. We will address your concerns and answer your questions, point by point.

- I consider that the authors should provide a proper description of the soils they used in the study

Thank you for addressing this important point. We indeed forgot the inclusion of a proper soil description. Hereby the revised part of the materials and method section:

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“Top soil and fresh rock samples from a nearby outcrop, were taken from a field site in Rambla Honda, Sierra de los Filabres (37°07'43'' N, 2°22'30'' W / Southern Spain). The area is located in the Nevado-Filabride complex and contains Devonian-Carboniferous slaty mica schist with graphite and garnets crossed by abundant quartz veins (Puigdefábregas et al., 1996). Carbonates found in the soil sample (0.18 % C) originated from pedogenesis and dust deposition, as the parent rock does not contain carbonates. Soil material was taken from the topsoil (0-5 cm, without sieving crust) under the grass tussock Macrochloa tenacissima to ensure a substantial amount of OC was present. The soil itself was classified as Skeletic Leptosol (colluvic) according to the World Reference Base for Soil Resources (WRB, 2014). Additional soil material was collected from a field site near the town of Alboloduy (37°04'09'' N, 2°36'43'' W), hereafter referred to as AB soil, with similar vegetation and climatic conditions. The lithology consists of feldspathic mica schist (IGME, 1979), but without natural graphite and with a relative high CaCO3 content (1.87 % C). The AB soil, classified as Skeletic Leptosol (WRB, 2014), was also sampled from the topsoil, without sieving crust, under the grass tussock Macrochloa tenacissima. The soil samples were dried at 40°C and sieved to ≤ 2 mm.”  
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1) Graphitic C can be found in rocks. Depending on P and T conditions experienced by the sediments, we do not get necessarily pure graphite. I am therefore wondering if the graphite standard material (Merck) is similar to the graphite found in the soils developed in micaschists. I would appreciate to see Raman signatures of the graphites used in the study. We can also imagine that some graphitic C with lots of defaults may evolved before the final oxidation step. In this case, such graphitic C would not be recovered in the ROC fraction. Can the authors discuss or rule out this hypothesis?

It is right that pressure and temperature conditions are important factors determining the degree of graphitization. From literature we have an estimation what the conditions have been for the sampled rock, but the degree of graphitization of the standard is not provided by the manufacturer. Raman analyses were carried out (see Figure S1 here below), which we
will include in the revised version to shed light on potential differences between the standard and natural graphite as found in the rock and soil sample. As both the graphitic schist and the standard show a highly similar pattern, it can be assumed that they have a similar state of graphitization. The D1 (~1350 cm⁻¹), G (~1580 cm⁻¹) and D’ (~1620 cm⁻¹) peaks indicated in Figure S1, can all be attributed to graphitic C, whereby the ratio between the D1 and the sum of all three peaks is a clear indication for the degree of graphitization (Beysacc et al., 2003; Ferrari, 2007). For the Standard (Merck) the ratio is 0.20, while for the graphitic schist we obtained 0.34, which are both indicating well organized carbon (< 0.5, Beysacc et al., 2003). No peaks are observed around the 1200 and 1500 cm⁻¹ bands, including the soil sample, which would have indicated the presence of pyrogenic / black carbon components (Sadezky et al., 2005, Schmidt et al. 2002). The G-peak for the soil sample (calibration 1) coincides with the D’ peak, as could be expected in a sample with different carbonaceous substances, but can also indicate defects in the crystalline graphite structure formed by weathering.

We will include the next paragraph to the method section together with the Raman spectra as in Figure S1:

“Raman spectra of the soil of calibration set 1, the standard (Merck) and graphitic schist were recorded, using a Thermo Scientific DXR Smart Raman Spectrometer, with 532 nm laser and a power output of 9 mW. Before the measurement, samples were pressed in aluminium cups. The obtained peaks were integrated using Lorentzian profiles fitting in Origin 2019.

The spectra of the graphite standard were highly similar to the graphitic schist (Fig. S1 in Supplement). The D1 (~1350 cm⁻¹), G (~1580 cm⁻¹) and D’ (~1620 cm⁻¹) peaks could be attributed to graphitic C. The ratio between the D1 and the sum of all three peaks are clear indicators for the degree of graphitization (Beysacc et al., 2003; Ferrari, 2007). Ratios of 0.20 for the graphite standard (Merck) and 0.34 for the graphitic schist indicated well organized carbon (< 0.5, Beysacc et al., 2003) for both samples. Peaks around the 1200 and 1500 cm⁻¹ bands, which would have indicated the presence of pyrogenic / black carbon components (Sadezky et al., 2005; Schmidt et al., 2002) could not be observed in all samples including the soil sample (Fig. S1).” p. 22, Line 3 –Line 12

Some graphitic C, especially with lots of defects or impurities in its mineral structure, might indeed evolve before the final oxidation step of the smart combustion method, resulting in an underestimation of the graphitic C content of the soil as it is not taken into the ROC fraction. We also hypothesized in the discussion that radicals, released from other minerals by temperatures of 700°C and higher, might induce graphite evolution under anoxic conditions (page 29, Line 7-14). According to the measurements with the smart combustion method about 6% of the total C was lost in samples of quartz + graphite standard (i.e. Calibration set 2), while the Graphitic schist lost 2% of the total C, although the graphite standard had a higher structural organization (lower ratio, as discussed above). We will extend line 28-33 (page 29) of the discussion to increase clarity about the loss of graphite before the final oxidation phase and include a Figure (Figure 8c) visualizing this loss as follow:
“As shown by Hayhurst and Parmar (1998), very small impurities in the graphite can cause a small part of the graphite to pyrolyse during anoxic conditions at higher temperatures. Graphitic C of lesser graphitization, might therefore result in a larger loss of graphitic C during pyrolysis and a greater underestimation of the graphitic C content. Taking a closer look at the measurements of the artificial soil, reveals that a small part of the graphite began to oxidize under anoxic conditions (Fig. 8). The measurement of graphite in quartz, as in calibration set 2, showed that about 6% of the total C was lost during the pyrolysis phase, while for the graphitic schist this loss was 2% (Fig. 8c), resulting in an underestimation of the graphitic C content. Bews et al. (2001) suggested that at temperatures higher than 700°C, radicals like HO₂ and OH might act as reactant with the pure C. Furthermore, in the method comparison study for recovering different types of black C, Roth et al. (2012) suggested a (relatively) strong catalytic effect of oxides on black C oxidation, which was most predominant in soils.” p.29 line 17-25

2) 2° Pyrogenic C (pyOC) can also resist to high T under anoxic conditions. In this case, some pyOC may be recovered in the ROC fraction. What would happen if the studied soil contains both graphite and pyOC? It may have been interesting to add charcoalin the tested mixtures. If the presence of pyOC is a limit to the method, it should be discussed.

We understand from your question that we did not discuss sufficiently the potential interference of thermally resistant OM (like pyrogenic C / black C). Therefore, we will elaborate more on this topic in the discussion section, as showed below. Including other forms of thermally resistant C fractions and examine how we could distinguish between them would be a very interesting topic for further investigation. Nonetheless we can expect, with the current settings, that pyrogenic C might end up in the ROC fraction of the smart combustion method.

“When the sample contains other forms of thermally resistant OM or even black C, which are not pyrolyzed during the anoxic phase, this C component is likely to end up in the graphitic C fraction with the smart combustion method. Especially as most temperature boundaries are empirically derived (Pallasser et al., 2013; Ussiri et al., 2014), a pre-test with continuous heating under oxic conditions is therefore recommended to assess the number of C-containing substances present in the sample by the occurrence of peaks. Further studies should focus on temperature boundaries of different substances in relation to their properties and see, for instance, how graphitic C can be distinguished from other thermally stable C components.” p.30, line 6- line 11

3) 10 samples for calibrating a model is definitely a too low number of samples. It would have been highly surprising to get nice results with such a low number of samples. It may have been interesting to use all the samples to design calibration models. We can’t exclude that with a nice sample set containing 500 samples with known graphite concentrations, a convincing IR-based model can be designed.
We agree that more samples could potentially improve the model. On the other hand, it is frequently shown in the literature that the performance of IR spectroscopic models for predicting soil properties increases with sample set homogeneity (e.g., Grinand et al., 2012), i.e., calibration and validation become more precise when focusing on samples from similar or identical sites and soil matrixes. Here, we like to point out that two representative matrix substances were used as calibration samples: quartz sand and the soil of interest (soil 1), which was later used for validation. The $R^2$ and RMSEP of the calibrations were quite sufficient ($R^2 = 0.96$ and 0.99; RMSEP = 0.24 and 0.10). The samples with the unknown graphite concentrations were of exactly the same matrix (Quartz, soil 1). So, the models we used were very specific in addition to the high $R^2$ and low RMSEP. Since further samples are not available, we calculated a model including both calibration data sets, soil 1 and quartz. This PLSR model used 3 components and an $R^2$ of 0.96 and an RMSEP of 0.24 (Fig. R1). These values were at the same level as found for the single models (see above). Nevertheless, all models substantially overestimated the graphite content. Therefore, we do not think that the use of more samples of different origin would improve the prediction/validation. Against the backdrop of the literature (specific graphite absorption bands that have been reported in the literature are only valid for oxidized graphite), the failed predictions of the graphite contents were plausible. We modified the discussion correspondingly, however, - if you agree - we do not intend to add the Figure (here shown as Fig. R1) to the text:

“The calibration between infrared spectra and graphite contents of the calibration sets yielded promising results (Figs. 1a and 1b) and could also be used for a cross-validation (Fig. 5). Although the same substrate materials and similar contents of graphitic C were used in the validation, the graphite contents were systematically over-predicted. Despite the apparent quality of the calibration, this failure could have been caused by the relatively low number of calibration samples. Note that the use of the two calibration data sets, soil and quartz, in a joint PLSR model ($R^2 = 0.96$ and an RMSEP = 0.24; 3 components) did not improve the calibration nor the prediction accuracy. It cannot be excluded that a higher number of samples for the calibration could improve the PLSR model and the prediction results. Further, Raman spectroscopy might be an alternative approach for quantifying graphite in soil samples (e.g., Sparkes et al., 2013; Jorio and Filho, 2016).”
4) The authors hypothesized that ROC content would match graphite content. It is not too far but not perfect. Why don’t the authors try to design a model based on ROC results as they did with IR and TGA results?

Thank you for the suggestion to create a model for correcting the ROC value. We considered this as well, but, as can be seen in supplementary Figure S5 of the manuscript, the total carbon measured with the smart combustion method (i.e. the Soli-TOC device) is the same as what is obtained by the more accurate elemental analyser. Therefore, the slight underestimated graphitic C content results from the differentiation between carbon fractions by the temperature-oxidation program and not as a result of the direct output of the sensor. Furthermore, correcting the ROC/graphitic C value using the calibration set would not improve the graphitic C estimation as a model build with calibration set 1 would result in a slight overestimation of calibration set 2 (notice slope > 1.00) and vice versa (see Figures R2 and R3, here below). This difference in underestimation of the ROC fraction was attributed to impurities in the graphite and/or presence of radicals, as discussed in the second paragraph of section 4.3. Especially the presence in radicals will differ from sample to sample, as it depends on the matrix composition, which would give the same matrix issue in the model creation as with the FTIR.
and TGA methods. Therefore, we suggest keeping the ROC content as it is derived by the device and clarify this decision in the method section by adding:

“**The Soli-TOC device** directly converts the NDIR signal to C content of the different components, as calibrated with CaCO$_3$. Creating an additional model to correct the C output, introduces an additional error in the measurements. Therefore, we analysed the direct C output, as measured in the ROC fraction. Triplicate measurements were averaged, whereby the average coefficient of variation between replicates was 2.7%. A Pearson correlation test was performed between the obtained ROC data and calibration sets to evaluate the graphite content measurements.” p.24, line 11-17

Figure R2. Correction of the ROC values by using a model based on calibration set 1 (graphitic soil with added graphite). In orange the original measured graphitic C content (ROC) of calibration set 1 is plotted against the added graphite. As can be seen by the linear trend line, the graphitic C content is originally overestimated. A simple correction model was created, resulting in an exact estimation (Blue), but the same model resulted in an underestimation of graphitic C in calibration set 2 (Grey, Quartz with added graphite).
Figure R3. Correction of the ROC values by using a model based on calibration set 2 (Quartz with added graphite). In yellow the original measured graphitic C content (ROC) of calibration set 2 is plotted against the added graphite. As can be seen by the linear trend line, the graphitic C content is originally slightly underestimated. A simple correction model was created, resulting in an exact estimation (purple), but the same model resulted in an overestimation of graphitic C in calibration set 1 (Green, Quartz with added graphite).

5) I do not understand the Figure 5. I suggest improving the explanations on this Figure or removing it.

We hope that Figure 5, in the manuscript, which summarizes the results of the three tested methods, becomes clearer with this extended Figure description:

“Figure 5: Overview of the predicted amount of graphite in the calibration sets (squares/diamonds), artificial soil (inset, circles/triangle), graphitic schist (inset, stars) and AB soil (right pointing triangle) as measured with the different methods. Black symbols: graphite prediction by FTIR, model from calibration set 1; White: graphite prediction with FTIR, calibration set 2; Orange: graphite prediction by TGA, model from calibration set 1; Green: graphite prediction by TGA, model from calibration set 2; Grey: graphite prediction by smart combustion. Exact data is given in Table 2.”
Reviewer 2:

This is a very well written manuscript that reports a methods comparison for the identification/quantification of graphite in soil. The writing is excellent, and the reporting of results is clear. But the study have a few important shortcomings that I’d like to see addressed before I would deem it acceptable for publication.

Dear Reviewer,

First of all, we want to thank you for your time reviewing our article and your effort helping us improving the work. Following the numbering in your review, we provide answers to your concerns and questions.

1. The manuscript does not provide sufficient context for focusing exclusively on graphite. The reader might interpret the current rationale as a narrow justification for using graphite in the experiment. I understand that graphite may form from metamorphic processes, but is the goal of the study to quantify graphite specifically, or geogenic organic C more broadly? The former seems far too narrow a prospect given the wide range of forms of geogenic C, and the latter is underdeveloped in the study. Without further elaboration, graphite seems a little too specific. Galy, Hemingway and others are not specific when they refer to “lithospheric” C, Ussiri (cited in the manuscript), Chan (2017, Themochimica Acta) and others have targeted coal, and there is a large and growing literature on pyrogenic C in soils. How would the presence of these affect results? Can graphite be distinguished from other forms of thermally recalcitrant organic C? The authors do well to distinguish between carbonates (which varies widely among dolomite, calcite, etc.) and thermally recalcitrant C, but have not adequately elaborated on graphite versus other forms of geogenic C, let alone pyrogenic C. The distinction between the latter two is obvious using 14C, but the issue here is among geogenic C forms.

   Yes, you’re right that we want to develop a quantification method just for graphite. As discussed by for instance Ussiri et al. (2014), there is a wide continuum of geologically altered organic compounds, whereby a general quantification method seems not to be possible as spectral and thermal properties gradually change by the degree of transformation of the organic matter. Furthermore, Roth et al. (2012) showed for several black carbon types that there is no ideal method to quantify all their tested black carbon types, especially in soil environments. As to our knowledge no previous study has attempted to quantify graphitic carbon, especially not in a soil environment, it hampers studying carbon dynamics in soils developed on sites with graphite containing parent materials, as experienced by ourselves. Therefore, we will re-write parts of the introduction to put a stronger focus on graphitic carbon:

   “Organic C (OC) of geogenic origin, which has gained less attention until now, is formed when organic compounds in sediments undergo coalification or kerogen transformation during diagenesis. Under high pressure and appropriate temperature conditions this process can continue into the formation of graphitic C, although well-crystallized pure C is rarely produced (Oohashi et al., 2012; Buseck and Beysack, 2014). Redox transformations during metamorphoses of carbonates leads also to the formation of highly crystalline graphite (Galvez et al., 2013). Intruding hydrothermal fluids in the earth’s crust forms a third source of graphitic C during rock formation, which produces the purest graphite crystals (Rumble, 2014).
relatively pure and stable form of C is highly chemical inert, although impurities from the parent material increase its chemical reactivity (Beyssac and Rumble, 2014). Via tectonic processes graphite bearing rocks can reach the earth surface where they are subjected to physical and chemical weathering. Therefore, graphitic C occurs mainly in rocks from orogenic belts and in metasedimentary rocks in old cratons and might be a quite common bedrock for soil development (Hartmann and Moosdorf, 2012; Buseck and Beyssac, 2014).

The fate of geogenic graphite under weathering and soil formation has rarely been studied, possibly due to the lack of methods for determining and quantifying geogenic graphite beyond the background of soil OC (OC). There are some indications that a substantial part of the geogenic graphitic C is actually lost in the pathway from rock weathering to (marine) sedimentation (Galy et al., 2008; Clark et al., 2017). Isolated naphthalene-degrading bacteria from contaminated soil proved to oxidize and degrade graphitic materials, questioning the assumed biological inactivity of graphite (Liu et al., 2015). In a recent study, Hemmingway et al. (2018) estimated that 2/3 of the graphitic C is oxidized during soil formation, strongly facilitated by soil microbial activity.” p.18, line 13-30

To answer your question if graphite can be distinguished from other forms of thermally recalcitrant organic C, we can be sure that with smart combustion method, in its current settings, is certainly not capable to do so as it lumps all the oxidizable carbon components that evolve between 400 and 900°C in one fraction. From FTIR spectroscopy, a valid proof of the existence of pyrogenic C in the soil samples is not possible. For a clear evidence, spectrometric techniques such as e.g. Pyrolysis-Field Ionization Mass Spectrometry (Py-FIMS) would be necessary, as shown in Leue et al. (2016), which were beyond the scope of the study.

Furthermore, we carried out Raman spectra of the soil of calibration set 1 and graphitic schist samples, as shown in Figure S1 at the end of this letter. The D1 (~1350 cm⁻¹), G (~1580 cm⁻¹) and D’ (~1620 cm⁻¹) peaks indicated in Figure S1, can all be attributed to graphitic C, whereby the ratio between the D1 and the sum of all three peaks is a clear indication for the degree of graphitization (Beysacc et al., 2003; Ferrari, 2007). The Raman spectra showed no signs of pyrogenic or black carbon in both the soil sample (Fig. S1), which would have created a peak at 1200 cm⁻¹ and/or a clear shoulder at 1500 cm⁻¹ (Sadezky, 2005; Schmidt et al., 2002). Nonetheless it is a good point that care should be taken to distinguish between pyrogenic/black carbon and the graphitic C. Therefore, we will include in the discussion section more clarification on this point:

“When the sample contains other forms of thermally resistant OM or even black C, which are not pyrolyzed during the anoxic phase, this C component is likely to end up in the graphitic C fraction with the smart combustion method. Especially as most temperature boundaries are empirically derived (Pallasser et al., 2013; Ussiri et al., 2014), a pre-test with continuous heating under oxic conditions, is therefore recommended to assess the number of C-containing substances present in the sample by the occurrence of peaks. Further studies should focus on temperature boundaries of different substances in relation to their
Concerning the mentioned literature, we want to clarify our choice for them hereby.

- Galy et al. (2008), doi: 10.1126/science.1161408, uses different sediment (samples) to study the origin of petrologic carbon. They state that part of the graphite, present in the parent material, is no longer present in the downstream sediments based on Raman spectra and transmission electron microscopic images, indicating that it has been oxidized during the erosion/weathering process. Although the degree of graphitization is important for the preservation, as mainly the lesser graphitized carbon was lost, we considered this reference as a good case study to show that graphite is somewhere lost (i.e. used) in the weathering process of the parent material, while this has hardly been documented or studied.

- Hemingway et al. (2018), doi: 10.1126/science.aao6463, although the title of their paper indicate only lithographic C, a more detailed description of the lithology of the study area revealed that they were dealing with graphite containing metamorphic rocks, whereby the rock contained graphite of different degrees of graphitization. More detailed information can be found in the supplementary of the Hemingway article and in the paper of Hilton et al. (2010), doi.org/10.1016/j.gca.2010.03.004

- Ussiri et al. (2014), doi: 10.1016/j.geoderma.2013.09.015, have made a comprehensive review on the current available methods and definitions to distinguish geogenic carbon (mainly coal) from other carbon sources, including inorganic carbon. They also consider that “coalification” (i.e. early stage of graphitization) is a process, whereby the stages of coalification determine their susceptibility for a certain analytic method. They furthermore emphasize that no standardized method exists to identify and separate geogenic carbon from other carbon sources in soils. In this context we used this paper, as, to our knowledge, no other comprehensive discussion on distinguishing of (geogenic) organic and inorganic carbon in soil samples exist.

Thank you for suggesting the paper of Chan et al. (2017), doi: 10.1016/j.tca.2017.02.006, as we were not aware of this study. However, the software they applied to process their TGA measurements further into the endmembers is not available to us and could therefore not be tested.

2. As a result of the discussion above, and more generally for a method development study, I found the number and range of materials used too small. Only one graphite-containing natural soil and one carbonate-containing natural soil were used for validation. The artificial soil mixtures were made from one OM source far removed from the natural soil (Germany vs. Spain), and “neat” mineral specimens (quartz, muscovite, CaCO3 (not dolomite or calcite?)). While I understand the desire to create a reductionist, simplified system for initial testing, the result is only nine samples of one mixed-matrix to generate the calibration. This is a highly undersampled relationship. This is critically important because the authors are right to highlight matrix effects, but they do not adequately account for these in the design of the calibration/correlation study. CaCO3 is not dolomite or calcite, soil minerals are often interstratified, and graphite likely exists in a mineral-associated form fused to the mineral matrix. None of these incipient properties of soil are accounted for in the method development – making the results of
limited value. It is a nice proof of principle, but the study needs to go well beyond this given the current state of the literature. I don’t see how this study substantially move us further than some of the studies cited within it.

As we are focussing on developing a quantification method for graphite in soil matrix, we started with creating samples of pure quartz mixed with different quantities of graphite to test the methods available to us. The next step was proofing that we could identify/distinguish graphite from other typical present substances in the soil, for which we created the artificial samples. To our opinion OM from a forest floor, although geographically not close to the natural samples, is still providing a typical input signal for OM as would be found in rangeland soils from Southern Spain.

Another important carbon component in semiarid soils is carbonate, mostly in the form of calcite. It is true that the used pure CaCO$_3$ is not exactly same as pedogenic or geogenic calcite, but for the simplified artificial soil and to test the differentiation between carbonates and graphitic C it should be sufficiently similar. The most important difference should be visible in the thermal properties, whereby pedogenic carbonates tend to start decomposing at 550-600°C, with the major decomposition peak coming around 750°C (e.g. Apesteguia et al. 2019; Pallaser et al. 2013), while the purer calcite started with decomposition just above 600°C, it reached its major decomposition peak around 725°C (Fig. 2 in the manuscript). Note that we also include an additional soil, “AB Soil”, which contains a large amount of pedogenic calcite. In the Figure 5 it can be seen that there is no difference in predicted graphite between the AB soil and the quartz, both spiked with graphite standard. This indicates that there is no significant influence of the pedogenic carbonates on the graphite prediction with the smart combustion method.

In the next step we tested natural soil and graphitic rock. By creating a sample set with different amounts of graphite added to the soil, we tested the methods for their ability to quantify graphite. By taking also a carbonate rich soil with a different mineral composition (i.e. feldspathic and without garnets) we also took the influence of mineralogy on the ability of graphite quantification into account, which resulted in the matrix effect highlighted in the manuscript.

In our point of view, further study is only realistic using smart combustion or a comparable method (like EGA or Rock-Eval) as they proved to be most promising. The alternating between oxic and anoxic conditions during a measurement is also a not often employed method to differentiate between soil carbon components. For FTIR it is frequently shown in the literature that the performance of IR spectroscopic models for predicting soil properties increases with sample set homogeneity (e.g., Grinand et al., 2012), i.e., calibration and validation become more precisely when focussing on samples from similar or identical sites and soil matrixes. Calculating a model including both calibration data sets, soil 1 and quartz resulted in a PLSR model with an $R^2$ of 0.96 and an RMSEP of 0.24. These values were the same level as found for the single models. Nevertheless, all models substantially overestimated the graphite content. We will highlight this even further in the discussion (see below and 2nd addition under point 1). The use of graphite addition might be most practical for testing quantification of graphitic C in different mineral matrixes. This should also shed more light on how geogenic C, pyrogenic C and carbonates could be distinguished from each other.
“The calibration between infrared spectra and graphite contents of the calibration sets yielded promising results (Figs. 1a and 1b) and could also be used for a cross-validation (Fig. 5). Although the same substrate materials and similar contents of graphitic C were used in the validation, the graphite contents were systematically overpredicted. Despite the apparent quality of the calibration, this failure could have been caused by the relatively low number of calibration samples. Note that the use of the two calibration data sets, soil and quartz, in a joint PLSR model ($R^2 = 0.96$ and an RMSEP = 0.24; 3 components) did not improve the calibration nor the prediction accuracy. It cannot be excluded that a higher number of samples for the calibration could improve the PLSR model and the prediction results. Further, Raman spectroscopy might be an alternative approach for quantifying graphite in soil samples (e.g., Sparkes et al., 2013; Jorio and Filho, 2016).”

3. The authors identified one of the challenges of thermally distinguishing forms of C as the determination of threshold temperatures. However, the description of how TGA data was processed is inadequate. Phrases like “models were created from the calibrations sets” and “the best temperature limits for quantification of graphitic C in the calibration sets was determined” are not reproducible. These steps may be the most critical step of the process, but even the most experienced expert in this field would be unable to verify and repeat it. More detail is required here. There are also no details provided on how these data were used in the calibration. The selection of different threshold temperatures somewhat undermines the “smart combustion” approach if it cannot be universally applied. Perhaps there is some elaboration required in the discussion about how much control over threshold temperature there is available with such an instrument, and whether the DIN methods are suitable/adequate standards.

We are sorry that the data processing was not clearly stated. Together with your next point, we have extended and revised our description of the data processing, which is hopefully now better reproducible (see point 4 for suggested revision).

We will emphasize our discussion of the DIN method to highlight that the smart combustion method is a rather standardized version of the EGA method and that it should therefore be applied with care as long as the thermal boundaries between substances are unknown.

“Because we focused in this study on the ROC component, which significantly correlated with the graphite content, a consideration of the other components mentioned in the DIN19539-standard was beyond the scope of this study. Nonetheless, we found indications that the thermal boundaries defined in the DIN19539-standard are not ideal to differentiate between soil OM and inorganic C (Fig. 8). As most carbonates start to decompose at temperatures of 550°C (Földvári, 2011), it might be more suitable to increase the level for the TOC component from 400 to 500°C. Only when black C is present in the sample, which might oxidize between 375 and 540°C (Roth et al., 2012), this might lead to an overestimation of the TOC content. Using TGA simultaneously with differential scanning calorimetry, water and CO$_2$/H$_2$O flux measurements (i.e. evolved gas analysis, Fernández et al., 2012) or with the Rock-Eval method focusing on hydrocarbon, CO$_2$ and CO release (Behar et al., 2001), could improve the development of a more standardized method applicable to soils using combustion elemental analysers. The overlap between the thermal properties of different C components emphasizes the need to always first consider
what is present in the sample and what might interfere with the considered methodology, before applying a fast and standardized analytic method.” p. 30, line 29 – p. 31, line 8

4. I would argue that since this is such a key/core component of the study, that it deserves its own separate subsection within the methods (ie, statistical analyses). I’m not 100% sure that my comments here will be relevant or correct because it was difficult to follow precisely how each of the calibrations were generated. But if I read it correctly, the authors appear to use different methods to generate the calibration curves depending on the quantification method. PLS was used for the FTIR method of quantification, not sure how it was done for what data from TGA, and Pearson correlation for the “smart combustion” data. In all cases, the independent variable should be clearly identified – presumably the “known” quantities of graphite in the mixture (though this should be verified using total C analyses). The dependent variable should also be clearly identified. Lastly, while the calibration/regression lines are shown with error envelopes, none of the data points have errors associated with them. Were the analyses replicated at all? I found on mention of these. Clearly, each of the analytical methods has instrument/analytical error associated with them. How were these accounted for in the study?

As every method requires its own statistical analysis, we decided to split the statistical analysis section over the different method sections. As this seems not to make it clearer, we will create subsections for each method, clearly stating the statistical analysis conducted. Furthermore, we have extended the part of the TGA method to increase clarity on the data processing:

“2.2.1 Statistical analysis of the FTIR spectroscopic data

The partial least squares regression (PLSR) analyses of correlations between the transmission or DRIFT spectra and the graphite contents (0.1 - 4 %) of the samples were performed using R, Version 3.1.1 (R Core Team, 2014) with module PLS (SIMPLS, cross-validation: leave-one-out) of Mevik et al. (2018). The signal intensities were used as independent variables, the graphite content as dependent. The number of components used in the calibration models followed the lowest predicted root-mean-square error (RMSEP) of the specific datasets. The scores and loadings were plotted for the two main components determining most of the variances of the DRIFT spectra. Larger absolute loading values of signal intensities in certain WN regions imply a greater importance of these WN for the cumulated values of the principal component 1 or 2 displayed in the score plot.” p. 22, line 22-30

“2.3.1 Statistical analysis of the TGA data

TGA measurements were processed and thermal mass loss data obtained via the Proteus Thermal Analysis software (NETZSCH, Hanau, Germany). Initial testing with triplicates revealed that the repeatability error, expressed as coefficient of variation of the mass loss in areas of interest, was below 2%. Measured sample mass data are fitted with the spline function over the temperature, with steps of 1°C. Further analyses of the obtained data were conducted using R, Version 3.5.1 (R Core Team, 2018). Using the module PLS (Mevik et al., 2018), a model was created for each calibration set to determine the graphite content based on the measured mass change in a certain temperature range and the known graphite addition, using the mass change as independent variable. By iterating the model creation over the temperature range from 400 to 1075°C with minimum step of 5°C difference and recording the slope, intercept and RMSEP of each created model, we evaluated the temperature range describing the graphite content of both the calibration sets most precisely. For this purpose, the RMSEP of these models
were used (Fig. S4, see Supplement), and a single temperature range that fit both calibration sets was determined.” p. 23, line 12-25

“2.4.1 Statistical analysis of the Smart Combustion data

The Soli-TOC device directly converts the NDIR signal to C content of the different components, as calibrated with CaCO₃. Creating an additional model to correct the C output was found, is introducing an additional error in the measurements. Therefore, we analysed the direct C output, as measured in the ROC fraction. Triplicate measurements were averaged, whereby the average coefficient of variation between replicates was 2.7%. A Pearson correlation test was performed between the obtained ROC data and calibration sets to evaluate the graphite content measurements.” p. 24, line 11-17

5. The methods chosen for comparison were not especially the best available or most appropriate given the current literature. It almost seems as though there was a foregone conclusion that “smart combustion” was going to be the best and the need was to validate these. However, as the authors correctly point out, the use of “smart combustion” doesn’t solve the problem of selecting a threshold temperature for distinguishing forms of organic C because it either uses the “wrong” temperature or at a minimum uses the same threshold for all samples. The selection of FTIR was intriguing since the presence of interference bands is well documented. I thought there might be a better spectral method (NIR, MIR, Raman, etc.) that would be better suited to the task. Similarly, the use of TGA has well documented shortcomings in that mass loss reactions are not all attributable to organic C combustion/pyrolysis. In fact, the only method that directly quantified carbon in the current study was the “smart combustion”. While the authors discuss the possibility of combining methods, they seem to have missed the opportunity for using EGA during ramped heating - which is essentially what “smart combustion” is.

We encountered an analytical problem during our work by the standard methods employed for carbon quantification and recognized the need for the development of a quantitative method to determine graphite in soils. Therefore, we developed a test program with methods available to us to overcome this issue. As we expect that others might encounter similar problems (because of the occurrence of graphite containing rocks as the base for soil formation), we decided to share our experience with the scientific community.

It is completely right that mass loss obtained by the TGA method are difficult to correlate with organic C, as OM tends to be chemically heterogeneous. Nonetheless, graphitic C is almost pure C and by using graphite addition tests, we expected the TGA method to correlate better with the graphitic C content than with OM content of soil samples.

Smart combustion, in our case with the Soli-TOC device of Elementar company, is indeed not the most flexible method to explore thermal properties of different carbonaceous substances as provided with most EGA methods. On the other hand, the smart combustion method provides a standardized set-up making the measurement of larger sample sets more feasible. We will highlight this further in the discussion section 4.4, as given under point 3.

Concerning the potential suitability of Raman spectroscopy, we added a sentence to the discussion. We want to notice that NIR and MIR (suggested by the reviewer as suitable for graphite determination) are FTIR techniques. As mentioned in the methods, we used FTIR spectroscopy in the mid-infrared wavelengths (2.5 \( – 25 \mu m \)). Unfortunately, Raman spectroscopy was not fully available for this study. Nonetheless we will add the Raman spectra below for clarification about the degree of
graphitization of the graphitic schist / standard graphite and to show that there were no indications for the presence of pyrogenic C, as discussed under point 1. Furthermore, we will include a few sentences on the potential of Raman spectroscopy for developing a quantification method:

“As Raman spectroscopy is suitable to distinguish graphitic C and determine its degree of graphitization, it seems to be a promising method. Nonetheless, to use Raman spectrometry for quantification of substances in a soil matrix, further studies should first focus on standardization of sample preparation, as it has a large influence on the measured intensities and baseline determination and thereby the direct quantification of components (Beysacc and Lazzeri, 2012; Sparkes et al., 2013).”

Between p.30, line 25-28

Figure S1. Raman spectra of the graphite standard (black), graphitic schist (red) and soil of calibration set 1 (i.e. natural graphite containing soil, blue). Vertical lines indicate the peaks for amorphous carbon (1342/1339 cm\(^{-1}\)) and peaks for graphitic carbon (1575 cm\(^{-1}\) standard/schist and 1596 cm\(^{-1}\) for soil of calibration set 1). Indicated are the D1 band (1350 cm\(^{-1}\)), caused by plane defects and heteroatoms in the carbon structure, G (1580 cm\(^{-1}\)), crystalline carbon i.e. pure graphite, and D’ band (1620 cm\(^{-1}\)), caused by disordered graphitic lattices.
Identifying and quantifying geogenic organic carbon in soils – the case of graphite

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Abstract

A widely overlooked source of carbon (C) in the soil environment is organic carbon (OC) of geogenic origin, e.g. graphite, occurring mostly in metamorphic rocks. Appropriate methods are not available to quantify graphite and to differentiate it from other organic and inorganic C sources in soils. This methodological shortcoming also complicates studies on OC in soils formed on graphite-containing bedrock, because of the unknown contribution of a very different soil OC source.

In this study, we examined Fourier-transform infrared (FTIR) spectroscopy, Thermogravimetric analysis (TGA) and the smart combustion methods for their ability of identifying and quantifying graphitic C in soils. For this purpose, several artificial soil samples with graphite, CaCO3 and plant litter as usual C components were created. A graphitic standard was mixed with pure quartz and a natural soil for calibration and validation of the methods over a graphitic C range of 0.1 to 4%.

Furthermore, rock and soil material from both a graphite bearing schist and a schist without natural graphite were used for method validation.

FTIR: As specific signal intensities of distinct graphite absorption bands were missing, calibration could only be performed on general effects of graphite contents on the energy transmitted through the samples. The use of samples from different mineral origin yielded significant matrix effects and hampered the prediction of geogenic graphite contents in soils.

TGA: Thermogravimetric analysis, based on changes in mass loss due to differences in thermal stabilities, are suggested as a useful method for graphite identification, although (calcium) carbonate and graphitic C have a similar thermal stability. However, the quantitative estimation of the graphite contents was challenging as dehydroxylation (mass loss) of a wide range of soil minerals occur in a similar temperature range.

Smart combustion: The method is based on measuring the release of C during a combustion program, quantified by a non-dispersive infrared detector (NDIR) being part of a commercial elemental analyser, whereby carbonates and graphitic C could be separated by switching between oxic and anoxic conditions during thermal decomposition. Samples were heated to 400°C under oxygen rich conditions, after which further heating was done under anoxic conditions till 900°C. The residual oxidizable carbon (ROC), hypothesized to be graphitic C, was measured by switching back to oxygenic conditions at 900°C.
Test samples showed promising results for quantifying graphitic C in soils. For the purpose of quantifying graphitic C content in soil samples, smart combustion was the most promising method of those who have been examined in this study. However, caution should be taken with carbonate rich soils as increasing amounts of carbonate resulted in an underestimation of graphitic C content.

5 1 Introduction

In the past decades, global carbon (C) cycling has gained more and more attention. As an important component in this cycle, the soil C reservoir consists of many different forms and types of carbonaceous substances, each with unique turnover times and functions. Probably the most widely studied and dynamic soil C component is fresh organic matter (OM), derived from litter input, decaying organisms and plant exudates (summarized in Blankinship et al., 2018). Another well studied soil C component is inorganic C in the form of carbonates, which form an important part of the soil C pool, especially under arid climate conditions (Zamanian et al., 2016; Apesteguie et al., 2018). Black C, defined as a broad set of highly condensed carbonaceous by-products (e.g. soot) and residues (e.g. charcoal) of incomplete fossil fuel and biomass combustion, has obtained an increasing interest during the past decades (Agarwal and Bucheli, 2011). Organic C (OC) of geogenic origin, which has gained less attention until now, is formed when organic compounds in sediments undergo coalification or kerogen transformation during diagenesis. Under high pressure and appropriate temperature conditions this process can continue into the formation of graphitic C, although well-crystallized pure C is rarely produced (Oohashi et al., 2012; Buseck and Beysacc, 2014). Redox transformations during metamorphoses of carbonates leads also to the formation of highly crystalline graphite (Galvez et al., 2013). Intruding hydrothermal fluids in the earth’s crust forms a second source of graphitic C during rock formation, which produces the purest graphite crystals (Rumble, 2014). This relatively pure and stable form of C is highly chemical inert, although impurities from the parent material increases its chemical reactivity (Beysacc and Rumble, 2014). Via tectonic processes graphite bearing rocks can reach the earth’s surface where they are subjected to physical and chemical weathering. Therefore, graphitic C occurs mainly in rocks from orogenic belts and in metasedimentary rocks in old cratons and might be a quite common bedrock for soil development (Hartmann and Moosdorf, 2012; Buseck and Beysacc, 2014).

The fate of this geogenic graphite under weathering and soil formation has rarely been studied, possibly due to the lack of methods for determining and quantifying geogenic graphite beyond the background of soil OC (OC). There are some indications that a substantial part of the geogenic graphitic C is actually lost in the pathway from rock weathering to (marine) sedimentation (Galy et al., 2008; Clark et al., 2017). Isolated naphthalene-degrading bacteria from contaminated soil proved to oxidize and degrade graphitic materials, questioning the assumed biological inactivity of graphite (Liu et al., 2015). In a recent study, Hemmingway et al. (2018) estimated that 2/3 of the graphitic C to getis oxidized during soil formation, strongly facilitated by soil microbial activity.

The necessity of identifying and quantifying geogenic C becomes obvious when considering the widely used 14C dating method to measure the mean age of substances and their turnover rates (Trumbore, 2000). As C is depleted in 14C over
50.000 years following burial, geogenic C will contain no longer $^{14}$C and might dilute the $^{14}$C content of younger C pools (Rumpel and Kögler-Knabner, 2011). Although the dilution effect might be of less importance for the C pool in topsoil, it can become more important in subsoil as the C gets older and geogenic C might have a more dominant share in the total C pool (Rumpel and Kögler-Knabner, 2011). If geogenic C cannot be distinguished from the “normal” soil organic C derived from fresh OM, age and turnover time of soil OC will be overestimated. For instance, in the case of Hemmingway et al. (2018), the OM in the upper soil layers should have aged for over 20.000 years to explain the derived $^{14}$C signal, which is unlikely under the local tropical conditions. Although exact figures are unknown, more than one fifth of the global lithology may potentially contain has the potential of containing graphitic C (Hartmann and Moosdorf, 2012). This illustrates the necessity to distinguish between the different C sources to be able in order to study their fate and residence time in soils.

Several quantification methods, based on optical, thermal and chemical properties, have been established for identification and quantification of various C sources. Probably the most widely applied method is measuring C released after dry combustion. However, dry combustion is not suitable to differentiate between when different (in)organic C components, are present, they might not be differentiated by dry combustion and which should therefore have to be corrected for or even removed. Several pre-treatments, like thermal differentiation (e.g. Apesteguia et al., 2018) or removal by acid fumigation (e.g. Harris et al., 2001) in the case of carbonates, have been established to differentiate between the different carbonaceous substances.

Other spectroscopic techniques, such as Fourier-transform infrared (FTIR) spectroscopy in the mid-infrared range (wave lengths 2.5 – 25 µm) is a widely used technique to qualify organic and mineral matter in soils in terms of its functional groups (Smith, 1995; Parikh et al., 2014). Transmission FTIR yields highly resolved spectra with clearly separated absorption peaks, but requires sample dilution. In contrast, diffuse reflectance infrared Fourier transform (DRIFT) can be applied to undiluted soil samples (Reeves, 2003), in particular to determine OC contents of soils (e.g. Reeves et al., 2002; McCarty et al., 2002) via partial least squares regression (PLSR) (e.g. Janik et al., 1998). For employing PLSR, DRIFT spectra are calibrated by OC contents obtained with standard techniques such as dry combustion (e.g. Vohland et al., 2014).

Mid-infrared spectra from graphite show few absorption bands. Among the bands at wave numbers 2200, 1587, 1362, and 830 cm$^{-1}$ reported by (Friedel and Carson, 1971), the bands at 1587 and 868 cm$^{-1}$ were attributed to optical lattice vibrational modes of graphite (Chung 2002) while the other two bands cannot be assigned to specific functional groups have not been found. Tan et al. (2013) reported no prominent peak in FTIR spectra from pure graphite powder. However, in case of oxidized or impure graphite, a number of infrared absorption bands assigned to C–OH (3400 cm$^{-1}$), C=O (1729 cm$^{-1}$), phenolic C–OH (1220 cm$^{-1}$), C–O (1052 cm$^{-1}$) and aromatic C–H (870 cm$^{-1}$) have been reported (Tan et al., 2013). Depending on the graphite C amount and transformation stage it is not clear until now, if they can be defined in soil samples.

Thermal / thermogravimetric analyses (TGA) have been applied for a long time to study the mineral components of soils and rocks. For instance, the Rock-Eval method has been developed for oil and gas exploration, whereby measuring the hydrocarbon, CO and CO$_2$ concentrations are measured during by a consecutive pyrolysis and oxidation program under
constant heating (Behar et al., 2001). More recent thermal analyses have been adopted to study the oxidative behaviour of soil OC, which might serve as a proxy for biogeochemical stability of these substances (Plant et al., 2009). The Rock-Eval method has been successfully applied to characterize the more stable part of OM remaining in the soil after long-term bare fallow (Barré et al., 2016). One of the advantages of TGA is the relatively inexpensive approach with minimal sample preparation needed to distinguish between different soil C components (Plant et al., 2009; Fernández et al., 2012; Kučerík et al., 2018).

Additionally, it is a promising method to differentiate between the thermally instable OM and highly stable geogenic C, like coal or graphite. A precondition for quantifying substances with the TGA method is that the thermal properties of the substance of interest are known, i.e. the temperature limits at which the oxidation/decomposition reactions take place. As no universally accepted temperature limits currently exists, the method still depends on empirically derived temperature boundaries to differentiate and quantify substances (Pallasser et al., 2013; Ussiri et al., 2014).

It has been demonstrated by Fernández et al. (2012) demonstrated that TGA coupled with differential scanning calorimetry and evolved gas analysis (CO₂/H₂O), increases the accuracy of quantifying organic substances during thermal analysis. Especially, in particular, the presence detection limit of CaCO₃ could be detected in smaller quantities as with decreased compared to conventional TGA since the decomposition could be masked by the dihydroxylation of (clay) minerals (Fernández et al., 2012). The same principle can also be expected for oxidation of graphitic C, as it takes place at roughly the same temperature ranges as the (clay) dihydroxylation and CaCO₃ decomposition (Hayhurst and Parmar, 1998; Bews et al., 2001).

Recently, a new method has been developed based on experience with TGA measurements, which is defined in the DIN 19539-standard (DIN Standards Committee Water Practice, 2016). In short, the DIN-standard defines biologically labile OM in solid samples, including soils, to be thermally oxidized at temperatures below 400°C (TOC₄₀₀), while residual oxidizable C (ROC), like lignite or soot, and inorganic C (TIC₉₀₀) are oxidized respectively decomposed between 400 and 900°C. Combustion elemental analysers, based on this DIN-standard, offer also the possibility to alternate between oxic and anoxic conditions during a measurement. In this method, dubbed named “smart combustion”, C components are consequently differentiated on both thermal and oxidizable properties. In theory, graphite, as a pure C, will oxidize poorly under anoxic conditions (Hayhurst and Parmar, 1998; Bews et al., 2001), while carbonates do not require oxygen to decompose at these temperatures. Contradictory to the pyrolyzing step of OM by the Rock-Eval method (Behar et al., 2001), OM is immediately oxidized in the first heating phase with the smart combustion method. Therefore, it is less likely that by-products of OM pyrolysis end up in the same fraction as graphitic C.

In summary, graphitic C content in soils has received very little attention as a quantification method is lacking. This study aims to test several available methods for identifying and quantifying graphitic C content of soils by examining Fourier Transform Infrared spectroscopy (FTIR), thermogravimetric analysis (TGA) and the smart combustion methods. To test the validity of the above methods for graphite identification and quantification, we analysed both natural and artificial soils that included widely present soil C components, i.e. carbonates and OM.
2 Material and Methods

2.1 Artificial mixtures and soil / rock samples

Top soil and fresh rock samples from a nearby outcrop were taken from a field site in Rambla Honda, Sierra de los Filabres (37°7'43" N, 2°22'30" W / Southern Spain). The area is located in the Nevado-Filabride complex and contains Devonian-Carboniferous slaty mica schist with graphite and garnets crossed by abundant quartz veins (Puigdefábregas et al., 1996). Carbonates found in the soil sample (0.18 % C) originated from pedogenesis and dust deposition, as the parent rock does not contain carbonates. **Soil material was taken from the topsoil (0-5 cm, without sieving crust) under the grass tussock Macrochloa tenacissima to ensure a substantial amount of OC was present. The soil itself was classified as Skeletic Leptosol (colluvic) according to the World Reference Base for Soil Resources (WRB, 2014).**

Additional soil material was collected from a field site near the town of Alboloduy (37°4’9” N, 2°36’43” W), hereafter referred to as AB soil, with similar vegetation and (climatic) conditions. The lithology consists of feldspathic mica schist (IGME, 1979), but without natural graphite and with a much higher CaCO$_3$ content (1.87 % C). **The AB soil, classified as Skeletic Leptosol (WRB, 2014), was also sampled from the topsoil, without sieving crust, under the grass tussock Macrochloa tenacissima.** The soil samples were dried at 40-°C and sieved to ≤ 2 mm.

Furthermore, three artificial soils were created, resembling a simplified version of a natural soil sample. For artificial soil 1 the organic horizon under deciduous tree species in Tharandter Wald (Saxony, Germany) was collected as OM substitute. Muscovite (American Educational, PN B00657LD62), a primary mineral present in the collected rock and soil samples (IGME, 1975), was taken as mineral component and ground in an agate disc mill. Together with CaCO$_3$ (Merck, Darmstadt, PN 1.02066) and graphite standard material (Merck, Darmstadt PN 1.04206), the components were mixed in a pure quartz matrix (Merck, Darmstadt, PN 1.07536) in the ratio of 10 % Muscovite, 2.4 % OM (=1.0 % C), 2.1 % CaCO$_3$ (=0.25 % C) and 0.5 % graphite (=0.5 % C). **Additionally, the individual components were mixed in quartz in the same concentration.** Artificial soil 2 was made without carbonate and Artificial soil 3 without graphite, whereby concentrations of the other components were kept the same constant (Table 1).

For testing and developing potential quantitative methods, the graphite standard was used as reference material and added to soil from Rambla Honda (hereafter denoted as calibration set 1) or with pure quartz (hereafter denoted calibration set 2) as a matrix in different quantities from 0.1 to 4 % (Table 1). All samples were ground in an agate disc mill (Retsch GmbH, Haan, Germany) in order to achieve homogenisation.

Total C content of the different mixtures and samples were measured and checked using an elemental CN analyser (Vario EL Cube, Elementar, Langenselbold, Germany). By acidifying samples, using an excess of HCl, the carbonates were removed. After drying at 60°C, the difference between acid and non-acid treatment was measured and called denoted as the total inorganic C (TIC). TC and TIC values of the used samples are summarized in Table 1. We assumed that with dry combustion at 950°C under pure oxygen atmosphere, all C holding components were decomposed or oxidized.
and therefore the total C content could be measured. This assumption was later validated by performing TGA temperatures up to 1100°C.

Raman spectra of the soil of calibration set 1, the standard (Merck) and graphitic schist were recorded, using a Thermo Scientific DXR Smart Raman Spectrometer, with 532 nm laser and a power output of 9 mW. Before the measurement, samples were pressed in aluminium cups. The obtained peaks were integrated using Lorentzian profiles fitting in Origin 2019.

The spectra of the graphite standard were highly similar to the graphitic schist (Fig. S1 in Supplement). The D1 (~1350 cm⁻¹), G (~1580 cm⁻¹) and D' (~1620 cm⁻¹) peaks could be attributed to graphitic C. The ratio between the D1 and the sum of all three peaks are clear indicators for the degree of graphitization (Beysacc et al., 2003; Ferrari, 2007). Ratios of 0.20 for the graphite standard (Merck) and 0.34 for the graphitic schist indicated well organized carbon (< 0.5, Beysacc et al., 2003) for both samples. Peaks around the 1200 and 1500 cm⁻¹ bands, which would have indicated the presence of pyrogenic / black carbon components (Sadezky et al., 2005; Schmidt et al., 2002) could not be observed in all samples including the soil sample (Fig. S1).

2.2 Fourier Transform Infrared (FTIR) spectroscopy

For transmission-FTIR analyses, 1 mg sample was mixed with 99 mg potassium bromide (KBr; Merck, Darmstadt; 3 sample replicates), finely ground in an agate mortar, and pressed to pellets. The transmission spectra were recorded in a Biorad FTS 135 spectrometer (BIO-RAD company, Cambridge, USA) as 16 co-added scans between wave number (WN) 4000 and 400 cm⁻¹ at a spectral resolution of 1 cm⁻¹. The spectra were corrected against ambient air as background and were converted to absorption units. For DRIFT analyses, ground mixtures of calibration set 1 and 2 were poured into standard cups (three replicates) without any dilution. The DRIFT spectra (16 co-added scans, WN 4000 and 400 cm⁻¹, resolution 4 cm⁻¹) were corrected for ambient air using a background spectrum of a gold target (99%; Infragold) and were converted to Kubelka–Munk units. All spectra were corrected for CO₂ absorption of the ambient air between WN 2400 and 2280 cm⁻¹ and smoothed (boxcar moving average algorithm, factor for transmission spectra: 25, factor for DRIFT spectra: 15) using the software WIN-IR Pro 3.4 (Digilab, MA, USA). For each sample one mean spectrum was calculated from the spectra of three replicate spectra.

2.2.1 Statistical analysis of the FTIR spectroscopic data

The partial least squares regression (PLSR) analyses of correlations between the transmission or DRIFT spectra and the graphite contents (0.1 - 4 %) of the samples were performed using R, Version 3.1.1 (R Core Team, 2014) with module PLS (SIMPLS, cross-validation: leave-one-out) of Mevik et al. (2018). The signal intensities were used as independent variables, the graphite content as dependent. The number of components used in the calibration models followed the lowest predicted root-mean-square error (RMSEP) of the specific datasets. The scores and loadings were plotted for the two main components determining most of the variances of the DRIFT spectra. Larger absolute loading values of signal intensities in certain WN regions imply a greater importance of these WN for the cumulated values of the principal component 1 or 2 displayed in the score plot.
2.3 Thermogravimetric analysis (TGA)

The TGAs were conducted on a STA 449 F5 Jupiter analyser (NETZSCH, Hanau, Germany). Therefore, 20-40 mg of sample material was placed in an Al₂O₃ crucible and heated under a constant heating program from ambient to 1100°C with a ramp of 20°C min⁻¹. First analyses were conducted under an oxygen-rich atmosphere, with an inflow of 250 ml O₂ min⁻¹ and 250 ml N₂ min⁻¹. Additional tests were done under anoxic conditions whereby the oxygen inflow was cut off between 500 and 850°C. The oxygen inflow was restored and the heating program continued until 1100°C.

As carbonates might interfere in the TGA measurement of graphite and high chloride concentrations damages the equipment, they were removed from the sample using the acid fumigation method of Harris et al. (2001). Briefly, about 40 mg of sample was weighed in a silver foil capsules, moistened to approximately field capacity and put in a desiccator under vacuum conditions with a beaker of 31% HCl and fumigated for 24 hours. Afterwards the sample was dried at 60°C overnight before it was transferred to an Al₂O₃ crucible for analysis.

2.3.1 Statistical analysis of the TGA method

TGA measurements were processed and thermal mass loss data obtained via the Proteus Thermal Analysis software (NETZSCH, Hanau, Germany). Initial testing with triplicates revealed that the repeatability error, expressed as coefficient of variation of the mass loss in areas of interest, was below 2%. Measured sample mass data are fitted with the spline function over the temperature, with steps of 1°C. Further analyses of the obtained data were conducted using R, Version 3.5.1 (R Core Team, 2018). Using the Fitting Linear Models function, models were created from the calibrations sets. With the module PLS (cross validation: leave one out) of (Mevik et al., 2018), a model was created for each calibration set to determine the graphite content based on the measured mass change in a certain temperature range and the known graphite addition, using the mass change as independent variable. By iterating the model creation over the temperature range from 400 to 1075°C with minimum step of 5°C difference and recording the slope, intercept and RMSEP of each created model, we evaluated the temperature range describing the graphite content of both the calibration sets most precisely. For this purpose, the RMSEP of these models were used (Fig. S4, see Supplement), and a single temperature range that fit both calibration sets was determined, following the lowest RMSEP of the specific datasets.

2.4 Smart combustion

Smart combustion denotes the method based on the DIN19539 (GS) standard (DIN Standards Committee Water Practice, 2016), whereby solid C components are separated based on their thermal and oxidizable properties. Smart combustion was conducted with the Soli-TOC cube analyser (Elementar, Langenselbold, Germany). The device is equipped with a non-dispersive infrared detector (NDIR), which measures the degree of infrared light absorbance caused by CO₂ concentration in the measuring gas (O₂/N₂). The NDIR has been calibrated with CaCO₃ and additionally CaCO₃ was used to control and
calculate a daily standard for the measured total C content measured. Depending on the expected carbonate and graphite concentration, 40-90 mg of homogenized sample was placed in the crucible. This was done to make sure that the peak surface fits well within the calibration range without causing unnecessarily large peak areas, which might influence the separation of the peaks / substances during analysis later on. Following the DIN19539 GS standard / standard gas switching program of the Soli-TOC cube analyser, the sample was first heated to and held at 400°C for 240 sec. whereby the “total organic carbon 400°C” (TOC400) was obtained. Subsequently the atmosphere was switched to inert gas (N2) and after an equilibration time of 100 sec. the sample was heated to 900°C and held for 150 sec. C released during this pyrolyzing phase is denoted as TIC, while it mainly consisting of carbonates, which do not need oxygen to decompose. After 150 sec., the oxygen gas flow was reintroduced and a third C component, the residual oxidizable C (ROC), was measured. It was hypothesized that this ROC fraction represents graphite.

2.4.1 Statistical analysis of the Smart Combustion data

The Soli-TOC device As the device directly converts the NDIR signal into the C content of the different components, as calibrated with CaCO3. Creating an additional model to correct the C output, introduces an additional error in the measurements. Therefore, we analysed the direct C output, as measured in the ROC fraction. Triplicate measurements were averaged, whereby the average coefficient of variation between replicates was 2.7%. A Pearson correlation test was performed between the obtained ROC data and calibration sets to evaluate how well the graphite content was measured.

3 Result

3.1 Overestimation of graphite contents by FTIR spectroscopy

The PLSR of the calibration set showed strong relations between the transmission-FTIR spectra from both calibration sets and the graphite concentrations when considering samples with 0.1 – 4 % graphite (Figs. 1a and 1b). For DRIFT spectra, the quality of these calibrations was at the same level (cal. Set 1: R² = 0.97, RMSEP = 0.16; cal. Set 2: R² = 0.98, RMSEP = 0.12). For calibration set 1 (based on natural soil) as well as for calibration set 2 (based on quartz), one main component of the PLSR presented most differences in the graphite concentration (Figs. 1c and 1d). This component showed the highest loading values across the entire range of wave numbers with some exceptions. For calibration set 1, wave numbers with decreased loading values were found at spectral regions 1077, 1031, 1013, 934, 913, 778, 536, 471, and 411 cm⁻¹, which all corresponded to the prominent absorption bands of the original soil used as matrix, comprising functional groups from organic and mineral matter (Hesse et al., 1984; Senesi et al., 2003; Van der Marel and Beutelspacher, 1976). For calibration set 2, the wave numbers with the smallest loadings at 1171, 1084, 796, 778, 694, 506, and 457 cm⁻¹ were exceptionally specific for quartz (Van der Marel and Beutelspacher, 1976).
The cross-validation of the PLSR calibration of both calibration sets 1 and 2 using spectra from the other calibration set, respectively, showed linear relationships. However, this relationship produced an overestimation in absolute graphite contents of ca. 2.59 % C for set 1 and 1.87 % C for set 2 (Fig. 5). The graphite content of 0.50 % for the artificial soil was 3.5-times underestimated by the PLSR calibration using set 1 (predicted: 1.75 % C) and 3-times underestimated by calibration set 2 (predicted: 0.17 % C). The graphite contents in the graphitic schist was estimated to be 1.91 % C by calibration set 1 and 3.71 % C by calibration set 2, which was respectively 2-times and 4.5-times higher than the total C content of the graphitic schist (Table 1).

3.2 Strong matrix dependency of TGA predictability

First qualitative TGA results revealed overlapping mass loss peaks of graphite and CaCO₃ (Fig. 2). Between 750 and 850°C, the sum of the mass losses of the individual components was smaller than the mass loss of the mixture of these components. Using the RMSEP values of the two calibration sets of soil and quartz with graphite, the most useful temperature range for modelling graphite content was identified. Based on the RMSEP values, the best range for modelling graphite content by mass loss was identified between 680°C and 840°C (Fig. S4, in Supplement), as visualized in Supplementary 3. According to PLSR, for both models, created with calibration set 1 (R² = 1.00, RMSEP = 0.05) and calibration set 2 (R² = 1.00, RMSEP = 0.04), predicted graphite content in their respective matrix was quite good (Fig. 3). Only cross-validation by predicting graphite content in the other matrix revealed a short coming of the TGA method (Fig. 5). The slope between predicted and actual graphite content is still parallel to the 1:1 line, but applying the model derived from calibration set 1 on set 2 underestimated the graphite content by 1.81 % C (Fig. 5). In contrast, the graphite content in calibration set 1 was underestimated by 1.81 % C using the model calibrated with set 2.

The graphite content of the artificial soil (0.5 %) and graphitic schist were estimated using the two calibrated models (Inset Fig. 5, inserted detail and Table 2). The content of artificial soil 1 was overestimated 3.5-times (predicted: 1.70 % C), using the model derived from calibration set 2. Using the model based on calibration set 1, the prediction of graphite content yielded negative values (predicted: -0.19 % C). In the additionally used What further stands out in Table 2 are the artificial soils 2 and 3, which were without did not contain respectively CaCO₃ and Graphite, the graphite contents – Both were underestimated with the model based on calibration set 1 (based on soil) and overestimated with the model based on calibration set 2 (based on quartz). Independent from the two models, the relative difference between them is in both cases 0.44, which resembles the actual graphite content. The graphite content of the graphitic schist was estimated to be 2.44 % C according to calibration set 2, but the model calibrated with set 1 showed better predictions. Where the total C content of the graphitic rock was 0.84 % C, the model of calibration set 1 estimated a graphite content of 0.64 % C (Table 2).

Furthermore, the artificial soils were used to explore whether changing between oxic and anoxic conditions during thermal analysis could separate between the mass loss peak of CaCO₃ and graphite – most important for potential application in soils containing both components. Changing the atmospheric composition resulted in an artificial mass gain when O₂/N₂ gas was switched to N₂ and an artificial mass loss when oxygen was reintroduced (Fig. 4), probably due to changes in pressure,
and thereby affecting the mass readings. It should be noted that a constant heating program was used and therefore the measurement time spans 4-5 minutes for both peaks. Nonetheless using artificial soil 2 (without CaCO₃) and artificial soil 3 (without graphite) revealed that qualitative separation between the two mass loss peaks was feasible using changes in oxic conditions. Based on TGA observations on some individual components and simplified artificial soil (Fig. 2), it seemed best to use anoxic conditions from 500 till 850°C as all OM will be oxidized at 500°C and carbonates should be fully decomposed at 850°C (Fig. 2). Furthermore, the mass loss peak had already returned to the baseline after reaching 850°C, indicating that the decomposition of the carbonates was completed (Fig. 4).

3.3 Direct graphite content quantified by smart combustion

First total carbon (TC) measured by the smart combustion method was compared with the TC obtained by dry combustion using the elemental analyser, but hardly any differences were found (Fig. S5 in the Supplement). Residual oxidizable C (ROC) values obtained by the smart combustion method were plotted against the added amount of graphite in calibration sets (Fig. 5). The graphite content in calibration set 1 seems to be overestimated by 0.26 % C. This observation can be explained by the fact that the used sample soil for calibration set 1 contains an unknown amount of natural graphite, which explains a constant overestimation. The content of graphitic C of the soils in calibration set 2 is slightly underestimated, especially with higher graphite concentrations (Fig. 5 and Table 2).

It can be seen in Table 2 that the graphite content of the artificial soil 1 was slightly underestimated (Table 2): 0.40 % ROC for the artificial soil 1 and 0.46 % ROC for the artificial soil 2, where it should have been compared to the real contents of 0.50 %. Artificial soil 3 revealed measured 0.00 % ROC as there was also no graphite in this sample. The graphitic schist had 0.79 % ROC, which was very similar to the total C of the rock (Table 1 and 2). Furthermore, the AB soil spiked with graphite, as an additional sample without natural graphite but with a high CaCO₃ content (1.87 % C), the AB soil spiked with graphite, showed a similar underestimation as observed with the calibration samples (Fig. 5).

The ROC results for the natural graphitic soil, also used for calibration set 1 and was also spiked with CaCO₃ revealed a downward trend of measured ROC content with increasing amounts of added CaCO₃ (from 0.0 to 2.5% added C) down to 0.01 % ROC absolute (Fig. 6) This observation was also evident when taking into account the, for which the ROC results are given in Fig. 6. Although, dilution of the sample by the addition of CaCO₃ was taken into account, a downward trend of measured ROC content of 0.01 % ROC absolute was observed with increasing amounts of added CaCO₃ (from 0.0 to 2.5% added C) and explained at least parts of the underestimation of graphite in this sample (Fig. 5).

4 Discussion

4.1 Matrix effects and the lack in specific absorption bands hamper graphite quantification via FTIR spectroscopy

The calibration between infrared spectra and graphite contents of the calibration sets yielded promising results (Figs. 1a and 1b) and could also be used for a cross-validation (Fig. 5). However, these validations showed a systematic overestimation of
graphite contents for both calibrations and the inability of these calibration sets for predicting the graphite content of artificial soil 1. Although the same substrate materials and similar contents of graphitic C were used in the validation, the graphite contents were systematically over-predicted. Despite the apparent quality of the calibration, this failure could have been caused by the relatively low number of calibration samples. Note that the use of the two calibration data sets, soil and quartz, in a joint PLSR model ($R^2 = 0.96$ and an RMSEP = 0.24; 3 components) did not improve the calibration nor the prediction accuracy. It cannot be excluded that a higher number of samples for the calibration could improve the PLSR model and the prediction results. Further, Raman spectroscopy might be an alternative approach for quantifying graphite in soil samples (e.g., Sparkes et al., 2013; Jorio and Filho, 2016).

The loading values revealed that the relationships used for the PLSR calibration (Figs. 1c and 1d) could not be attributed to absorption bands of graphite reported in literature such as wave numbers 2200, 1587, 1362, and 830 cm$^{-1}$ (Friedel and Carson, 1971) or 3400, 1729, 1220, 1052 and 870 cm$^{-1}$ (Tan et al., 2013). These wave number positions did not match with the absorption bands of the spectra obtained from calibration set 1 and 2, the graphitic schist and the artificial soil 1 (Fig. 7). The only exception is WN 3400 cm$^{-1}$, which is mainly caused by O–H of free or adsorbed water and is thus not specific for graphite.

The height of the loading values across broad spectral regions, i.e. across the entire wave number range in case of the soil samples, hints on effects of the general optical conditions within the samples. The transmission, i.e. the energy throughput in the sample pellet (transmission FTIR) or the reflectance of the sample surface (DRIFT), seems to be a measure for the amount of graphite added rather than specific graphite signal intensities in the calibration sets. In addition, but transmission/reflection characteristics are highly influenced by the mineral composition of a sample. Generally, increasing graphite concentrations caused decreasing transmission over the entire spectral range, which is a characteristic of the mineral composition due to increasing proportions of primary absorption (Kortüm, 1969; Hesse et al., 1984). This assumption is underlined by the fact that calibration was not possible with both calibration sets when using baseline-corrected spectra, because baseline correction compensated the described effect of decreasing transmission over the entire wave number range with increasing graphite content. In addition, DRIFT measurements of pure, i.e. undiluted, graphite material (not shown) did not reveal any prominent absorption bands. Note that specific graphite absorption bands that have been reported in the literature are only valid for oxidized graphite where C–O and C=O groups have been formed to a certain extent (Tan et al., 2013). While FTIR spectroscopy may be feasible for determining oxidized or impure graphite, it was found not useful in our case, where an oxidization or impureness of the used graphite material obviously did not appear. Further, the mentioned potential signal intensities would occur in the same spectral ranges as compared to signals from SOM functional groups, thus hampering a quantification of graphite in soil samples. Consequently, the lack of specific absorption bands resulted in a strong dependency of the calibration and validation quality on the sample matrix, i.e. its main mineral component. This matrix effect was illustrated by the incapability of the PLSR models to predict the graphite contents of the artificial soil 1 or graphitic schist (Fig. 5 and Table 2).
4.2 Strong matrix effects did not allow using TGA as a universal quantification method

In recent work, the TGA method has been tested and further developed for differentiating between carbonates and OC / OM (e.g. Apesteguia et al., 2018). First analysis of individual components in a quartz matrix revealed that graphite has a similar thermal stability as carbonates (Fig. 2) and overlap in the thermal region where dehydroxylation of various minerals takes place (Földvári, 2011; Fernández et al., 2012). As shown in Fig. S2 and S3 in the Supplement supp. 4 and discussed by other studies, complete removal of carbonates from soil sample by acid fumigation is difficult and might affect further quantitative estimations. For graphite, estimation as sought for graphite, via TGA becomes challenging as the acid affects the thermal stability of other soil constituents and makes the sample hysgrosopic (Agarwal and Bucheli, 2011; Apesteguia et al., 2018). Additionally, sample grinding in an agate disk mill, representing common homogenization process used for small sample amounts (10’s of mg), introduce some changes in thermogravimetric patterns for some minerals, e.g. micas, but makes it also more “reactive” (Földvári, 2011). This would mean that mass loss peaks for minerals, like the used muscovite, can appear sharper and at lower temperatures, in comparison with non-ground materials.

The best temperature ranges to relate mass loss to the amount of added graphite was between 680°C and 840°C. For calibration set 2 (pure quartz matrix) a lower temperature (range) would also be able to predict the graphite content (Supp. 3 Fig. S4 in Supplement), which indicates an indicating interferences in the soil matrix of calibration set 1. The best temperature range was in line with the observation that the mass loss peak of graphite spans a large range (Fig. 2), most likely a result of the slow oxidation of this pure C. Other studies found that graphite in a (fluid) sand bed already oxidized slowly under oxygen rich conditions at temperatures below 670°C accelerating at higher temperatures (Hayhurst et al., 1998; Bews et al., 2001).

Validation of the created models from the two calibration sets revealed that interference with other soil components required an individual calibration for every sample set of specific (mineral) composition (Fig. 5). As shown in this study, fresh but ground muscovite dehydroxylates between 600 and 1000°C (Fig. 2), which influenced the total mass loss measured in this temperature range. Other present (minor) minerals, like chlorites (500-860°C) or apatites (200-1400°C), might also increase the bias by influencing mass loss (Földvári, 2011; Tõnsuaadu et al., 2011). This observation could explain why the model of calibration set 1, using the soil spiked with graphite, showed a good predictability of the graphite content in the graphitic schist, as mineral composition is highly similar between these two samples (Fig. 5 and Table 2).

Roth et al. (2012) suggested that the use of anoxic conditions / a pyrolyzing phase during measurement might be useful to differentiate between wood and black C. According to our gas switching experiment with the TGA, it is at least a useful approach to differentiate between graphite and CaCO₃ (Fig. 4). Due to the artificial mass gain/loss induced by switching the gases during the measurement, exact temperature ranges for developing a quantitative method could not be established. As no universally-accepted temperature limits for the quantification of TOC, TIC or other carbonaceous substances exist, the best temperature ranges for switching between oxic and anoxic conditions are difficult to define (Pallasser et al., 2013; Ussiri et al., 2014). For instance, according to the DIN19539 - standard, TOC is defined as the oxidizable C at maximum of 400°C. Others showed that 1/5 or even 1/3 of the TOC is not oxidized at 400°C (Pallasser et al., 2013; Schiedung et al., 2017). For
the artificial soil in our study a temperature limit of 400-500°C seems to be too low to oxidize all OM, as indicated by the TGA in Fig. 2, and therefore the pyrolyzing phase was set to 500°C. To obtain a clear peak for the graphite oxidation, it is important that the other substances, i.e. CaCO₃, are already decomposed. In the case of the artificial soil, it was found that at 850°C all CaCO₃ was decomposed and a clear peak for graphite was formed upon re-establishing the oxic conditions (Fig. 4). With higher CaCO₃ levels or dolomitic carbonates, a higher temperature might be needed to create a clear separation between the substances (Földvári, 2011).

TGA seemed to be a good method to identify different organic components of samples and thus can be used as complemenetal technique to other methods for (organic) C content estimation. For high graphite content with negligible amounts of dehydroxylating minerals and/or decomposing carbonates, TGA might be a useful method to quantify graphite.

4.3 Minor effect by CaCO₃ and radicals on direct graphite quantification using smart combustion

With the TGA method it was already shown that qualitative differentiation between carbonates and graphite was possible by changing between oxic and anoxic conditions during heating of the sample (Fig. 4). Using the Soli-TOC device, a direct measurement of the released C could be achieved during the heating/gas changing program, which correspond very closely to the amount of (added) graphite (Fig. 5 and Table 2). The fact that the Soli-TOC device measured almost the same TC values as the elemental analyser (Fig. S5 in Supplement), supported the idea that a direct comparison between the ROC fraction and (added) graphite content is possible.

As shown by Hayhurst and Parmar (1998), very small impurities in the graphite can cause a small part of the graphite to pyrolyse during anoxic conditions at higher temperatures. **Graphitic C of lesser graphitization might therefore result in a larger loss of graphitic C during pyrolysis and a greater underestimation of the graphitic C content.** Taking a closer look at the measurements of the artificial soil, reveals that a small part of the graphite started to oxidize under anoxic conditions (Fig. 8). **The measurement of graphite in quartz, as done for calibration set 2, showed that about 6% of the total C was lost during the pyrolysis phase, while for the graphitic schist this loss was 2% (Fig. 8c), resulting in a small underestimation of the graphitic C content.** Bews et al. (2001) suggested that at temperatures higher than 700°C, radicals like HO₂ and OH might act as reactant with the pure C. Furthermore, in the method comparison study for recovering different black C types, Roth et al. (2012) suggested a (relatively) strong catalytic effect of oxides on black C oxidation, which was most predominant in soils. These ideas are also supported by our observation that artificial soil 2 (without CaCO₃) measured higher ROC values (0.06% more C absolute) than artificial soil 1 (with CaCO₃, Table2). Also in the carbonate-rich AB soil the added graphite was underestimated by 7% (Fig. 5). Furthermore, graphitic C was underestimated with increasing CaCO₃ content (Fig. 6). The 7% underestimation by the AB soil, which contains 1.87 % C-CaCO₃, coincided with ROC underestimation of the calibration soil with 2 %C-CaCO₃ addition.

When the soil contains more thermally resistant OM, which is not oxidizable at 400°C and can be 1/4th of the OM (Schiedung et al., 2017), the question rises if this fraction is pyrolyzed with during heating under anoxic conditions or if it is taken as part of the ROC fraction when oxygen is again available. The TGA method showed that not all OM has been oxidized
at 400°C (Fig. 2). Taking a closer look on the smart combustion measurement of the artificial soil 2 and only particularly its fresh OM component (Fig. 8), revealed a small peak formed upon heating the sample above 400°C. This peak represents, which is only a few percent of the total OC, but a clearly indication that not 100% of the OM was oxidized incompletely at 400°C. Although this study focuses on the ROC component, this point it might be important when considering the TOC and TIC fractions of the smart combustion method.

When a sample contains other forms of thermally resistant OM or even black C, which are not pyrolyzed during the anoxic phase, this C component is likely to end up in the graphitic C fraction with the smart combustion method. Especially as most temperature boundaries are empirically derived (Pallasser et al., 2013; Ussiri et al., 2014), a pre-test with continuous heating under oxic conditions, is therefore recommended to assess the number of C-containing substances present in the sample by the occurrence of peaks. Further studies should focus on temperature boundaries of different substances in relation to their properties and see, for instance, how graphitic C can be distinguished from other thermally stable C components.

Although the thermal boundaries for the different C fractions are given in the DIN19539 (GS) standard are debatable (Ussiri et al., 2014; Schiedung et al., 2017), we showed that the ROC fraction corresponded closely to the graphitic C content. Through the smart combustion method, graphitic C could be differentiated from the other C components in soil matrix and quantified satisfactorily as indicated by the offset in the calibration with graphite estimation in the artificial/spiked samples.

4.4 Potential for combining methods

Comparing the ability of the examined methods on predicting graphite content, it becomes clear that FTIR overestimated, TGA was highly variable, and smart combustion was most accurate in predicting the graphite content (Fig. 5). An interesting observation was the similar predicted graphite content in calibration set 1 by both the FTIR and TGA methods, especially as FTIR is based on spectral properties and TGA on the thermal stability of the graphite. It has previously been suggested to combine FTIR and TGA systems to rapidly characterize the soil OM (Demyan et al., 2013). Oxidation of graphite upon heating could result in specific infrared absorption bands (Tan et al., 2013), which nevertheless would still be superimposed by SOM-specific bands in natural soil samples. As discussed by Demyan et al. (2013), not only the available oxygen, but also the heating rate has an important effect on the charring of OM and thereby on the thermal and spectral properties of the studied material.

As Raman spectroscopy is suitable to distinguish graphitic C and determine its degree of graphitization, it seems to be a promising method. Nonetheless, to use Raman spectroscopy for quantification of substances in a soil matrix, further studies should first focus on standardization of sample preparation, as it has a large influence on the measured intensities and baseline determination and thereby the direct quantification of components (Beysace and Lazzeri, 2012; Sparkes et al., 2013).

Although Because we focused in this study on the ROC component, which significantly correlated with the graphite content, a consideration of other components mentioned in the DIN19539-standard was beyond the scope of this study. Nonetheless, we found indications that the thermal boundaries defined in the DIN19539-standard are not ideal to differentiate between soil OM and inorganic C (Fig. 8). As most carbonates start to decompose at temperatures of 550°C (Földvári, 2011), it might be more suitable to increase the level for the TOC component from 400 to 500°C. Only when black C, which might
oxidize between 375 and 540°C (Roth et al., 2012), is present in the sample, this might lead to an overestimation of the TOC content. Using TGA simultaneously with differential scanning calorimetry, water and CO₂/H₂O flux measurements (i.e. evolved gas analysis, as previously suggested by Fernández et al. (2012) or with the Rock-Eval method focusing on hydrocarbon, CO₂ and CO release (Behar et al., 2001), could improve the development of a more standardized method applicable to soils using combustion elemental analysers. The overlap between the thermal properties of different C components emphasizes the need to always first consider what is present in the potential as well as possible interferences with the sample and what might interfere with the considered applied methodology, before applying a fast and standardized analytic method.

5 Conclusion

Three widely used methods were examined for their potential to quantify graphitic C content in soil samples. Calibrations between mid-infrared transmission as well as DRIFT spectra and graphite contents of well-defined samples are principally possible via PLSR. However, these calibrations depend on general effects of graphite contents on the energy transmitted through the samples rather than on signal intensities of specific graphite absorption bands. The use of samples from different origins yield strong matrix effects and hampers the prediction of geogenic graphite contents in soils. Thermogravimetric analysis of the samples revealed that the suitability of this it is a useful qualitative method for identifying graphitic C in soil samples, although care should be taken for carbonates as they have a similar thermal stability. Quantitative estimation of the graphite content seems challenging as dihydroxylation of several soil minerals occur at similar temperatures, requiring the calibration with an empirical model necessary. With alteration between oxic and anoxic conditions during heating of a sample, a differentiation between other soil components and graphite could be established, like applied in the using smart combustion method. Further quantification of the released C during the gas changing heating program revealed a close correspondence between the measured ROC and original graphite content. Among of the examined methods, the smart combustion method performs best in differentiating between graphite and other soil components and thereby also in quantifying graphitic C in soil samples.

Author contribution

ML contributed the FTIR part to the article, including analysis of the data, and editing the paper. CV contributed by suggesting and evaluating test set-up and by commenting on and editing the paper. SS contributed by providing measurements with the Soli-TOC device, technical assistance and editing the paper. KK contributed by suggesting and evaluating test set-up and by commenting on and editing the paper.
Competing interests

The authors declare that they have no conflict of interest.

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References


Figure captions

Figure 1: Prediction plots with 95% prediction bands (a, b) and loading plots (c, d) after FTIR analyses of the PLSR calibration sets using soil (calibration set 1) and pure quartz (calibration set 2) with graphite concentrations of 0.1 – 4 % added as indicated in table 1.

Figure 2: Thermogravimetric analysis of artificial soil 1 and its components measured individually. The summation (dash-dotted, grey) is the combined mass loss of the individual components.

Figure 3: Prediction plots after thermogravimetric analysis (TGA) of calibration set 1 (squares) and 2 (diamonds). 95% predictions band (dotted grey) are displayed besides the linear regression line (black). Graphite was added in the concentrations between 0% and 4.0% graphite as indicated in table 1.

Figure 4: Thermogravimetric analysis of the artificial soils, with one lacking either carbonate (Artificial soil 2, green) or graphite (Artificial soil 3, orange), whereby the oxygen gas supply was cut off during part of the standard heating program (GC – Gas Change). For comparison the artificial soil 1 under normal program (without gas change) is also displayed in grey.

Figure 5: Overview of the predicted amount of graphite in the calibration sets (squares/diamonds), artificial soil (inset, circles/triangle), graphitic schist (inset, stars) and AB soil (right pointing triangle) as measured with the different methods Black symbols: graphite prediction by FTIR, model from calibration set 1; White: graphite prediction with FTIR, calibration set 2; Orange: graphite prediction by TGA, model from calibration set 1; Green: graphite prediction by TGA, model from calibration set 2; Grey: graphite prediction by smart combustion. - Exact data is given in Table 2.

Figure 6: Residual oxidizable carbon (ROC) as measured with the smart combustion method plotted against the added CaCO3 content to soil sample, used for creating calibration set 1.

Figure 7: Transmission spectra of the pure quartz (Calibration set 2, sample 1), quartz + 4 % graphite (Calibration set 2, sample 10), soil + 4 % graphite (Calibration set 1, sample 10), artificial soil 1 (0.5 % graphite added) and graphitic schist. The vertical lines denote wave numbers for which absorption peaks have been reported in literature (see text).

Figure 8: Examples of smart combustion measurements of the artificial soils (a) and the fresh OM component (b), the graphite standard and graphitic schist (c). The blue area delineates the interval where O2 gas is substituted for N2 and the temperature program is displayed by the red dashed line. Note that artificial soil 2 (green) is without CaCO3 and artificial soil 3 (orange) is without graphite.
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Figure 8: Examples of smart combustion measurements of the artificial soils (a) and the fresh OM component (b), the graphite standard and graphitic schist (c). The blue area delineates the interval where O$_2$ gas is substituted for N$_2$. The temperature program is displayed by the red dashed line. Note that artificial soil 2 (green) is without CaCO$_3$ and artificial soil 3 (orange) is without graphite.
Table 1: Overview of the used samples. Total Carbon (TC) and Total Inorganic Carbon (TIC) data is derived by the elemental analyzer, as described in section 2.1. Crosses note for which method testing the sample has been used.

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<th>Sample</th>
<th>TC [% C]</th>
<th>TIC [% C]</th>
<th>Added Graphite [% C]</th>
<th>FTIR</th>
<th>TGA</th>
<th>Smart combustion</th>
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* -contain unknown amount of natural graphite // # -as measured in the sample with 0.0 added graphite
Table 2: Overview of the predicted graphitic carbon by the different examined methods.

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<th>Sample</th>
<th>Added Graphite</th>
<th>FTIR Calibrated with Cal. Set 1 [% C]</th>
<th>FTIR Calibrated with Cal. Set 2 [% C]</th>
<th>TGA Calibrated with Cal. Set 1 [% C]</th>
<th>TGA Calibrated with Cal. Set 2 [% C]</th>
<th>Smart combustion [% C]</th>
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* -containing unknown amount of natural graphite