Soil COS exchange: a comparison of three European ecosystems

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Key Points:
• COS emissions increased after UV exposure of soil samples from three European soils
• Primary abiotic driver of COS cycling differs between three European soils

Supporting Information:
• Supporting Information S1

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Abstract The potential of carbonyl sulfide (COS) flux measurements as an additional constraint for estimating the gross primary production depends, among other preconditions, on our understanding of the soil COS exchange and its contribution to the overall net ecosystem COS flux. We conducted soil chamber measurements of COS, with transparent chambers, in three different ecosystems across Europe. The in situ measurements were followed by laboratory measurements of soil samples collected at the study sites. The soil samples were exposed to UV radiation to investigate the role of photo-degradation for COS exchange. In situ and laboratory measurements revealed pronounced intersite and intrasite variability of COS exchange. In situ COS fluxes were primarily governed by radiation in the savannah-like grassland (SAV), soil temperature and intrasite heterogeneity in the deciduous broadleaf forest, and soil water content and intrasite heterogeneity in the evergreen needleleaf forest. The soil of the ecosystem with the highest light intensity incident on the soil surface, SAV, was a net source for COS, while the soils of the other two ecosystems were COS sinks. UV radiation increased COS emissions and/or reduced COS uptake from all soil samples under laboratory conditions. The impact of UV on the COS flux differed between soil samples, with a tendency toward a stronger response of the COS flux to UV radiation exposure in samples with higher soil organic matter content. Our results emphasize the importance of photo-degradation for the soil COS flux and stress the substantial spatial variability of soil COS exchange in ecosystems.

1. Introduction

Measurements of the trace gas carbonyl sulfide (COS) have been discussed as a promising tool for inferring gross primary production (GPP) and stomatal conductance at the ecosystem scale (Sandoval-Soto et al., 2005; Seibt et al., 2010; Wehr et al., 2017; Whelan et al., 2018; Wohlfahrt et al., 2012). COS sparked the interest of researchers because its uptake by plants is closely linked to the uptake of CO2 (Seibt et al., 2010). Both gases enter the leaf via similar pathways and are subsequently consumed by the same enzymes, especially the enzyme carbonic anhydrase (CA) (Seibt et al., 2010; Stimler et al., 2010). So far, no COS emissions from vascular plants have been reported. This would in theory and in the absence of other ecosystem sources and sinks equate the measured COS flux at the ecosystem level to the COS uptake by plants, which could be used to infer the CO2 plant uptake. COS and CO2 measurements together, both on the ecosystem level, could therefore improve the partitioning of the net ecosystem CO2 flux into GPP and respiration.

If COS uptake is indeed primarily driven by plant uptake, it might be a useful GPP proxy. Other sinks and sources of COS, apart from plants, should then either be negligible or easy to parameterize. Soil is potentially the largest nonleaf contributor to ecosystem COS uptake because it contains a multitude of organisms equipped with CA (Conrad, 1996; Meredith et al., 2019; Ogawa et al., 2016; Seibt et al., 2006; Wingate et al., 2008). In addition to CA, other enzymes were found to hydrolyze COS, among them RubisCO (Lorimer & Pierce, 1989), nitrogenase (Seefeldt et al., 1995), CO dehydrogenase (Ensign, 1995), and CS2 hydrolyase (Smeulders et al., 2013). Substantial COS consumption by CA in soils has been shown by laboratory experiments (Kesselmeier et al., 1999; Meredith et al., 2019). Together with in situ measurements of soil COS fluxes (Steinbacher et al., 2004; Yi et al., 2007), these findings underline the role of soil as a substantial COS sink. Attempts to close the COS budget on the global scale characterized oxic soils as COS sinks (Berry...
et al., 2013; Kettle et al., 2002; Launois et al., 2015; Montzka et al., 2007). Recent mechanistic soil COS flux models (Ogée et al., 2015; Sun et al., 2015) led to similar results, consolidating the role of oxic soils as a net COS sink. But there is also evidence for oxic soils acting, under hot, dry, and/or high-light conditions, as a net source for COS (Florian Kitz et al., 2017; Maseyk et al., 2014; Meredith et al., 2018; Whelan & Rhew, 2015), presumably driven by abiotic production of COS from precursors of biotic origin (Meredith et al., 2018). The motivation to understand how soil COS fluxes vary with soil moisture, soil temperature, and radiation is therefore growing, since more knowledge is needed to adequately represent the soil COS exchange in models (Whelan et al., 2018). Advances have been made in understanding the impact of soil moisture on the soil COS flux by recent laboratory studies (Bunk et al., 2017; Kaisermann, Jones, et al., 2018), which investigated soil samples from different ecosystems, concluding that COS production increases under drier conditions. Bunk et al. (2017) showed that soil samples can act as both a net COS sink and source depending on their soil moisture. Knowledge about the light contribution to COS emissions from soil is much scarcer, with only few studies (Meredith et al., 2018; Whelan & Rhew, 2015) investigating the effect of radiation under controlled conditions in the laboratory. Data for in situ measurements examining the link between radiation and soil COS exchange is equally scarce (Florian Kitz et al., 2017), since most studies used opaque chambers (Berkelhammer et al., 2014; Maseyk et al., 2014; Sun et al., 2016), therefore excluding any radiation effect on decomposition (Yanni et al., 2015). From studies investigating the ocean (von Hobe, 2003), a lake (Du et al., 2017), and precipitation (Mu et al., 2004) we know that especially the energy-rich UV fraction of the solar spectrum is involved in COS production. COS production in seawater was the highest at shorter wavelengths and decreased with increasing wavelengths (in an investigated range of 280–380 nm) (Zepp & Andreae, 1994).

The magnitude and sign of soil COS fluxes varies between ecosystems (Liu et al., 2010; Maseyk et al., 2014; Yi et al., 2007), and if our goal is a robust global estimate of the soil COS flux, we need more data from different soils. But processes in soils, for example, respiration, can vary greatly within ecosystems (Li et al., 2017; Wang et al., 2006) depending on the heterogeneity of the system, which should incentivize scientists to also cover the intrasite heterogeneity of soil COS fluxes.

We investigated three different ecosystems across Europe to address shortcomings in our current knowledge on soil COS fluxes (Whelan et al., 2018), especially the importance of abiotic versus biotic controls and the magnitude of small-scale heterogeneity. We conducted in situ measurements in an oak savannah-like grassland (SAV) in Spain, a deciduous broadleaf forest (DBF) in Denmark, and an evergreen needleleaf forest (ENF) in Estonia with transparent soil chambers. The in situ setup allowed us to address the following hypotheses: soil COS fluxes (I) are primarily driven by differences in radiation; (II) vary in response to other abiotic and biotic site characteristics such as different light, temperature, and soil moisture conditions; and (III) can change with small-scale variations in abiotic and biotic conditions within a site. Soil samples were collected from the investigated sites in addition to the in situ measurements and were compared to each other in the dark and under UV-A exposure under controlled conditions in the laboratory. This approach allowed for “real-world” data, that is, from an intact soil column and the full set of environmental parameters, to be combined with data derived from a controlled system, that is, the laboratory.

2. Material and Methods

2.1. Sites

2.1.1. Savannah-Like Grassland

The FLUXNET site (Site-ID: ES-LMa) Las Majadas del Tietar (39°56′29.4″N, 5°46′24.1″W) is located in the province of Cáceres in Spain at an altitude of 260 m above sea level. The climate is Mediterranean with a mean annual temperature of 16 °C and annual rainfall of 700 mm (Perez-Priego et al., 2017). The vegetation is described as a holm oak savannah-like grassland, grazed by cows, with tree canopy coverage of about 20%. The dominant tree species is Quercus ilex, the grass layer is dominated by Vulpia bromoides, Vulpia geniculata, Trifolium subterraneum, and Ornithopus compressus. The soil was classified as Abruptic Luvisol (IUSS Working Group WRB, 2006). For more details about the study site, see Lopez-Sangil et al. (2011) and Weiner et al. (2018).

2.1.2. Deciduous Broadleaf Forest

The FLUXNET site (Site-ID: DK-Sor) Soro (55°29′9.24″N, 11°38′40.56″E) is located on the Danish Island of Zealand at an altitude of 40 m above sea level. The climate is temperate maritime with a mean annual...
COS and CO₂ concentrations were measured with a Quantum Cascade Laser (QCL) Mini Monitor. In the forest, microrelief of the forest floor has a distinct structure, with elevations, which consist of decomposing tree stubs and litter, and depressions, lower, typically very wet, areas. For more details about the study site, see Pilegaard et al. (2011) and Wu et al. (2013).

2.1.3. Evergreen Needleleaf Forest

The SMEAR site Järseljä (58°16′39.416″N, 27°18′29.637″E) is located in southeastern Estonia near Lake Peipsi. The climate is temperate with a mean annual temperature of 4–6 °C and annual precipitation of 500–750 mm. The site is located in a hemiboreal forest dominated by Norway spruce (Picea abies). The ground vegetation consists of Oxalis acetosella, Vaccinium myrtillus, Calamagrostis arundinacea, Convallaria majalis, Melampyrum pratense, and several moss species (at especially moist location even Sphagnum spp.). The soil was classified as Haplic Gleysol (eutric) with a thick humus horizon. The forest floor has a distinct structure, with elevations, which consist of decomposing tree stubs and litter, and depressions, lower, typically very wet, areas. For more details about the study site, see Noe et al. (2011, 2015).

2.2. In Situ Measurements

In situ measurements were done in May (SAV), June (DBF), and August (ENF) of 2016. At each site, 6–9 stainless steel (SAE grade: 316 L) rings were inserted 5 cm into the soil and remained in place for the whole month. The aboveground vegetation, if any, was removed. Some small patches of moss remained inside the ring at ENF, because removing these was not possible without a severe disturbance of the top surface layer. Litter was not removed from the soil surface in the chambers, since its contribution to the COS flux was found to be significant (Sun et al., 2016) and is often inseparable from the soil under natural conditions. Since litter (in the DBF and ENF) and a little bit of moss (in the ENF) was inside the rings, the term soil includes moss and litter typical for each site with regard to the in situ measurements to avoid introducing a new term. We attempted to cover the heterogeneity of the ecosystems by placing rings in different locations, which had distinct features as explained below. In SAV rings were placed under a tree (T), in the open grassland (O), and in between (I). In the DBF measurements were performed with chambers containing litter and chambers without litter, but only the chambers with litter were used in the analyses (to see how the rings with litter compare to the rings without litter, see supporting information Figure 1). In DBF rings were placed in areas with low (LV), medium (MV), and high (HV) surrounding ground vegetation. In ENF rings were placed on top of elevations (E) and in depressions (D) forming the characteristic microrelief of the forest floor at the site.

COS and CO₂ concentrations were measured with a Quantum Cascade Laser (QCL) Mini Monitor (Aerodyne Research, Billerica, MA, USA) at a wave number of ca. 2056 cm⁻¹. The QCL was operated at a pressure of 20 Torr using a built-in pressure controller and temperature of the optical bench and housing controlled to 35 °C. Fitting of absorption spectra at 1 Hz, storing of calculated mole fractions, and switching of zero/calibration valves, control of pressure lock, and other system controls were done by the TDLWintel software (Aerodyne Research, Billerica, MA, USA). The QCL was housed in a temperature-regulated custom-built case (DE Casebuilder.com GmbH, Hamburg, Germany). A handheld sensor (WET-2, Delta-T Devices, Cambridge, England) was used to measure soil water content (SWC) and soil temperature at a soil depth of 5 cm next to the rings simultaneously with the soil chamber measurements. Photosynthetic photon flux density (PPFD) was measured with a line quantum sensor (SQ-316-SS, Apogee, USA). Environmental parameters measured during the campaigns are summarized in Figure 1.

To measure the soil flux a fused silica bell, transmitting light between 170 and 5,000 nm was placed in a water-filled channel on top of a stainless steel ring (see supporting information for a sketch of the experimental setup, Figure S2). All tubing and fittings used were either made of perfluoroalkoxy alkane (PFA) or stainless steel. Each measurement cycle started with air being drawn with approximately 1.5 L/min from a tube outside, but near the chamber to quantify the ambient concentration, while the chamber line was flushed at the same flow rate. After 5 min, the lines were switched and the air in the headspace of the chamber was measured until a steady state was reached (11.6 ± 4.6 min); meanwhile, the ambient line was flushed at the same flow rate. Afterwards, the ambient concentration was measured a second time and a linear interpolation was performed to estimate the ambient concentration at the time of the steady state conditions within the chamber. The soil COS and CO₂ flux was calculated according to equation (1).
\[ F = \frac{q(C_2 - C_1)}{A} \]  

(1)

\( F \) denotes the soil flux in pmol m\(^{-2}\) s\(^{-1}\) for COS and \( \mu \)mol m\(^{-2}\) s\(^{-1}\) for CO\(_2\) (air temperature and pressure were measured to calculate the molar density), \( q \) is the flow rate (mol/s) while measuring the chamber concentrations, \( C_1 \) is the ambient concentration derived from the linear interpolation (pmol mol\(^{-1}\) for COS and \( \mu \)mol mol\(^{-1}\) for CO\(_2\) ), \( C_2 \) is the concentration in the chamber at steady state, and \( A \) is the surface area (0.032 m\(^2\)) covered by the chamber.

Differential pressure measurements (MKS Baratron Type 226A Differential Pressure Transducer, MKS Instruments Inc., Andover, MA, USA) were conducted with the same setup in Florian Kitz et al. (2017) to find the best flow rate, avoiding under inflation or overpressure within the chamber (Rayment & Jarvis, 1997). A flow rate of 1.55 standard liter (sl) per minute resulted in pressure differences being lower than 0.2 Pa, corresponding to the instrument resolution. Small COS emissions, with a maximum of 2 pmol m\(^{-2}\) s\(^{-1}\), from the setup itself were reported by Florian Kitz et al. (2017).

### 2.3. Laboratory Measurements

Soil samples were taken at each study site as soil cores of the uppermost 10 cm of the soil profile and stored at \(-20\) °C, a temperature where no changes in the microbial community composition can be expected (Lauber et al., 2010). The samples were sieved, with a mesh size of 2 mm, in preparation for the gas exchange measurements. To determine dry weight, a portion of each soil sample was oven dried at 105 °C for at least 24 h and subsequently weighted. The pH was determined in soil:water extracts (1:5, w/v) by using a pH Meter.
Metrohm 744 (Herisau, Switzerland). Total C and N contents were analyzed in dried samples, using a CN analyzer (TruSpec CHN; LECO, Michigan, USA) (supporting information Figure S3).

Soil samples were taken out of the freezer 24 h prior to measuring their gas exchange. They were filled into glass tubes with a diameter of 28 mm and a length of 150 mm. The glass tubes were connected to valves (Parker-Hannafin, Cleveland, OH, USA) switching between empty and filled, with soil, tubes every 5 min. The same instrument as used in situ, the Aerodyne Mini QCL, with the same settings was used for the lab measurements. Pressurized air was drawn into the QCL through water bottles, to humidify the air, and the glass tubes at a flow rate of 0.103 sl min⁻¹. Fluxes were calculated with equation (2).

\[ F = \frac{q(C_2 - C_1)}{g} \]  (2)

where \( F \) is the soil flux in pmol g⁻¹ s⁻¹ for COS and \( \mu \)mol g⁻¹ s⁻¹ for CO₂, \( q \) is the flow rate (mol/s), \( C_1 \) is the mean concentration of the empty tube measurements before and after the sample tube measurement (pmol mol⁻¹ and \( \mu \)mol mol⁻¹), \( C_2 \) is the concentration in the sample tube, and \( g \) is the soil fresh weight in grams. Blank measurements, with all tubes empty, were performed before the experiment started, and no contaminations were detected. The weight (g) used for each site and treatment varied to avoid huge volume differences, since soil with more organic material has a lower density compared to mineral soil. The system was given 1 hour to equilibrate for each set of measurements; data recorded during this period were discarded. Measurements were conducted in the dark, with aluminum foil covering the glass tubes, and under UV exposure. To expose the soil samples to UV radiation, a mercury vapor lamp (Actinic BL TL TL-K 40 W/10-R; Philips Lighting, Eindhoven, Netherlands), emitting primarily light between 350 and 400 nm with peak intensity at 370 nm, was used.

The programming language Python (Anaconda distribution, version 4.3.25) was used to calculate the fluxes and the “uncertainties” package (Lebigot n.d.) to propagate errors.

### 2.4. Statistics

Statistical analyses were carried out with R version 4.5.0—“Joy in Playing” (R Core Team, 2018), RStudio (RStudio Team, 2016), and the package “data.table” (Dowle & Srinivasan, 2018). In preparation for all statistical analyses the respective data were investigated as described in Zuur et al. (2010). In case of a violation of homogeneity of residuals, robust sandwich errors (Zeileis, 2004), implemented in the “sandwich package,” were used. In the DBF only the data from chambers with litter were used, and one observation, identified as outlier, was excluded. The predictors (PPFD, Tsoil, and SWC) were centered when interactions were included in the linear model to reduce multicollinearity. The relative importance of the predictors for the soil COS flux was calculated using the relaimpo package (Grömping, 2006) with the method “lmig” (Lindeman et al., 1980). Kruskal-Wallis and Dunn’s tests were performed to test for significant differences between dark/light, night/day and sites. An analyses of variance (ANOVA) was performed for the laboratory measurements.

### 3. Results

#### 3.1. In Situ Measurements

Mean COS fluxes during daytime were highest in the SAV and lowest in the ENF, while nighttime COS fluxes were lowest in the DBF (Table 1). Mean in situ COS and CO₂ fluxes from the SAV were significantly different from the DBF (\( p \) value <0.001) and the ENF (\( p \) value <0.001) during daytime. Nighttime in situ COS fluxes from the SAV were significantly different from the DBF (\( p \) value <0.001) and the ENF (\( p \) value <0.001); CO₂ fluxes from the SAV were only significantly different from the DBF (\( p \) value <0.01). DBF and ENF in situ COS fluxes were significantly different from each other during the day (\( p \) value <0.001), but not during the night (\( p \) value >0.05), while in situ CO₂ fluxes of the two sites were significantly different during daytime and nighttime (\( p \) value <0.01).

No significant location or interaction effects were detected in the multiple linear regression for the SAV. In both the DBF and the ENF interaction effects with SWC were significant (\( p \) value <0.05) and were subsequently included in the final models (for interaction plots see Figure 2).
For the simple SAV multiple linear regression \[ \text{COS flux} = \beta_0 + \beta_1 \text{PAR} + \beta_2 \text{Tsoil} + \beta_3 \text{SWC} \] no multicollinearity (variance inflation factors < 4) was detected, and the whole model had predictive capability (F Statistic: \( p < 0.05 \)). Since the residuals were heteroscedastic, robust sandwich standard errors were used.

For the SAV the multiple linear regression explained 74\% of the variation in the response variable (COS flux) and the predictors PAR and Tsoil were highly significant (\( p \) value < 0.001). PAR was the most important predictor in the model followed by Tsoil (Figure 3). For the DBF multiple linear regression \[ \text{COS flux} = \beta_0 + \beta_1 \text{PAR} + \beta_2 \text{Tsoil} + \beta_3 \text{SWC} + \beta_4 \text{Location 2} + \beta_5 \text{Location 3} + \beta_6 (\text{PAR*Location 2}) + \beta_7 (\text{PAR*Location 3}) + \beta_8 (\text{SWC*Location 2}) + \beta_9 (\text{SWC*Location 3}) + \beta_{10} (\text{Tsoil*Location 2}) + \beta_{11} (\text{Tsoil*Location 3}) \] multicollinearity between the predictors and interactions including them was detected, and the whole model had predictive capability (F Statistic: \( p < 0.05 \)). The model explained 36\% of the variation in the COS flux, Tsoil and interactions between the locations and SWC were significant (\( p \) value < 0.05). Tsoil was the most important predictor in the model (Figure 3). For the ENF multiple linear regression \[ \text{COS flux} = \beta_0 + \beta_1 \text{PAR} + \beta_2 \text{Tsoil} + \beta_3 \text{SWC} + \beta_4 \text{Location 2} + \beta_5 \text{Location 3} + \beta_6 (\text{PAR*Location 2}) + \beta_7 (\text{PAR*Location 3}) + \beta_8 (\text{SWC*Location 2}) + \beta_9 (\text{SWC*Location 3}) + \beta_{10} (\text{Tsoil*Location 2}) + \beta_{11} (\text{Tsoil*Location 3}) \] multicollinearity between the predictors and interactions including them was detected, and the whole model had predictive capability (F Statistic: \( p < 0.05 \)). The model explained 36\% of the variation in the COS flux, Tsoil and interactions between the locations and SWC were significant (\( p \) value < 0.05). Tsoil was the most important predictor in the model (Figure 3). For the ENF multiple linear regression \[ \text{COS flux} = \beta_0 + \beta_1 \text{PAR} + \beta_2 \text{Tsoil} + \beta_3 \text{SWC} + \beta_4 \text{Location 2} + \beta_5 \text{Location 3} + \beta_6 (\text{PAR*Location 2}) + \beta_7 (\text{PAR*Location 3}) + \beta_8 (\text{SWC*Location 2}) + \beta_9 (\text{SWC*Location 3}) + \beta_{10} (\text{Tsoil*Location 2}) + \beta_{11} (\text{Tsoil*Location 3}) \] multicollinearity between the predictors and interactions including them was detected, and the whole model had predictive capability (F Statistic: \( p < 0.05 \)). The model explained 36\% of the variation in the COS flux, Tsoil and interactions between the locations and SWC were significant (\( p \) value < 0.05). Tsoil was the most important predictor in the model (Figure 3). For the ENF multiple linear regression \[ \text{COS flux} = \beta_0 + \beta_1 \text{PAR} + \beta_2 \text{Tsoil} + \beta_3 \text{SWC} + \beta_4 \text{Location 2} + \beta_5 \text{Location 3} + \beta_6 (\text{PAR*Location 2}) + \beta_7 (\text{PAR*Location 3}) + \beta_8 (\text{SWC*Location 2}) + \beta_9 (\text{SWC*Location 3}) + \beta_{10} (\text{Tsoil*Location 2}) + \beta_{11} (\text{Tsoil*Location 3}) \] multicollinearity between the predictors and interactions including them was detected, and the whole model had predictive capability (F Statistic: \( p < 0.05 \)). The model explained 36\% of the variation in the COS flux, Tsoil and interactions between the locations and SWC were significant (\( p \) value < 0.05). Tsoil was the most important predictor in the model (Figure 3).

Table 1

Mean COS and CO2 fluxes (Mean ± SD) for the Lab and Field Experiments

| Site     | COS flux (pmol m\(^{-2}\) s\(^{-1}\)) | CO2 flux (μmol m\(^{-2}\) s\(^{-1}\)) |
|----------|-------------------------------------|-------------------------------------|
|          | Daytime                            | Nighttime                           |
| SAV      | 5.8 ± 7.8                          | 0.2 ± 1.5                           |
| DBF      | −3.2 ± 3.7                         | −6.6 ± 0.6                          |
| ENF      | −3.5 ± 1.1                         | −2.6 ± 0.9                          |
| Laboratory |                                      |                                      |
| SAV      | 5.4*10\(^{-4}\) ± 1.4*10\(^{-4}\) | 2.4*10\(^{-4}\) ± 9.1*10\(^{-5}\) |
| DBF      | −2.6*10\(^{-4}\) ± 7.0*10\(^{-5}\) | −6.6*10\(^{-4}\) ± 6.7*10\(^{-5}\) |
| ENF      | −2.9*10\(^{-5}\) ± 1.7*10\(^{-4}\) | −8.9*10\(^{-5}\) ± 6.6*10\(^{-5}\) |

Figure 2. Interaction plots for the interactions SWC * Location in the DBF (a) and the ENF (b) as included in the multiple linear regression models described in the text.
Multicollinearity between the predictors and interactions including them was detected, and the whole model had predictive capability (F Statistic: $p < 0.05$). The model explained 55% of the variation in the COS flux. SWC, location, and interactions between the locations and SWC were significant ($p$ values <0.05). SWC was the most important predictor in the model followed by the interaction between SWC and location (Figure 3). Coefficients and standard errors for the linear models are summarized in Table 2.

### 3.2. Laboratory Incubations

Mean COS fluxes were highest in the SAV and lowest in the DBF, both in the dark and under UV radiation (Table 1). All three sites were significantly different ($p < 0.001$) to each other in their COS and CO$_2$ flux means.

Mean COS fluxes from SAV soil samples were significantly different between samples collected under the tree and in the open and between the dark and light treatment (see ANOVA summary in Table 3). There was also a significant interaction between those two factors (see ANOVA results, Table 3). CO$_2$ fluxes were significantly different between the SAV sampling locations and between the dark and light exposed samples, but there was no significant interaction. Mean COS fluxes from DBF soil samples were significantly different between samples collected in differently dense vegetated areas and between the dark and light treatment applied in the lab (see ANOVA results, Table 3). There was no significant interaction between the factors (see ANOVA results, Table 3). CO$_2$ fluxes were significantly different between DBF sampling locations and between the dark and light exposed samples. There was also a significant interaction effect between the sample location and dark/light exposure. Mean COS fluxes from ENF soil samples were significantly different between samples collected at the elevation and in the depression and between the dark and light treatment applied in the lab (see ANOVA results, Table 3). There was also a significant interaction between those two factors (see ANOVA results, Table 3). CO$_2$ fluxes were significantly different between ENF sampling locations and between the dark and light exposed samples, but the interaction was not significant. The COS/CO$_2$ ratio was significantly different between locations and dark/light treatments in samples from all
three sites (SAV, DBF, and ENF). The mean difference in the COS flux between light and dark treatment correlated with the soil organic matter (SOM) content of the soil samples. Samples with higher SOM content had a higher mean difference between the treatments (e.g., the difference in the COS flux between the light and dark treatment was higher) compared to samples with lower SOM content, even across sites, with the exception of the ENF depression samples (see supporting information Figure S4).

4. Discussion

A recent synthesis paper (Whelan et al., 2018) summarized soil COS fluxes from a multitude of different studies and ecosystems with a majority of reported COS uptake in oxic soils in the range of $-10$ to $0$ pmol m$^{-2}$ s$^{-1}$. Our in situ measurements of soil COS fluxes in both forest ecosystems (DBF and ENF), in Denmark and Estonia, were well in this range (Figure 4). The oak savannah (SAV) in Spain though was characterized by very high positive soil COS fluxes (Figure 4), especially in the open grass during daytime, exceeding most values found in the literature (Whelan et al., 2018). These high fluxes correlate with high light intensities on the soil surface, favored by the removal of the above-ground vegetation in the chambers, which would otherwise provide shading. In order to estimate the surface COS fluxes in the SAV under natural conditions, hence with intact vegetation, and to put it into perspective with the measured fluxes, the linear model (Table 2), driven by an estimate of the PPFD at the soil surface based on the Beer-Lambert law (Campbell & Norman, 1998), can be used. A measured PPFD of $2,191 \mu$mol m$^{-2}$ s$^{-1}$, the maximum observed at this site, would equate to 693.8 $\mu$mol m$^{-2}$ s$^{-1}$ if shading by the vegetation, with a mean leaf area index (LAI) of 2.3 (calculated from the removed biomass) and an estimated extinction coefficient of 0.5 (Zhang et al., 2014), was factored in. Plugging in the derived PPFD into the linear model (Table 2), with a soil temperature of 25 °C and a SWC of 20%, would yield a soil COS flux of 10.03 pmol m$^{-2}$ s$^{-1}$ under maximum light conditions but with shading by the vegetation, still a high value but comparable to COS fluxes measured by Maseyk et al. (2014) or Florian Kitz et al. (2017).

Table 2

| Predictor | Estimate | Std. error | Estimate | Std. error | Estimate | Std. error |
|-----------|----------|------------|----------|------------|----------|------------|
| Intercept | 1.59E+01*** | 2.62E+01 | -2.38 | 1.30E+01 | -2.57E+00*** | 2.10E+01 |
| PAR       | 9.57E-03**  | 3.72E+00 | -2.21E+00 | 3.71E+03 | -3.62E+00*** | 5.78E+00 |
| SWC       | 1.58E-02*** | 3.79E-02 | 6.11E+00 | 1.34E+02 | 8.37E+00**  | 1.35E+02 |
| Location 2| 1.12E+00   | 1.12E+00 | 3.27E+00 | 1.50E+00 | 9.29E+00**  | 1.15E+00 |
| Location 3| -3.16E+00  | 1.97E+00 | -2.63E+00 | 1.92E+00 | 7.17E+00**  | 2.12E+00 |
| PAR * location 2 | 4.31E+00 | 6.31E+00 | -2.37E+00 | 5.78E+00 |
| PAR * location 3 | 1.43E+00 | 7.31E+00 | 1.98E+00 | 1.35E+00 |
| SWC * location 2 | 5.28E+00 | 1.99E+00 | 9.66E+00*** | 1.35E+00 |
| SWC * location 3 | 1.34E+00 | 2.14E+00 | 1.34E+00 | 1.15E+00 |
| Soil * location 2 | 1.76E-01 | 1.94E-01 | 1.24E-01 | 1.15E-01 |
| Soil * location 3 | 4.33E-02 | 2.13E-01 | 1.24E-02 | 1.15E-01 |

Table 3

| Predictor | F value | P value |
|-----------|---------|---------|
| SAV       |         |         |
| Location 1| 2.46E-06 | 1050.3  | <2e-16*** |
| UV        | 1.56E-06 | 2334.2  | <2e-16*** |
| Location * UV | 9.20E-08 | 39.4    | 1.76E-09*** |
| Residuals | 224     | 5.24E-07 |         |
| ENF       |         |         |
| Location 1| 2.90E-07 | 94      | <2e-16*** |
| UV        | 1.82E-06 | 2396.7  | <2e-16*** |
| Location * UV | 0.00E+00 | 0.001   | 9.70E-01 |
| Residuals | 224     | 7.74E-07 |         |

Note. DF = degrees of freedom; SS = sum of squares. *** p < 0.001, ** p < 0.01, * p < 0.051.
The linear regression confirms that the COS flux in the SAV was primarily driven by radiation, as indicated by PPFD being the most important predictor, explaining 66% of the variability in the observed surface COS flux. In the two denser forest sites (DBF and ENF) in Denmark and Estonia PPFD was not a good explanatory variable for the in situ soil COS flux, most likely due to a very narrow range of PPFD on the soil surface and low light levels (DBF: 44.4 ± 74 μmol m⁻² s⁻¹; ENF: 18.9 ± 21.2 μmol m⁻² s⁻¹) in general. But as it can be seen in the no litter chambers in the DBF, sunlight penetrating the forest canopy for a short period of time

**Figure 4.** CO₂ (a) and COS (b) fluxes measured in situ in the savannah (SAV) in Spain, the deciduous broadleaf forest (DBF) in Denmark, and the evergreen needleleaf forest (ENF) in Estonia. Colors denote the location within the ecosystems, with under the Tree (T), open savannah (O), and intermediate (I) in the SAV, low (LV), medium (MV), and high vegetation (HV) surrounding the rings in the DBF and elevation (E) and depression (D) in the ENF. (c) The ratio of the COS and CO₂ flux.

The linear regression confirms that the COS flux in the SAV was primarily driven by radiation, as indicated by PPFD being the most important predictor, explaining 66% of the variability in the observed surface COS flux. In the two denser forest sites (DBF and ENF) in Denmark and Estonia PPFD was not a good explanatory variable for the in situ soil COS flux, most likely due to a very narrow range of PPFD on the soil surface and low light levels (DBF: 44.4 ± 74 μmol m⁻² s⁻¹; ENF: 18.9 ± 21.2 μmol m⁻² s⁻¹) in general. But as it can be seen in the no litter chambers in the DBF, sunlight penetrating the forest canopy for a short period of time
can lead also in this ecosystem to a strong response in the soil COS flux (supporting information Figure S1). The strong light response of the soil COS flux was also seen in the lab, where all soil samples showed a strong reaction to UV-A exposure (Figure S5), especially remarkable was the strong intrasite variability (Figure S5). Whelan and Rhew (2015), who exposed their soil samples to visible light, also saw an increase of COS production under light exposure, even in sterilized soil samples, a response that increased even more when roots were kept in the soil. Our samples exhibited a comparable behavior, samples with higher soil organic matter content usually exhibiting a stronger reaction to light (see supporting information Figure S4), which holds true within and between sites, with the exception of samples from the ENF depression (Figures 5 and 1). The effect of radiation on organic material is understood to be complex (King et al., 2012) and can vary with identity, for example, chemical composition, and age of the substrate (Austin et al., 2016; Day et al., 2015; Gallo et al., 2009). SOM and its decomposability can vary greatly in soils (Marschner et al., 2008) depending on the composition of organic input to the soil system (Kögel-Knabner, 2002) and interactions with mineral components of the soil (Marschner et al., 2008; Mikutta et al., 2006), but knowledge about its chemical and physical properties is limited (Dungait et al., 2012; Hedges et al., 2000). Meredith et al. (2018) showed that the soil sulfur species composition can vary greatly in soils, which in turn impacts COS production. In their study they were able to link higher sulfate concentrations to increased COS production. The question whether the source of the soil COS emission under light exposure is primarily abiotic or biotic is difficult to investigate, since the experimental setup was not designed to distinguish between facilitation of microbial decomposition (Lin et al., 2018), through photo-priming, and direct abiotic destruction of organic material by radiation. Previous studies (F. Kitz et al., 2019; Meredith et al., 2018) hypothesized that COS production is caused by coupled abiotic and biotic processes, with biotic processes often providing the precursors from which COS is produced abiotically. With the absence of high light levels as a driving factor in DBF and ENF other parameters gained importance in explaining the in situ variability, but the overall explained variance by the linear models decreased substantially (Table 2). In DBF soil temperature was the factor best explaining the observed variability in the soil COS fluxes (Figure 3) but not so in SAV. The soil temperature range in DBF and ENF was too narrow to investigate an optimum curve, indicating temperature-dependent biological activity with an optimum for biological COS uptake, as proposed by Kesselmeier et al. (1999), while in SAV COS emissions increased linearly with soil temperature, indicating primarily abiotic processes, a response reported previously in the literature (Maseyk et al., 2014; Whelan et al., 2016; Whelan & Rhew, 2015). When soil temperature increased at the SAV site, soil CO$_2$ fluxes increased as well. The soil CO$_2$ flux can serve as an estimator for biological activity in the soil and litter layer, since it primarily originates from respiration processes (Hanson et al., 2000; Insam & Haselwandter, 1989; Jensen et al., 1996). The increase in CO$_2$ emissions in SAV therefore indicates an increase in biological activity in the soil with increasing temperature, a common but very variable response of soils observed in many different ecosystems (Conant et al., 2011; Davidson et al., 1998; Davidson & Janssens, 2006; Jensen et al., 1996; Lloyd & Taylor, 1994). Since there are more accounts for biotic COS consumption (Kato et al., 2008; Kato et al., 2012; Kesselmeier et al., 1999; Masaki et al., 2016; Ogawa et al., 2013; Ogawa et al., 2016; Sauze et al., 2017) than biotic COS production (Conrad, 1996; Masaki et al., 2016) under ambient COS concentrations, it is tempting to link biological activity in oxic soils primarily to COS uptake. The co-occurring increase in COS and CO$_2$ emissions in SAV could therefore mean that abiotic production of COS is increasing more strongly with temperature outweighing increased COS uptake due to higher biological activity or the underlying assumption of biotic COS contribution being primarily characterized by COS uptake in oxic soils is not universally true, and emissions associated to the soil biological activity could also occur.

Another parameter, which is connected to microbial activity and is considered to have a decisive influence on soil COS fluxes, is the soil water content (Bunk et al., 2017; Kesselmeier et al., 1999; Ogée et al., 2016; Van Diest & Kesselmeier, 2008). Soil COS uptake is expected to exhibit a bell-shaped curve along a SWC gradient (Kesselmeier et al., 1999; Van Diest & Kesselmeier, 2008), with a reduction of soil COS uptake at low SWC, due to reduced microbial and enzyme activity (Borken & Matzner, 2009; Davidson & Janssens, 2006; Lavigne et al., 2004; Sardans & Penuelas, 2005) caused by water limitation, and reduced diffusivity of COS in the soil column at high SWC (Ogee et al., 2016; Sun et al., 2015). Even though our in situ measurements cover a wide range of SWC (Figure 1) the impact of soil moisture is difficult to separate from associated soil parameters, like soil organic matter content, which has to be kept in mind when evaluating the SWC contribution to the linear models. In the two wetter ecosystems, DBF and ENF, the interaction between SWC and location was
significant, which complicates the interpretation of the main effect, SWC, in the model (Irwin & McClelland, 2001; McClelland & Judd, 1993). But if we investigate the SWC impact on soil COS fluxes in the different locations separately (Figure 2), a decrease in COS uptake with increasing SWC in ENF and the reverse in DBF is observed. A striking phenomenon is the strong intrasite variability, which is likely caused by the large differences in soil properties (Figure 1) within the sites, especially in ENF. This suggests that the response of the soil COS flux to changing SWC is very variable and depends greatly on soil and litter layer (Sun et al., 2016) properties.

Figure 5. COS (a) and CO₂ (b) fluxes measured in the laboratory from SAV, DBF, and ENF soil samples. Colors denote the sampling location within the ecosystems, with under the Tree (T) and open savannah (O) in the SAV, low (LV) and high vegetation (HV) surrounding the rings in the DBF and elevation (E) and depression (D) in the ENF. “light” indicates soil samples exposed to UV radiation. (c) The ratio of the COS and CO₂ flux.
Soil COS fluxes in the laboratory, in the dark and under the same temperature conditions, did also vary between soil samples from the different sites. Such variability between different soils under laboratory conditions was observed previously (Bunk et al., 2017; Kaisermann, Ogée, et al., 2018; Meredith et al., 2019; Whelan et al., 2016); notable in our experiment is the consistently positive COS flux from the SAV soil. Positive soil COS fluxes were rarely observed in untreated soil samples in previous laboratory studies. In the few experiments where COS emission were detected (Bunk et al., 2017; Whelan et al., 2016), the corresponding soil samples were very dry, as with the SAV samples (Figure 1). Other soil parameters, found to influence the soil COS exchange in the literature, are soil pH (Sauze et al., 2017) and soil nitrogen content (Kaisermann, Jones, et al., 2018; Meredith et al., 2018). Soil pH differed between our samples (Figure 1) with COS uptake in the more acidic samples and COS production in the less acidic samples, while no apparent pattern emerged from the C/N ratio of our samples with regard to soil COS fluxes (supporting information Figure S3).

5. Conclusions

Our approach to complement field with laboratory measurements allowed us to address the three formulated hypotheses under in situ and controlled conditions. From both the field and laboratory measurements, radiation emerged as the most impactful driver for the soil COS flux, confirming hypothesis (I). But in ecosystems with low light levels on the soil surface, like in dense forests, the importance of radiation for in situ soil COS flux was reduced, despite its potential impact seen in the lab. In situ, intersite variability in soil COS fluxes, as proposed in hypothesis (II), was observed in the forest sites, DBF and ENF, and the savannah. And while in situ measurements did not reveal a significant difference in the mean of the soil COS flux between the forests, laboratory measurements did, probably originating from the differences between soil samples in the lab and natural soils with a developed litter layer and living roots. Most surprising to us was the large intrasite variability, hypothesis (III), in both in situ and laboratory measurements, which did also change the response to UV radiation. We suspect especially the amount and chemical composition of soil organic matter to be of great importance for explaining the observed differences in the light response and the overall differences within and between sites. It is important to note that while we have separated parameters, like SWC and SOM, in analyses and discussion, they are of course related to each other (Ankenbauer & Loheide, 2017) and indivisible in situ. Among the biggest obstacles we currently face in explaining soil COS fluxes is our lack of understanding for the basic chemical processes leading to COS emissions. Future studies should therefore explore the effect of environmental parameters, especially radiation, on organic matter in the soil and on the soil surface and their interactions with biological decomposition processes.

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