



Changes in soil pore structure generated by the root systems of *maize*, *sorghum* and switchgrass affect in situ N_2O emissions and bacterial denitrification

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Abstract

Due to the heterogeneous nature of soil pore structure, processes such as nitrification and denitrification can occur simultaneously at microscopic levels, making prediction of small-scale nitrous oxide (N_2O) emissions in the field notoriously difficult. We assessed N_2O+N_2 emissions from soils under maize (*Zea mays* L.), switchgrass (*Panicum virgatum* L.), and energy sorghum (*Sorghum bicolor* L.), three potential bioenergy crops in order to identify the importance of different N_2O sources to microsite production, and relate N_2O source differences to crop-associated differences in pore structure formation. The combination of isotopic surveys of N_2O in the field during one growing season and X-ray computed tomography (CT) enabled us to link results from isotopic mappings to soil structural properties. Further, our methodology allowed us to evaluate the potential for in situ N_2O suppression by biological nitrification inhibition (BNI) in energy sorghum. Our results demonstrated that the fraction of N_2O originating from bacterial denitrification and reduction of N_2O to N_2 is largely determined by the volume of particulate organic matter occluded within the soil matrix and the anaerobic soil volume. Bacterial denitrification was greater in switchgrass than in the annual crops, related to changes in pore structure caused by the coarse root system. This led to high N-losses through N_2 emissions in the switchgrass system throughout the season a novel finding given the lack of data in the literature for total denitrification. Isotopic mapping indicated no differences in N_2O -fluxes or their source processes between maize and energy sorghum that could be associated with the release of BNI by the investigated sorghum variety. The results of this research show how differences in soil pore structures among cropping systems can determine both N_2O production via denitrification and total denitrification N losses in situ.

Keywords Nitrification · N_2O isotope mapping · X-ray CT · Anaerobic soil volume · Plant roots · Pore structure · BNI

Introduction

N_2O is a highly potent greenhouse gas with a substantial global warming impact that can also harm the stratospheric ozone layer (Ravishankara et al. 2009; Tian et al. 2020). Agriculture is responsible for the majority (60%) of anthropogenic N_2O emissions (Syakila and Kroeze 2011), through management practices like tillage and fertilizer application (Butterbach-Bahl and Dannenmann 2011; McGill et al. 2018; McSwiney and Robertson 2005). Biofuels from cellulosic bioenergy feedstocks make it possible to reduce the climate impact of fossil fuel energy consumption; their positive contribution to mitigating climate change, however, might be diminished by greenhouse gas (GHG) emissions, including N_2O (Oates et al. 2016; Qin et al. 2015; Walter et al. 2015; Wightman et al. 2015). While N_2O production is known to be affected by plant species composition (Butterbach-Bahl

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and Dannenmann 2011; Stehfest and Bouwman 2006) we do not fully understand the mechanisms influencing N_2O emission from soils under different vegetation systems well enough to develop effective solutions for curbing overall GHG emissions.

In soils, N_2O is a product of an array of N transformations (Robertson and Groffman 2015), with heterotrophic bacterial denitrification (bD), autotrophic nitrification (Ni), nitrifier denitrification (nD), and fungal denitrification (fD) regarded as primary sources (Butterbach-Bahl et al. 2013; Müller et al. 2014). Even though these processes may drastically differ from each other in terms of conditions necessary for their occurrence and the microorganisms involved, the extremely high micro-scale heterogeneity of the soil matrix enables them to produce N_2O in a close spatial proximity (Braker and Conrad 2011; Rohe et al. 2021). Disentangling the specific drivers responsible for enhanced N_2O emissions in the field is therefore notoriously difficult.

Availability of O_2 is one of the major physical factors controlling N_2O fluxes (Bollmann and Conrad 1998; Groffman et al. 1988; Rohe et al. 2021). For example, O_2 discriminates between N_2O production via denitrification, i.e. the anoxic reduction of nitrate (NO_3^-) to N_2 with N_2O as an intermediate, which takes place in the absence of O_2 , and nitrification, where N_2O is a by-product during the oxidation of hydroxylamine (NH_2OH) to nitrite (NO_2^-), which requires aerobic conditions. In contrast to bD, the major end product of fD is N_2O since fungal denitrifiers lack N_2O reductase (Baggs 2011; Philippot et al. 2011). In contrast to coupled nitrification-denitrification where the formation of nitrate and subsequent nitrate reduction are conducted by different microorganisms, in nD these tasks are performed by the same organism, the process benefitting from low organic C and O_2 but high N availability (Braker and Conrad 2011; Wrage et al. 2001).

Denitrification is thus favored by wet conditions, available C and nitrate, and is often a major N_2O source during high N_2O flux events (Baggs 2011; Wrage et al. 2004). Yet, denitrification also can make surprisingly sizeable contributions to N_2O fluxes from relatively dry soils. The latter is an outcome of local anaerobicity observed within decomposing plant residues or within soil matrix, e.g., centers of soil aggregates (Schlüter et al. 2018; Wrage et al. 2001). Occurrence and proliferation of anaerobic microsites within the soil matrix, which we will refer to as anaerobic soil volume fraction, is hard to quantify directly, but can be assessed indirectly through measures of diffusivity as well as model simulations based on air-filled pore volumes (Andersen and Petersen 2009; Balaine et al. 2013). Anaerobic soil volume fraction can also be manipulated in bulk (whole) soil samples by controlling the water filled pore space (WFPS) (Chen et al. 2016; Kim et al. 2022; Kravchenko et al. 2017). The WFPS 70–80% has been shown to be optimal for denitrification (Butterbach-Bahl

et al. 2013). However, such bulk measurements do not consider pore structure, a key factor controlling the microscale distribution patterns in anaerobic soil volume fraction. Recent advancements in X-ray computed tomography (X-ray CT) imaging allowed visualization of anaerobic soil volume fraction at micro-scales and demonstrated that anaerobic soil volume fraction, quantified via distances to air-filled pores, can predict N_2O emissions as well as magnitudes of complete denitrification (Kravchenko et al. 2018; Rabot et al. 2015; Rohe et al. 2021).

Not only O_2 influx, but also O_2 demand is highly controlled by the pore architecture. Intense microbial activity boosts oxygen consumption, creating local anaerobic microsites which stimulates N_2O production (Kim et al. 2020, 2021; Kravchenko et al. 2017). Microbial activity's impact on N_2O emissions thus depends on pore architecture in a complex not yet fully understood manner. Some studies suggest that narrow macropores (30–150 μm diameter (\varnothing)) can provide a perfect environment for microbial decomposers, hence, for close spatial coupling of N_2O production and emission during decomposition of fresh plant residues (Kim et al. 2020, 2021, 2022; Kravchenko et al. 2017). Yet, lower availability of O_2 stimulated by greater distances from pores to decomposing soil particulate organic matter (POM) can enhance N_2O emissions (Ortega-Ramírez et al. 2023).

Plant roots play the most important role in shaping soil pore architecture through direct formation of biopores and indirect repacking and rearranging of soil solids (Lucas et al. 2019, 2022). Biopores formed by roots range from $\sim 30 \mu\text{m}$ to $5000 \mu\text{m}$, i.e. span three orders of magnitude (Yunusa and Newton 2003). Narrow biopores (e.g., 30–150 μm \varnothing) can be particularly important due to their oversize contribution to the overall connectivity of the pore system (Lucas et al. 2020). Differences in root architectures in plants of different species or plant communities can have a sizeable influence on narrow macropore formation (Bacq-Labreuil et al. 2019; Bodner et al. 2014; Lucas et al. 2022), hence potentially influencing the size of anaerobic soil volume fraction. For example, a comparison of several bioenergy cropping systems demonstrated that polyculture vegetation communities decreased soil anaerobic soil volume fraction compared to monoculture maize and switchgrass, and that such decreases were associated with lower N_2O emissions (Kravchenko et al. 2018).

The other two important routes through which roots can affect N_2O emissions are through direct alterations of soil N balance through N uptake vs. N inputs via exudation and rhizodeposition (Jones et al. 2009; Moreau et al. 2019), and by shaping the composition and functions of the soil microbiome and rhizosphere processes (Berendsen et al. 2012; Hinsinger et al. 2009). However, the actual significance of plant-microbe interactions for the soil N-cycle, e.g. through the release of labile C as drivers of rhizosphere

denitrification, is currently not well understood (Baggs 2011; Moreau et al. 2019; Philippot et al. 2009). Certain plant species, e.g., sorghum a promising bioenergy crop, can produce and release biological nitrification inhibitors (BNIs) into the soil, potentially suppressing nitrification, through the reduction of ammonia-oxidizing bacteria and archaea (Li et al. 2021; Sarr et al. 2020; Subbarao et al. 2007, 2015; Tesfamariam et al. 2014). However, sizes and ecological impacts of biological nitrification inhibition of sorghum in the field are yet to be determined by measurements of in-situ gross nitrification and denitrification, such as through isotopic techniques (Nardi et al. 2022).

With few exceptions, for example, Rohe et al. (2021), previous studies measured only the emitted N_2O , lacking the ability to identify the processes that led to its production or to assess the full extent of denitrification. Therefore, while N losses in the soil in the form of N_2 can be substantial, their measurements are complicated by a high atmospheric N_2 background (Lewicka-Szczebak et al. 2017; Yu et al. 2020). By analyzing the isotopic signatures of N_2O , including the $\delta^{18}\text{O}$ value of oxygen, the bulk $\delta^{15}\text{N}$ value, and the intramolecular distribution of ^{15}N in N_2O (site preference, SP), it is possible to gain insights into the origins of N_2O emissions (Yu et al. 2020). A way to derive quantitative information on N_2O sources from such isotopic analyses is isotopic mapping using $\delta^{15}\text{N}^{\text{SP}} / \delta^{18}\text{O}$ (Yu et al. 2020). Based on the isotopic enrichment of residual N_2O during the reduction to N_2 , it further allows to derive denitrification product ratio [$\text{N}_2\text{O} / (\text{N}_2\text{O} + \text{N}_2)$] (*pr*) and thus to quantify complete denitrification through $\text{N}_2\text{O} + \text{N}_2$ fluxes (Lewicka-Szczebak et al. 2017).

The objectives of the study were 1) to conduct field monitoring of N_2O and comparisons of $\text{N}_2\text{O} + \text{N}_2$ emissions and their N_2O component from the soils under energy sorghum, maize and switchgrass crops, grown for bioenergy stock production; 2) to distinguish among the prevalent pathways of N_2O production in the soils of these crops; and 3) to elucidate the potential role of soil pore structure for influencing $\text{N}_2\text{O} + \text{N}_2$ emissions and their sources. We collected N_2O throughout the growing season using static flux chambers and implemented $\delta^{15}\text{N}^{\text{SP}} / \delta^{18}\text{O}$ isotope mapping to estimate the relative contribution of different microbial pathways to N_2O production as well as to quantify the reduction of N_2O to N_2 . X-ray CT imaging of undisturbed soil cores allowed us to quantify pore structure and elucidate its contribution to micro-environmental conditions prevalent within the soil matrix. We also hypothesized that the reported ability of sorghum roots, in contrast to maize, to reduce nitrification potential by BNI (Subbarao et al. 2007; Tesfamariam et al. 2014) will be manifested in the field through greater N_2O production via denitrification than nitrification pathway. We also hypothesized that a capacity of switchgrass to reduce narrow macropores and increase the anaerobic soil volume

(Kravchenko et al. 2019, 2022) will result in an increased importance of complete denitrification, as compared to that in the other two crops.

Methods

Field design and management

The DOE-Great Lakes Bioenergy Research Center (GLBRC) Biofuel Cropping System Experiment (BCSE) site was established in 2008 at the Kellogg Biological Station (KBS) Long-term Ecological Research site (Robertson and Hamilton 2015) in Hickory Corners, Michigan [42°23'47" N, -85°22'26" W, 288 m a.s.l.]. Site soils are loamy, well-drained Alfisols developed on glacial outwash with loess inputs. The experiment is a randomized complete block design with five replicate blocks. We evaluated N_2O emissions from three systems during the 2021 field season: Monocultures of switchgrass (*P. virgatum* L. variety Cave-in-rock), maize (*Zea mays* L., Pioneer P0306Q) and energy sorghum (*Sorghum bicolor* L., TAM 17651). Before 2018, the energy sorghum plots contained continuous maize + cover crops.

Seeding and fertilizer application differed between the treatments (Fig. 2a, top row). Details on the agricultural management in 2021 and before can be found on <https://aglog.kbs.msu.edu/>. In summary, maize was seeded on the 15th of May with a starter fertilizer supplying 34 kg N ha⁻¹ and energy sorghum was seeded on the 19th of May with 56 kg N ha⁻¹. At the end of June an additional 137 kg N ha⁻¹ (28% Urea Ammonium Nitrate, UAN) was injected in the middle of the rows of the two crops. The switchgrass plots were sprayed with 28% UAN on the 13th of May supplying 56 kg N ha⁻¹ with no further N additions. All systems were managed without tillage.

N_2O sampling

Two static (closed-cover) flux chambers were installed within 2 m of one another in each of the five blocks in May 2021 and removed only for agronomic operations for a total of 10 chambers per cropping system. Each chamber consisted of a cylindrical metal base and an airtight plastic lid (surface area = 641 cm², headspace volume = 16.6 L) and was hammered 5 cm deep in the soil. Atmospheric pressure within the sealed chamber was maintained by a piece of coiled stainless-steel tubing (0.5 m X 0.32 cm OD and 0.18 cm ID) extending from the interior to exterior of the chamber. Gas samples were taken from an approx. 50 cm long and 0.6 mm outer diameter polyurethane tubing connected to the headspace (Fig. 1a). During sampling a pre-evacuated 250 ml glass bottle was connected to a steel needle at the

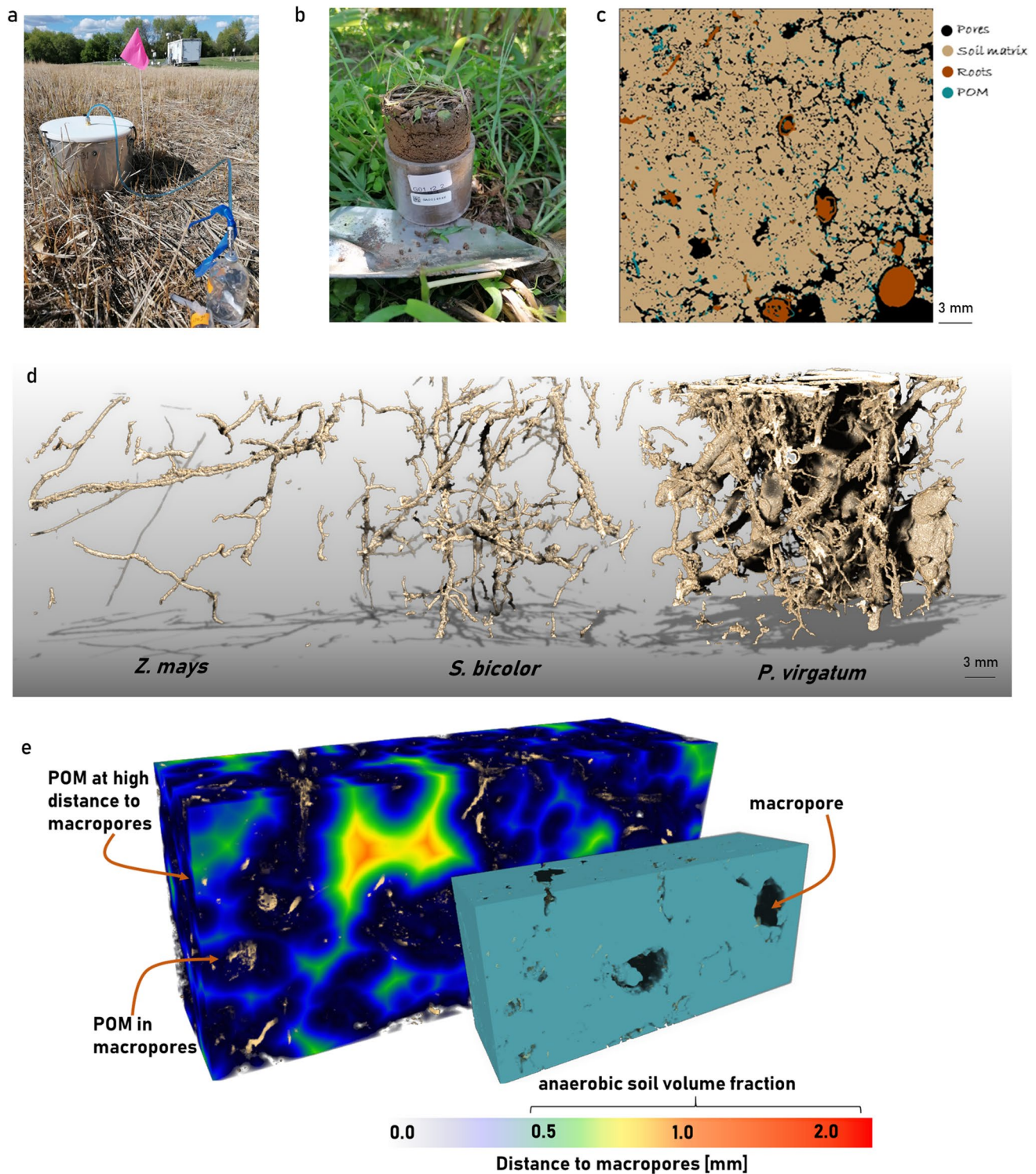
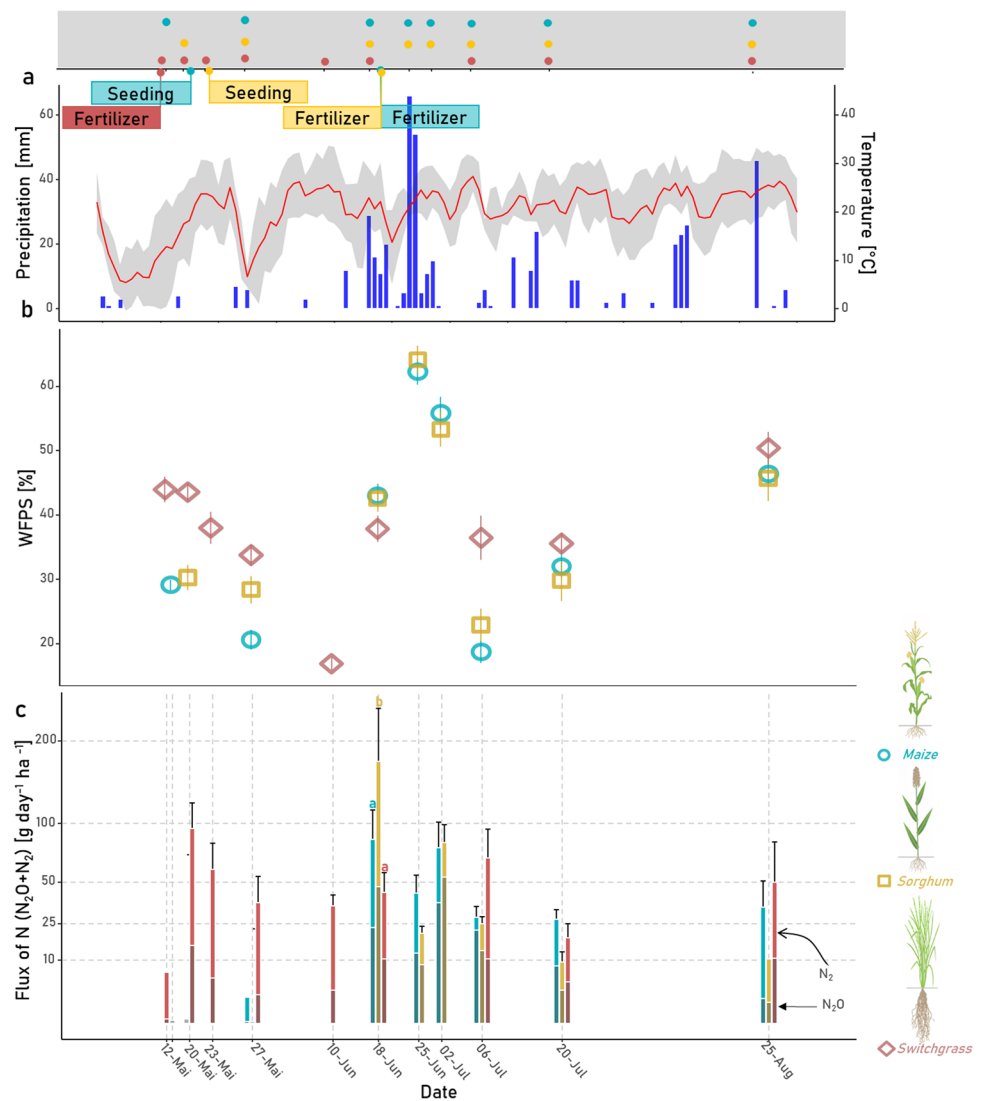


Fig. 1 Photographs and CT visualizations. **a)** Collecting N_2O from a static flux chamber at the beginning of the season. **b)** A soil core taken within the base of the chamber. **c)** An image slice from a CT-scanned soil core showing pores, roots, particulate organic matter (POM), and soil solid matrix identified on the image. **d)** Visualization

of the root system of maize, energy sorghum and switchgrass within the soil core obtained using X-ray CT and **e)** visualization of the soil matrix (turquoise) and macropores (black) as well as the visualization of POM (brown) within the 3D distance map to air-filled macropores.

Fig. 2 a) Daily precipitation (blue bars) and mean daily temperature (red line). Shadow represents max. and min. temperature range. b) Water filled pore space (WFPS). c) $\text{N}_2\text{O}+\text{N}_2$ fluxes during the cropping season for the three bioenergy crops. The darkest parts of the bars represent the fraction of N_2O . The top row shows the sampling days for the three studied crops and marks the dates of seeding and fertilization events. Whiskers show the standard errors of the means. Different letters mark significant differences among the crops within the same sampling date ($p<0.05$). On some of the sampling dates N fluxes were too low to provide reliable isotopic values, thus no values are reported on c). Note that N_2 was calculated based on the pr derived from the isotopic mapping approach.



end of the tubing for 1 min to assure diffusive equilibration. Additionally, a 30 ml pre-evacuated glass bottle was filled. Chamber closing times ranged between 70 min and 470 min to ensure sufficient N_2O concentrations for later isotopic characterization. Initial closures were adjusted to the expected fluxes, i.e. with the shortest closing times for maize directly after fertilization and the longest for switchgrass before fertilization and timed to complete sampling by noon. Additionally, 250 ml atmospheric gas samples were taken on the respective sampling days in plot 1 of each of the respective plant treatments at the height of the top of the chamber.

Sampling was conducted throughout the 2021 season with 7–8 sampling events per crop. Some of the sampling events took place on a regular basis, i.e., once a month, throughout the growing season, with the first sampling on 12th of May and the last on 25th of August. Other sampling events targeted anticipated enhanced GHG emissions, including samplings at 3, 7, 14, and 30 days after N fertilization and

sampling on the day of the first large rain event (>30 mm, Fig. 2a) that followed a long May–June drought.

N_2O flux calculations

Measurements of N_2O concentration in the 30 ml samples were carried out using a gas chromatograph (GC-ECD, Shimadzu GC-2014) with an analytical precision of approx. 2%. Using these concentrations, the N_2O fluxes were calculated based on an increase in the N_2O concentration from that of air using the ambient mean of the measured N_2O concentration of 335 ppbv during the closing time. For all samples with N_2O concentrations > 300 ppbv, the paired taken 250 ml samples were used for isotopic characterization (see below). In that case, the N_2O concentration from respective analyses was used to calculate N_2O fluxes.

Isotopic characterisation

An Elementar IsoPrime 100 stable isotope ratio mass spectrometer (IRMS) interfaced to a Trace Gas inlet system (Elementar; Mt. Laurel, NJ) was used to measure $\delta^{15}\text{N}_{\text{bulk}}$, $\delta^{15}\text{N}_{\alpha}$, $\delta^{15}\text{N}_{\beta}$ and $\delta^{18}\text{O}$ of N_2O as previously described (Sutka et al. 2003). By analyzing the mass-to-charge (m/z) 44, 45, and 46 in intact N_2O^+ molecular ions, we determined the bulk $\delta^{15}\text{N}$ and $\delta^{18}\text{O}$ isotope signatures, while the $\delta^{15}\text{N}_{\alpha}$ values were detected by the m/z 30 and 31 of NO^+ fragment ions generated in the mass spectrometer. N_2O is a linear molecule consisting of two N atoms (NNO), with one of the N atoms in the central position (α site) and the other at the terminal position (β site). The distribution of ^{15}N within the N_2O molecule is called site preference (SP) and is defined as the difference in $\delta^{15}\text{N}$ values between the α ($\delta^{15}\text{N}_{\alpha}$) and β ($\delta^{15}\text{N}_{\beta}$) sites. The isotopic values are presented as deviation from the $^{15}\text{N}/^{14}\text{N}$ and $^{18}\text{O}/^{16}\text{O}$ ratios of atmospheric N_2 and the Vienna Standard Mean Ocean Water (VSMOW), respectively. The analytical precision determined as standard deviation (1σ) of primary standards measurements was 0.5‰ for $\delta^{15}\text{N}_{\text{bulk}}$ and $\delta^{18}\text{O}$, 0.4‰ for $\delta^{15}\text{N}_{\alpha}$, and $\delta^{15}\text{N}_{\beta}$ and 0.6‰ SP. The $\delta^{15}\text{N}_{\text{bulk}}$, $\delta^{15}\text{N}_{\alpha}$, $\delta^{15}\text{N}_{\beta}$, $\delta^{18}\text{O}$ and SP values of the two laboratory N_2O primary standards are -0.69‰, 11.51‰, -12.88‰, 40.16‰ and 24.39‰ and -0.77‰, -1.12‰, -0.42‰, 39.17‰ and -0.70‰, respectively as determined by calibration against international reference material USGS51 and USGS52 (Ostrom et al. 2018). The dilution of the ambient air in the flux chamber was corrected based on the increase in the N_2O concentration from that of the measured mean ambient concentration and the respective isotopic values throughout the season ($\delta^{15}\text{N}$ 6.1 ± 0.2 , $\delta^{15}\text{N}^{\text{SP}}$ 13.4 ± 0.5 , $\delta^{18}\text{O}$ 41.7 ± 0.2). To assure high accuracy, we further analyzed only isotopic readings of samples with N_2O concentrations greater than 130% of the ambient N_2O concentration.

We followed the " $\delta^{15}\text{N}^{\text{SP}} / \delta^{18}\text{O}$ isotope mapping technique" to estimate the relative contribution of the different microbial N_2O production pathways to the total $\text{N}_2\text{O} + \text{N}_2$ production (Lewicka-Szczebak et al. 2020; Yu et al. 2020). We used recently summarized data from Yu et al. (2020) to map the endmembers, i.e. the microbial source isotope values for bD/ nD and Ni processes and between bD/nD and fD as well as the N_2O reduction (Tab. S1). The $\delta^{18}\text{O}$ values were corrected for by the mean $\delta^{18}\text{O}$ of annual precipitation water (-7.3 ‰) derived from the Waterisotopes Database (<http://waterisotopesDB.org>, Accessed 01.11.2022). Adding the endmembers into the $\delta^{15}\text{N}^{\text{SP}} / \delta^{18}\text{O}$ isotope plot allows us to derive the different slopes of the mixing line between bD+nD and fD or Ni, as well as the reduction line for isotopic enrichment of residual N_2O . The latter allows us to calculate the denitrification product ratio [$\text{N}_2\text{O} / (\text{N}_2\text{O} + \text{N}_2)$] (pr) and

thus to derive total $\text{N}_2\text{O} + \text{N}_2$ emissions. Note that bD and nD cannot be distinguished by this method.

We followed a protocol (Lewicka-Szczebak et al. 2017, 2018) taking into account the sample position in the and $\delta^{15}\text{N}^{\text{SP}} / \delta^{18}\text{O}$ map using a mixing equation for the bacterial fraction and the Rayleigh equation for N_2O reduction. In short, to derive the relative contributions of the endmembers, two scenarios are assumed, of which we present mean values: In the first, N_2O is produced by bD and partially reduced; then a mixing of residual N_2O with unreduced N_2O from Ni or fD appears. In the second, these processes happen vice versa. The $\delta^{18}\text{O}$ endmembers for Ni and fD taken from Yu et al. (2020) are sufficiently differentiated to allow both mixing-lines between bD-Ni and bD-fD to be distinguished (Lewicka-Szczebak et al. 2020). Since multiple mixing curves cannot be evaluated simultaneously with the equations of Lewicka-Szczebak (2018), we distinguished between bD-Ni and bD-fD mixing as two possible instances of end-member mixing as suggested in Lewicka-Szczebak et al. (2020). In the event that the samples were located below the mean reduction line, the calculation results provide the fraction of bD values slightly higher than 1, which were set to 1 for further summaries.

WFPS and N forms

Soil moisture was measured (0–10 cm depth) at three locations close to each chamber at every sampling event using a volumetric soil moisture sensor (HydroSense II, Campbell Scientific, Logan UT, USA).

Additionally, we took disturbed soil samples (~100g) from around the chambers at various time points throughout the season (Tab. S2), including sample dates before and after fertilization as well as after the rain event. The soil samples were stored at -20°C before extraction. For this, a homogenized sample of approx. 10 g of fresh soil was extracted with 0.1 mol KCl. Available NH_4^+ and NO_3^- were analyzed in the MSU soil test laboratory according to Sinsabaugh et al. (2000) and Doane and Horwath (2003), respectively.

After scanning the undisturbed cores with X-ray CT (see below), we derived the bulk density of the cores gravimetrically. The water filled pore space was then calculated based on the measured water contents and the bulk density within the different plots. In addition, soil of these cores was used to measure the pH-value (in water).

Weather data (daily precipitation and temperature) are from <https://lter.kbs.msu.edu/datatables/12>.

X-ray CT

After the last sampling campaign in late August, one intact soil core (5 cm Ø, 5 cm height) was taken from 1 to 6 cm depth under the base of each static chamber (Fig. 1b). These

cores were subjected to X-ray CT shortly after collection. The soil cores were scanned using an X-ray microtomograph (X3000, North Star Imaging, Rogers, USA) at 75 kV and 470 μ A. Since the samples were scanned using a continuous subpiX mode, a resolution of 18.2 μ m could be achieved, although the respective energy settings resulted in a larger focal spot on the VarianL07 detector panel (size 1920 * 1536 pixels). During a scan with four subimages (2 rows and 2 columns), 2880 projections were acquired at 3 frames per second with an average of 2 frames. The 3D image reconstruction was performed with the efX reconstruction software.

Image processing and analyses

The reconstructed images were cut into cubes of 1850x1850x2300 voxels in Fiji (V. 153n, Schindelin et al. 2012). This was done, to avoid analyzing disturbed regions at the core walls. Then the images were segmented into four classes, namely pores, soil matrix, POM and roots (Fig. 1c). For this, we used a random forest classifier trained in ILASTIK (Berg et al. 2019) to pore segments, soil matrix, and a class that includes roots and POM. To train the classifier, we used subvolumes of five randomly chosen images for annotation. The out-of-bag error was <0.01. We were not able to further compare image-based POM to POM conventionally analyzed, but a similar protocol was used by Schlüter et al. (2022) to show good agreement of image-based POM and conventionally analyzed POM. Moreover, visual expectation of the images showed no over-segmentation of POM particles, although under-segmentation could potentially happen for small POM particles due to our resolution of 18.2 μ m. Such small POM particles, however, seem to be distributed more evenly in the soil matrix (Schlüter et al. 2022) and therefore would not affect the analyzed distribution of POM. After segmenting all plant residues, we further differentiated between POM and roots in Fiji, in which objects of the mixed class were assigned to roots (Fig. 1d) only if they were connected to the outer boundary of the image and were larger than 10.000 voxels, i.e. approx. 0.06 cm³. For the latter, we used the “connected components labeling” and the “size opening” functions of the plugin MorphoLibJ (Version 1.4.3, Legland et al. 2016).

The anaerobic soil volume fraction, i.e., the volume fraction of air distance larger than a threshold (Fig. 1e), was calculated as written in Rohe et al. (2021) by computing the Euclidean Distance Transform for the pore image. In addition to computing the anaerobic soil volume fraction and the visible porosity (pores > 0.036 mm), we computed the Γ -indicator as a third metric to estimate oxygen supply. For this, the pore image was labelled using the connected component labelling from the plugin BoneJ2 plugin (V. 7.10, Domander et al. 2021). This image was used to calculate the

Γ -indicator, which is a metric of pore connectivity (Lucas et al. 2020).

To measure the volume of pores between 0.036 mm and 0.15 mm \varnothing , we used the local thickness method (Hildebrand and Rüegsegger 1997) in Fiji. We refer to the volume of pores between 0.036 mm and 0.15 mm \varnothing by pores < 0.15 mm \varnothing and report their volume relative to the volume of the soil core. Additionally, the image of the Euclidean Distance Transform (Fig. 1e) was also used to calculate the mean distance of POM to macropores similar to Ortega-Ramírez et al. (2023). Note that one image of the switchgrass cores contained a massive volume of roots, as only the root sod was sampled (Fig. 1d). This sample was handled as an outlier, as the large root volume (>10 %) led to unreliable information on the soil matrix. It was also excluded from the bulk density estimation.

Statistical analysis

Effects of the plant treatment (maize vs. switchgrass vs. energy sorghum) on the studied flux data and isotopic characteristics as well soil structural properties derived from X-ray CT scans were investigated using linear mixed model approach implemented in the lme4-package (Bates et al. 2015) of R (V. 4.1.1). These models extend simple linear models to include the non-independent nature of our sample hierarchical structure, i.e. the different chambers within one plot. The random effects assigned consisted of the treatment plots, used as an error term for testing the plant treatment effect, and the flux chambers nested within the plots, used as an error term. Additionally, for the studied flux data and isotopic characteristics, the time point of sampling was added as a fixed factor to the model. Because on some dates only maize and energy sorghum were sampled, we constructed two models, the first including only dates with these two plant treatments, while the second contained only sampling campaigns of all three plant treatments. The assumptions of normality and homogeneity of variances were assessed using normal probability plots of the residuals and Levene's tests for equal variances, respectively. When the normality assumption was found to be violated, the data were logarithmically transformed; when the equal variance assumption was violated, the unequal variance models were fitted using the package ‘nlme’ in R, respectively.

To address our second research question concerning the evaluating soil structural properties as predictors of denitrification, we computed the correlation matrix of Pearson's correlation showing coefficients in R using the ‘corrplot’ package. In addition to the correlations including data from all sampling days, we computed these correlation matrixes for specific days for which all data (pore structure, N₂O, and soil chemistry) were available.

Due to the different management strategies of bioenergy crops we did not analyze the pore structural correlations with the total fluxes across the plant systems. For this, we computed plant treatment specific linear regressions of parameters derived from X-ray CT (anaerobic soil volume fraction, distance of POM to macropores, Pores <150µm) with the mean N_2+N_2O emissions of the flux chambers. In addition, linear regression show the response of the fraction of bD and the pr to the anaerobic soil volume fraction and the distance of POM to macropores across the plant systems, where the fraction of bD as well as the pr were log scaled. Available NH_4^+ and NO_3^- were analyzed only in one sample per plot only, that is, no mixed effect model was necessary and we used an analysis of variance (ANOVA) in conjunction with Tukey's HSD test implemented in the 'agricolae' package (Mendiburu and Yaseen 2020).

Results

N_2O+N_2 fluxes

At the beginning of the season (early May), the perennial system switchgrass was wetter compared to maize and energy sorghum (Fig. 2b) systems. Fertilization of switchgrass during this time led to a peak in N_2O+N_2 emissions with fluxes > 100 g N day⁻¹ ha⁻¹, with N_2O less than 25 g N day⁻¹ ha⁻¹ (Fig. 2c). Due to low precipitation (Fig. 2a), by late May WFPS had dropped substantially (Fig. 2b)

and N_2O emissions were barely detectable (Fig. 2c). The large rain event of mid-June led to high fluxes of N_2O+N_2 in all three systems, with both N_2O (approx. 75 g N day⁻¹ ha⁻¹) and N_2O+N_2 emissions significantly higher in energy sorghum compared to maize and switchgrass. Shortly after the first large rain event, the two annual crops were fertilized, and multiple heavy rains followed within two days, greatly increasing soil water contents. The N_2O+N_2 emissions after fertilization did not differ between energy sorghum and maize and were substantially lower than the emission peaks observed after the first rain event of 18 June. Seven days after this fertilization their N_2O+N_2 fluxes again increased substantially (approx. 100 g N day⁻¹ ha⁻¹), with a particularly large share of N_2O (approx. 50 g N day⁻¹ ha⁻¹). Interestingly, after the low June fluxes and despite the lack of fertilization, N_2O+N_2 emissions from switchgrass increased in early July and then again in late August, with a continuously low share of N_2O .

The mean N_2O+N_2 fluxes for the entire season did not differ significantly among the three crops (Table 1). Mean N_2O fluxes were significantly higher in energy sorghum (18.6 g N day⁻¹ ha⁻¹) compared to maize (14.6 g N day⁻¹ ha⁻¹), while numerically the lowest in switchgrass stands (9.2 g N day⁻¹ ha⁻¹).

Soil NH_4^+ and NO_3^- concentrations did not significantly differ among the systems at any point during the season (Table S2) except in July when the soil of maize plots had significantly higher NO_3^- concentrations compared to energy sorghum and switchgrass.

Table 1 Mean values (\pm standard errors) of the main N_2O flux parameters and pore structural properties of the three bioenergy systems. Shown are measured N_2O fluxes, calculated N_2O+N_2 fluxes, the corresponding fraction of bacterial denitrification (bd), and the product ratio (pr) assuming a mixing of bacterial and fungal denitri-

fication. In addition, also shown are the values of f_{bd} 2 and pr 2 calculated assuming a mixing of bD and nitrification (Ni). The Γ -indicator is the probability which describes the connectivity of the pore system. Different letters indicate significant differences between plant systems ($p<0.05$).

Treatment	Maize		Sorghum		Switchgrass	
N_2O flux [g N day ⁻¹ ha ⁻¹]	14.57	$\pm 2.33a$	18.64	$\pm 3.13b$	9.22	$\pm 2.61a$
$N_2O + N_2$ flux [g N day ⁻¹ ha ⁻¹]	29.64	$\pm 7.87a$	24.98	$\pm 6.75a$	31.70	$\pm 7.57a$
Fraction of bD [-]	0.79	$\pm 0.03a$	0.75	$\pm 0.05a$	0.95	$\pm 0.01b$
pr [-]	0.49	$\pm 0.03b$	0.49	$\pm 0.06b$	0.24	$\pm 0.01a$
Fraction of bD 2 [-]	0.86	$\pm 0.02a$	0.82	$\pm 0.04a$	0.96	$\pm 0.01b$
pr 2 [-]	0.27	$\pm 0.02a$	0.27	$\pm 0.02a$	0.19	$\pm 0.01a$
Macroporosity [% of total volume]	14.12	$\pm 1.17a$	13.18	$\pm 1.45a$	12.55	$\pm 0.80a$
Pores <150 µm Ø [% of total volume]	5.77	$\pm 0.52b$	5.28	$\pm 1.05ab$	3.19	$\pm 0.45a$
POM [% of total volume]	0.95	$\pm 0.11a$	1.18	$\pm 0.16a$	1.21	$\pm 0.19a$
Root [% of total volume]	0.18	$\pm 0.01a$	0.21	$\pm 0.06a$	1.38	$\pm 0.28b$
anaerobic soil volume fraction [% of total volume]	4.44	$\pm 1.21a$	7.19	$\pm 1.97a$	14.51	$\pm 2.63b$
Distance of POM to macropores [mm]	0.09	$\pm 0.01a$	0.11	$\pm 0.01b$	0.14	$\pm 0.01b$
Γ -indicator [-]	0.75	$\pm 0.04a$	0.77	$\pm 0.06a$	0.79	$\pm 0.04a$
Bulk density [g cm ⁻³]	1.54	$\pm 0.05a$	1.56	$\pm 0.05a$	1.48	$\pm 0.06a$
pH [-]	6.53	$\pm 0.43a$	6.32	$\pm 0.43a$	5.86	$\pm 0.16a$

Isotopic Characterization

On most dates when isotopic characterization was performed for all three cropping systems switchgrass had higher $\delta^{18}\text{O}$, $\delta^{15}\text{N}_{\text{bulk}}$, and lower $\delta^{15}\text{N}_{\text{SP}}$ values compared to maize and energy sorghum (Fig. S1). Only in late August, the N_2O from switchgrass had significantly lower $\delta^{18}\text{O}$ values compared to the two other plants. Consequently, in the isotopic mapping of $\delta^{18}\text{O} / \delta^{15}\text{N}_{\text{SP}}$ the maize and energy sorghum data are distributed mainly in-between the fD-bD mixing line and the reduction line, while the switchgrass data are mostly clustered around the reduction line (Fig. 3). High fluxes were directly associated with a higher fraction of bD (Fig. S2), thus yellow to red points in Fig. 3 are found only close to the reduction line.

Reasoned summarized isotopic endmember values from Yu et al. (2020) enable us to differentiate between two potential mixing lines (bD-Ni and bD-fD) and thus we must select the appropriate mixing scenario (Lewicka-Szczepak et al. 2020). In this study, the pr and the fraction of bD are nearly identical for both mixing scenarios (Fig. 4, Fig. S3). This is because of the generally high fraction of bD especially in switchgrass. As most points lay far below the bD-fD mixing line and are distant from Ni, indicating the low importance of Ni, we focus our analysis on the results in Fig. 4, assuming a primary mixing of bD and fD. That said, the influence of Ni (Fig. S3) cannot be excluded for all points and later is discussed separately.

The calculated values of the fraction of bD and pr from Fig. 3 revealed a high importance of bD in switchgrass systems and a large share of N_2O 's being reduced to N_2 as compared to maize and energy sorghum systems (Table 1, Fig. 4). An exception was the late August sampling date, where the fraction bD in switchgrass became numerically

lower and the pr numerically higher than those in maize and energy sorghum. While throughout the season energy sorghum and maize did not differ from each other in terms of either the fraction of bD and pr, shortly after fertilization energy sorghum had a significantly lower bD relative contribution compared to maize (Fig. 4). If we assume the mixing of bD and Ni only (Fig. S3), this would suggest that the importance of Ni was higher in sorghum systems compared to maize that particular sampling event, while on all other days there was no difference between the two crops in the relative contributions of Ni and bD to N_2O production. Indeed, the corresponding points can be found above the bD-fD mixing line in the subplot of energy sorghum (Fig. 3).

Correlation with pore structural properties

We conducted a sensitivity test and correlated the volume fraction of varying minimum distances to pores with the $\text{N}_2\text{O} + \text{N}_2$ fluxes to derive the anaerobic soil volume fraction (Rohe et al. 2021). While there was no significant correlation for switchgrass throughout all distances, the two other plants had significant correlations for a range of minimum distances (Fig. S4). For computing the final anaerobic soil volume fraction, we used the distance >0.41 mm away from pores, for which the lowest mean p-value for the two annual plants was found.

pH-values, bulk density, macroporosity (percent of pores $>40 \mu\text{m}$ Ø), Γ -indicator (connectivity of the pore system), and percent POM were not significantly different among the three systems (Table 1). Yet, switchgrass had significantly higher root volumes and anaerobic soil volume fraction compared to energy sorghum and maize. Furthermore, switchgrass decreased the $<150 \mu\text{m}$ Ø pores and increased the mean distance from POM to macropores.

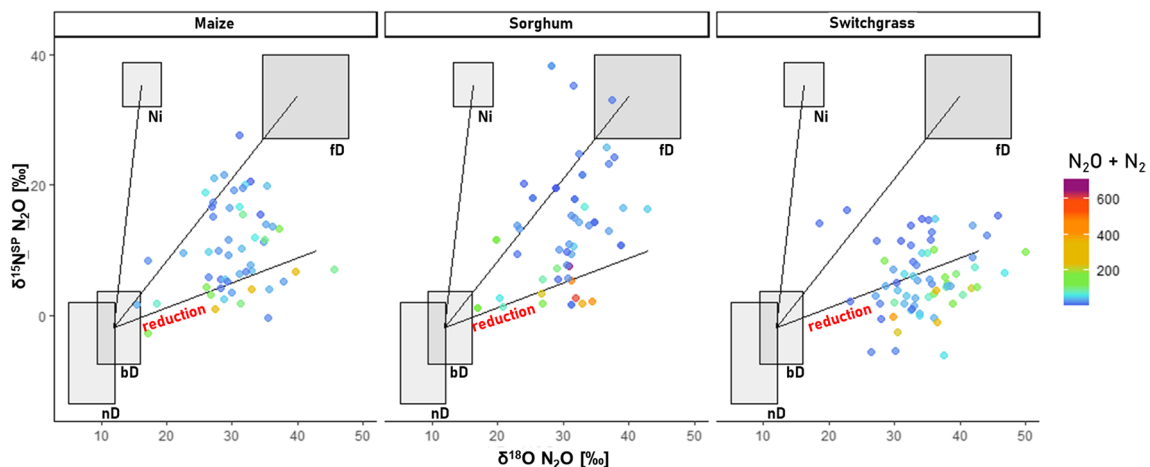
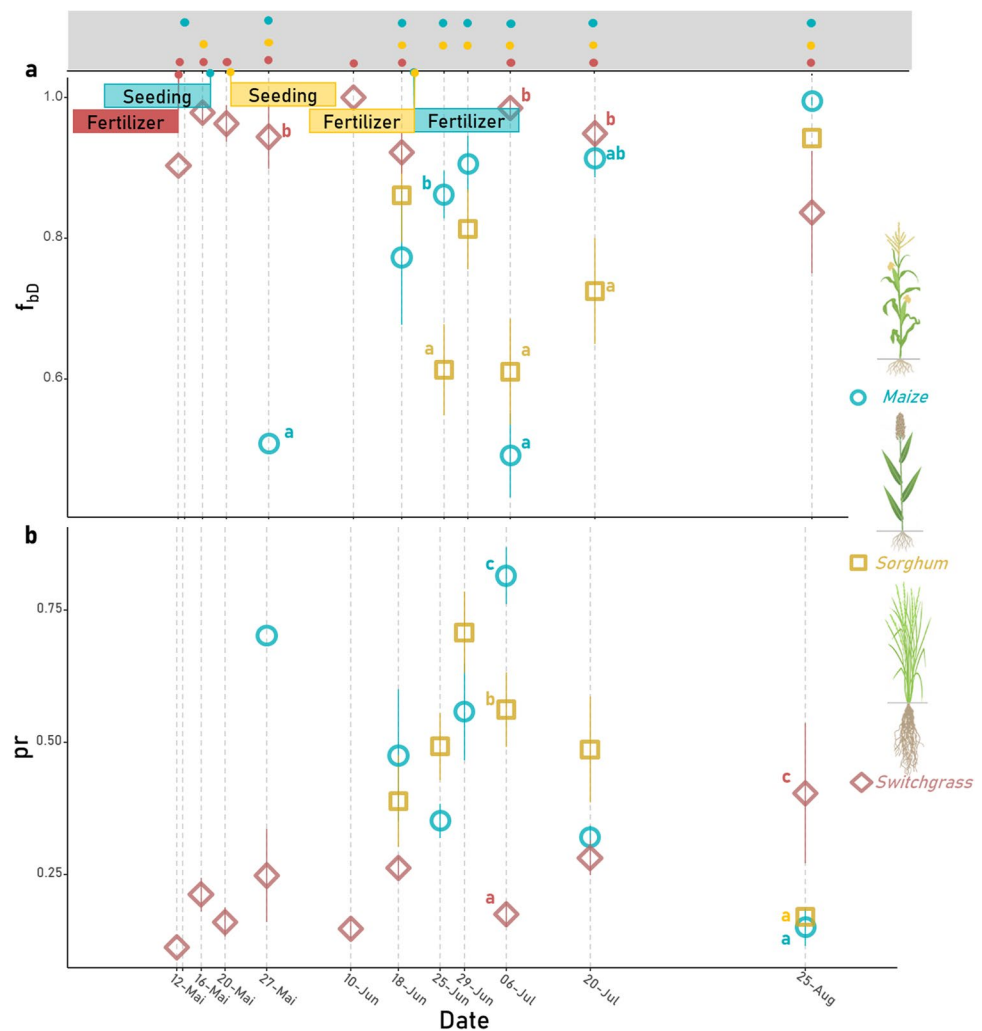


Fig. 3 Isotopic mapping for the three bioenergy crops. Squares show the end-members of Ni (nitrification), fD (fungal Denitrification), nD (nitrifier Denitrification) and bD (bacterial Denitrification)

Fig. 4 Fraction of N_2O originating from bacterial denitrification of total emitted N_2O (f_{bd}) and the denitrification product ratio (pr). The calculation assumes a mixing between bacterial denitrification and fungal denitrification.



Several of these properties were correlated with each other (Fig. S2a). The anaerobic soil volume fraction and the distance of POM to macropores were highly positively correlated, and both were negatively correlated to pores $<150 \mu m \varnothing$ and positively to the root volume.

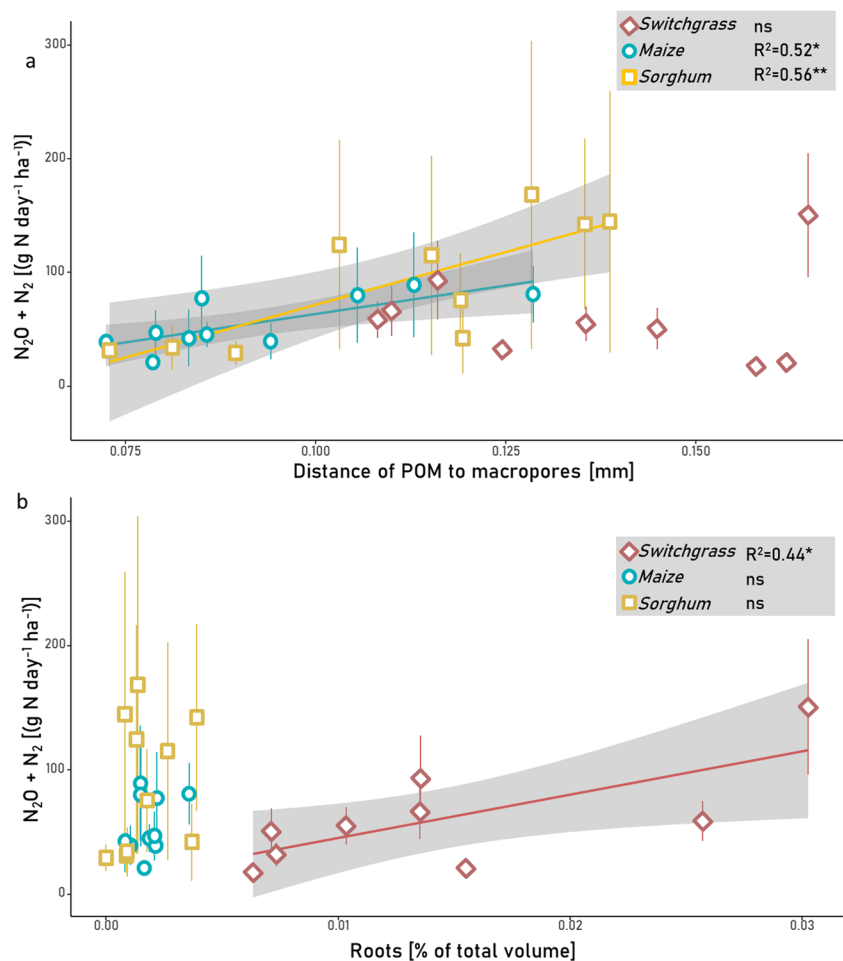
The distance of POM to macropores was positively correlated to N_2O+N_2 emissions in soils of maize and energy sorghum, but not in switchgrass (Fig. 5a). The same pattern was observed for the association between anaerobic soil volume fraction and N_2O+N_2 emissions (Fig. S5a). The abundance of $<150 \mu m \varnothing$ pores was related to N_2O+N_2 (negatively) only in sorghum soils (Fig. S5b). In addition, only in switchgrass was the root volume positively associated with the N_2O+N_2 emissions throughout the season (Fig. 5b).

Associations between N_2O+N_2 emissions and pore structure characteristics in the studied crops varied throughout the season (Fig. S2). For example, N_2O+N_2 emissions 7 days after fertilization (20th of May) were positively correlated to $<150 \mu m \varnothing$ pores in switchgrass (Fig. S5c). The unexpectedly high N_2O+N_2 emissions in switchgrass soil

in late August was positively associated with the volume of roots observed within intact soil cores (Fig. S5d, $p < 0.1$), while not with other pore structural properties (not shown).

When examined across all three systems, there was a non-linear positive trend of the fraction of bd's increasing with increases in the anaerobic soil volume fraction (Fig. S6a) and the distance of POM to macropores (Fig. 6a). The fraction of bd was particularly low in soils of energy sorghum and maize when the soil anaerobic soil volume fraction was at its lowest and the distance from POM was at its highest, with bd fraction's as low as $<55\%$ in some energy sorghum plots. Note that this distance represents the mean of all POM within a given sample. The fraction of bd increased to $>90\%$ in switchgrass and plateaued after the distance of POM to macropores exceeded $0.1 mm$. The trend was opposite for pr, which in energy sorghum and maize soils exceeded 60% at small anaerobic soil volume fractions and short distances to POM, and then decreased to $<50\%$ with increasing the distance of POM to macropores. Yet, pr was $<40\%$ in switchgrass soils across the entire range of observed distances of

Fig. 5 Relationship of the distance of POM to macropores (a) and roots (b) with $\text{N}_2\text{O} + \text{N}_2$ fluxes. Each point represents mean values of fluxes throughout the season for one static flux chamber. Stars indicate the significant association between the two parameters for the specific plant; $p < 0.05 = *$, $p < 0.01 = **$, ns = non-significant. Shadows show 95% confidence level interval for the predictions of the linear model.



POM to macropores. Note that the mean WFPS during the season did correlate with the total $\text{N}_2\text{O} + \text{N}_2$ emissions, but not with the fraction of bD and pr (Fig. S2a).

Discussion

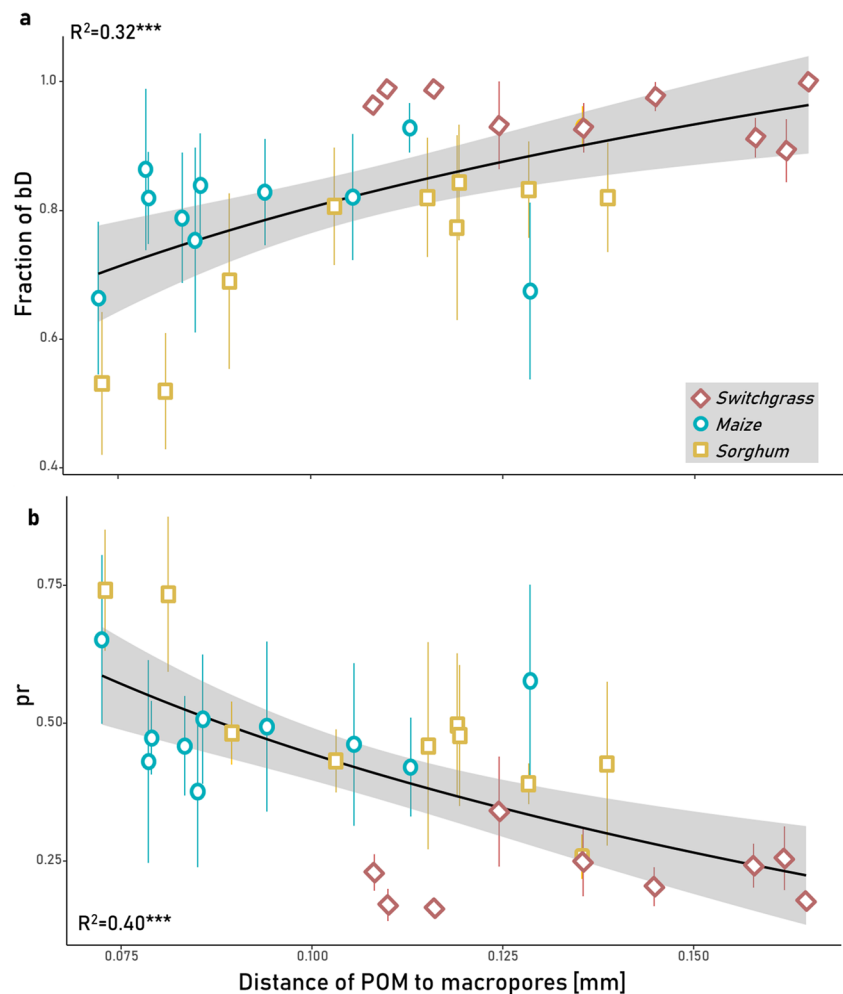
The effects of the pore structure on denitrification and N_2 production across bioenergy systems.

Our observations support the notion of high importance of bD in N_2O production, especially during events of high emissions (Fig. 3, Fig. 4a, Di Liang and Robertson 2021; Gao et al. 2023; Kravchenko et al. 2018; Ostrom et al. 2021). Furthermore, high anaerobic soil volume fractions and the distance of POM to macropores lead to a high amount of N_2O being reduced to N_2 (Fig. 6b,d). The reason for this could be that N_2O formed in the anaerobic soil volume fraction, e.g. at hotspots of occluded POM, can potentially be reduced to N_2 before it reaches the air-filled pore space (Braker and Conrad 2011; Rohe et al. 2021). As expected, high emissions of $\text{N}_2\text{O} + \text{N}_2$ corresponded to fertilizer

application dates and changing soil moisture (Fig. 2a). The change in pore structure, however, led to high N_2O emissions only in energy sorghum and maize, e.g. after fertilization (28th of June and 6th of July), where a relatively high denitrification pr [$\text{N}_2\text{O} / (\text{N}_2\text{O} + \text{N}_2)$] (Fig. 4b) resulted in N_2O fluxes $> 50 \text{ g N day}^{-1} \text{ha}^{-1}$ (Fig. 2c). Thus, mean N_2O emissions increased in the order switchgrass $<$ maize $<$ energy sorghum (Table 1).

In rainfed areas of the US, bioenergy systems using sorghum were found to have comparable N_2O emissions to those using maize (Kent et al. 2020). Indeed, in our study the two systems behaved very similarly with 1) a peak after fertilization and 2) the reduced importance of bD one month after fertilization (Fig. 3a) as measured on the same fields two years before for maize (Fig. 2c, Ostrom et al. 2021), while the plots of switchgrass deviated from the others not only due to management strategies, e.g. different timing and amount of fertilization. Switchgrass was found to have lower N_2O emissions compared to maize (Tab. 1 and as shown before by Abraha et al. 2018 and Ostrom et al. 2021). The large anaerobic soil volume fractions in switchgrass, which were associated with a high fraction of bD and a low pr

Fig. 6 Influence of the distance of POM to macropores on (a) fraction of bacterial denitrification (bD) and the (b) denitrification product ratio (pr). Each point represents mean values of fluxes throughout the season for one static flux chamber. Stars indicate the significant association between the two parameters; $p < 0.05 = *$, $p < 0.01 = **$, ns = non-significant. Shadows show 95% confidence level interval for the predictions of the linear model.



(Tab. 1, Fig. S6), lead to relatively low N_2O emissions in switchgrass throughout the season (Fig. 2c). Even after the fertilization event in switchgrass (20th of May), N_2O emissions were low ($< 25 \text{ g N day}^{-1} \text{ ha}^{-1}$), while $N_2O + N_2$ emissions peaked at approx. $100 \text{ g N day}^{-1} \text{ ha}^{-1}$ (Fig. 2c). In such a system with high relative contribution of bD, gross N_2O consumption can exceed N_2O production, which can make denitrification a net N_2O sink (Philippot et al. 2011). Note that due to error propagation the calculation of N_2 emissions from isotopic mapping tend to be imprecise especially at low fluxes, where methodological uncertainties are highest. A comparison to the ^{15}N tracing technique applied to the field, however, showed that the mapping technique offers valid qualitative information about the N_2 emissions, which are due to the high atmospheric N_2 background rarely measured (Lewicka-Szczebak et al. 2020).

However, the distance threshold for anaerobic conditions was set to 0.41 mm, based on a sensitivity analyses before further evaluation (Fig. S4). This is more than double than that found by Kravchenko et al. (2018), but also about 1/10 of the 5 mm found by Rohe et al. (2021). The

latter's large distances, however, were related to repacked soil conditions, which can create completely different conditions with trapped gas pockets and reduced hotspots due to missing POM as compared to structured soil (Rohe et al. 2021). When, locally, the O_2 demand exceeds the O_2 supply, denitrification is favored (Rohe et al. 2021). Our results confirm former studies that show the capacity for X-ray CT to estimate the anaerobic soil volume fraction to describe O_2 availability (Kravchenko et al. 2018; Rabot et al. 2015; Rohe et al. 2021) and additionally to quantify local hotspots created by POM (Kim et al. 2020; Kravchenko et al. 2018; Ortega-Ramírez et al. 2023). These hotspots are microsites of high O_2 demand and unless connected to an air-filled macropore, anaerobic conditions develop that allow denitrification to occur. The anaerobic soil volume fraction and the distance of POM to macropores were highly correlated in our study (Fig. S2a), making both parameters equally good predictors of $N_2 + N_2O$ fluxes (Fig. 5a, Fig. S5a) and of the activity of denitrifying bacteria (Fig. S6, Fig. S6a).

The highly negative correlation of pores $< 150 \mu\text{m}$ \varnothing with the anaerobic soil volume fraction (Fig. 2a) is in

agreement with previous findings showing the importance of these pores to reduce the volume of anaerobic microsites (Kravchenko et al. 2018). Therefore, the $<150\ \mu\text{m}$ \varnothing pores were negatively associated with $\text{N}_2\text{O}+\text{N}_2$ fluxes in energy sorghum, while no significant association was found in the two other plant systems (Fig. S5b). This is in contrast to the laboratory incubation experiments with fresh residue additions or roots of young recently terminated plants (Kim et al. 2020, 2022; Kravchenko et al. 2017). In these studies, N_2O emissions were enhanced by greater presence of $<150\ \mu\text{m}$ pores \varnothing and an associated enhanced creation of hotspots due to the sponge effect of fresh decomposing plant and reveals that in the field the effect of the anaerobic soil volume fraction can counteract the sponge effect. The intact cores of our study contained residues of old as well as young roots as well as POM in a wide range of decomposition stages and thus provided a more realistic assessment of field processes.

The anaerobic soil volume fraction varies under changing water contents at the same time that local hotspots created through POM will change O_2 demand (Kravchenko et al. 2018; Rabot et al. 2015; Rohe et al. 2021; Schlüter et al. 2019). The relationship of bD with the anaerobic soil volume fraction and related structural properties derived from the cores taken in late August is therefore not on all days significant (Fig. S2). This was true, for instance, on 18 June, i.e. the day of the large rain event, when the anaerobic soil volume fraction was potentially much higher as estimated by our image analysis. On this day, however, Γ -indicator, i.e. the connectivity of the pore space, was negatively correlated with the $\text{N}_2\text{O} + \text{N}_2$ fluxes. The potential lower infiltration in plots with low connectivity increased the anaerobic soil volume fraction during the rain event and led, in combination with the higher amounts of available NO_3^- (Fig. S2d), to the burst of $\text{N}_2\text{O} + \text{N}_2$ especially in energy sorghum (Fig. 2c).

Despite these variabilities, our findings highlight the value of our method to connect N_2O production pathways during a crop season to microscale properties. In the future, pore scale modelling could be used to simulate the anaerobic soil volume fraction under contrasting conditions throughout the year and therefore improve the predictability of $\text{N}_2\text{O}+\text{N}_2$ emissions. To summarize, we have discovered a strong correlation between the anaerobic soil volume fraction and the distance of POM to macropores with the average fraction of bD in the bioenergy systems.

How the large roots system of the perennial switchgrass changes the N-cycle.

Switchgrass had lower N_2O emissions compared to maize (Tab. 1), making it a promising biofuel crop option for mitigating climate change due to its low greenhouse gas emissions (Monti et al. 2012). Switchgrass as a perennial crop builds its large root system over several years and then

maintains it, while most of the root system of sorghum and maize degrades shortly after harvest. The large volume of switchgrass roots create a pore structure dominated by large root-holding macropores at the expense of narrow macropores (Tab. 1), leading to larger anaerobic soil volume fraction and consequently high potential for bD (Fig. S6a, Fig. 4a). In addition, large volumes of POM generated from switchgrass massive root system serve as sources for denitrification (Fig. 6a). Thus, despite being supplied with only 1/3 of the N fertilizer, switchgrass showed numerically the highest $\text{N}_2\text{O}+\text{N}_2$ emissions compared to the annual systems (Tab. 1, Fig. 6a). Even though continuous measurements of the gas fluxes would be necessary to obtain unequivocal estimations for the entire season, our intermittently sampled data strongly suggest that switchgrass loses large amounts of the applied fertilizer N as N_2 . Higher $\text{N}_2\text{O}+\text{N}_2$ emissions from switchgrass is a novel finding given the lack of data in the literature for total denitrification.

Despite the high fraction of bD in switchgrass we found no significant association of the anaerobic soil volume fraction and the total $\text{N}_2+\text{N}_2\text{O}$ emitted. One reason could be that we only used 5 cm cores from the topsoil and therefore were not able to include potential subsoil properties guiding denitrification (Shcherbak and Robertson 2019). Another could be that the lower application of urea and the large root system in switchgrass systems enhance the role of the root system's affecting N-availability and consequently $\text{N}_2\text{O}+\text{N}_2$ production (Fig. 5b). Plant roots can modify the N balance in multiple ways; for example, the direct release of exudates can lead to priming effects, and the allocation of C to ectomycorrhizal fungi can increase the mineralization of org-N (Moreau et al. 2019). Such processes would release N in close proximity to switchgrass roots, i.e. the rhizosphere, which is potentially often compacted (Tab.1) and would result in local anaerobic hotspots' driving complete denitrification (N_2O reduction to N_2). Note that the availability of such labile C form also drives denitrification (Baggs 2011; Groffman et al. 1988; Wrage et al. 2004). In addition, switchgrass is known to harness free living N-fixing bacteria (Roley et al. 2021), which could additionally fuel $\text{N}_2 + \text{N}_2\text{O}$ emissions.

After full aboveground development and pollination switchgrass undergoes senescence. During this period remobilization of plant N into roots prior to harvest occurs (Yang and Udvardi 2018). During the sampling of N_2O in late August, when the plant flowered, we found a decreasing importance of bD (Fig. 3a) and an increase in pr (Fig. 3b), while $\text{N}_2\text{O}+\text{N}_2$ fluxes were high (Fig. 2c). Interestingly, during this sampling date the root volume of switchgrass was positively correlated with total fluxes (Fig. S5d, $p<0.1$). Considering that switchgrass roots, when grown in monoculture, are predominantly located in large macropores (Lucas et al. 2023), we can surmise that N_2O formed during the

decomposition of senescent roots with elevated N content (Yang et al. 2016) readily escapes to the atmosphere, leading to the observed increases in p_r . From this, it follows that the N source for N_2O formation has shifted from the anaerobic soil volume fraction/matrix to the roots and their rhizodeposition, which however needs further investigation.

In summary, switchgrass with its massive root system stimulates denitrification both directly, by releasing labile N into the compacted rhizosphere, and indirectly, by forming large amounts of POM at large distances from macropores. This resulted in large N losses through N_2O+N_2 emissions.

The impact of BNI for the investigated sorghum variety

Our expectation was that 4 years of continuous cropping of energy sorghum with its potential for BNI (Subbarao et al. 2007) will lead to greater N_2O production through the denitrification pathway relative to the nitrification pathway by reducing gross nitrification as compared to maize (Nardi et al. 2022). Indeed, energy sorghum seem to be capable of reducing the amount of ammonia oxidizing bacteria under field conditions (Bozal-Leorri et al. 2023). However, during the study season, the denitrification-based N_2O gross production from the soil under energy sorghum was not greater than that from maize (Table 1). Moreover, on none of the sampling dates energy sorghum had significantly higher bD fraction than maize, while three days after fertilization (25th of June) the fraction of bD from sorghum soil was significantly lower compared to maize, suggesting a greater importance of nitrification (Fig. S3a) or fungal denitrification (Fig. 4a) there. The fertilization events with 28% urea-ammonium nitrate fertilizer and the rain event resulted in occasionally relatively high NH_4^+ -concentrations in the soils of both energy sorghum and maize systems (Tab. S2). The presence of NH_4^+ in the rhizosphere could have stimulated the exudation of BNIs (Subbarao et al. 2015). Note that the largest N_2O+N_2 emissions from energy sorghum occurred at the beginning of the season (Fig. 2c), while large effects on gross nitrification by BNIs requires root proliferation as found only later in the growing season. This is because sorgoleone, a major BNI component of sorghum, is hydrophobic and thus likely to be restricted to the rhizosphere (Dayan et al. 2010; Subbarao et al. 2013). But even hydrophilic BNIs that can diffuse into the soil (Gao et al. 2022) would impact nitrification in a limited soil volume if the root system is still under development.

However, the release of BNIs by sorghum can be highly variable depending on genotype (Gao et al. 2022; Sarr et al. 2020; Subbarao et al. 2015; Tesfamariam et al. 2014). As there are, to the best of our knowledge, no detailed analyses of BNIs releases by the variety used in our study (TAM 17651) we cannot draw any conclusions about other sorghum

varieties. Thus, additional research with other varieties by using the isotopic mapping approach is needed to investigate the effect of BNI by sorghum on gross-denitrification on the field. In summary, the importance of bD was already high for all plants throughout the season (Fig. 4a), and we could not identify any effect of potential nitrification inhibition by sorghum, which would have led to even higher importance of bD.

Conclusions

For the first time, we were able to link microscale properties with N_2O+N_2 emissions and production pathways in an agricultural cropping season. We show that the anaerobic soil volume fraction and the distance of POM to macropores derived from X-ray CT are important factors for bacterial denitrification and can be used to cover local variability in N_2O+N_2 emissions in the field. The large changes in these microscale properties measured across and within the investigated plant systems highlight the effect of certain plant species on pore structure and their potential use to mitigate GHG emissions. Although our study reflects fluxes from only a single season, and a multi-year observation period would be needed to assess long-term effects of cropping histories and soil structural differences on N_2O+N_2 emissions in field settings, results nonetheless demonstrate the feasibility and usefulness of N_2O+N_2 monitoring in combination with pore structural analysis in the field. Results also show the value of a future research focus on root traits' leading to changes in pore structure and the distribution of POM, with the potential to reduce N losses by N_2O+N_2 . Finally, our data questions the importance of BNI for the investigated sorghum variety, as we found no evidence for reduced gross nitrification.

Supplementary Information The online version contains supplementary material available at <https://doi.org/10.1007/s00374-023-01761-1>.

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Declarations

Competing interests The authors declare no competing interests.

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Supplementary material

Changes in soil pore structure generated by the root systems of *maize*, *sorghum* and switchgrass affect in situ N₂O emissions and bacterial denitrification

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Table S 1: Isotopic values used for $\delta^{15}\text{N}^{\text{SP}}$ / $\delta^{18}\text{O}$ isotope mapping. Shown are minimum, maximum and mean values for $\delta^{15}\text{N}^{\text{SP}}$ and $\delta^{18}\text{O}$ based on (Yu et al. 2020). In addition, the $\delta^{18}\text{O}$ values are corrected based on precipitation water at the research station KBS.

Process	$\delta^{15}\text{N}^{\text{SP}}$			$\delta^{18}\text{O}$			$\delta^{18}\text{O}$ KBS	$\delta^{18}\text{O}$ corrected		
	min	max	mean	min	max	mean		min	max	mean
ND	-13.6	1.9	-5.9	12.4	19.4	NA	-7.3	5.1	12.1	NA
BD	-7.5	3.7	-1.9	16.7	23.3	19.2	-7.3	9.4	16	11.9
FD	27.2	39.9	33.5	42	55.1	47.2	-7.3	34.7	47.8	39.9
NI	32	38.7	35	20.5	26.5	23.5	-7.3	13.2	19.2	16.2
R	8.9	9.5	9.9	66.9	33.5	50.2	-7.3	59.6	26.2	42.9

Table S 2: Mean values and standard errors for available ammonium (NH_4^+) and nitrate (NO_3^-) for the three investigated plant systems. Different letters indicate significant differences between the plant systems.

Date	NH_4^+ [mg kg ⁻¹]		NO_3^- [mg kg ⁻¹]		Plant	Note
27.05.2021	29.87	±16.94	4.14	±3.8	Z. mays	12 days after seeding
27.05.2021	149.74	±100.46	23.13	±11.83	S. bicolor	8 days after seeding
27.05.2021	56.95	±7.86	19.71	±2.09	P. virgatum	2 weeks after fertilization
18.06.2021	10.81	±3.67	0.38	±0.11	Z. mays	rain event
18.06.2021	27.10	±18.02	9.37	±8.92	S. bicolor	rain event
18.06.2021	41.42	±14.3	2.80	±2.25	P. virgatum	rain event
29.06.2021	10.36	±5.23	2.55	±2.31	Z. mays	7 days after fertilization
29.06.2021	7.83	±2.58	0.29	±0.04	S. bicolor	7 days after fertilization
06.07.2021	50.73	±14.48	47.50	±12.39b	Z. mays	14 days after fertilization
06.07.2021	27.98	±21.2	10.56	±6.34a	S. bicolor	14 days after fertilization
06.07.2021	7.90	±0.55	0.38	±0.12a	P. virgatum	

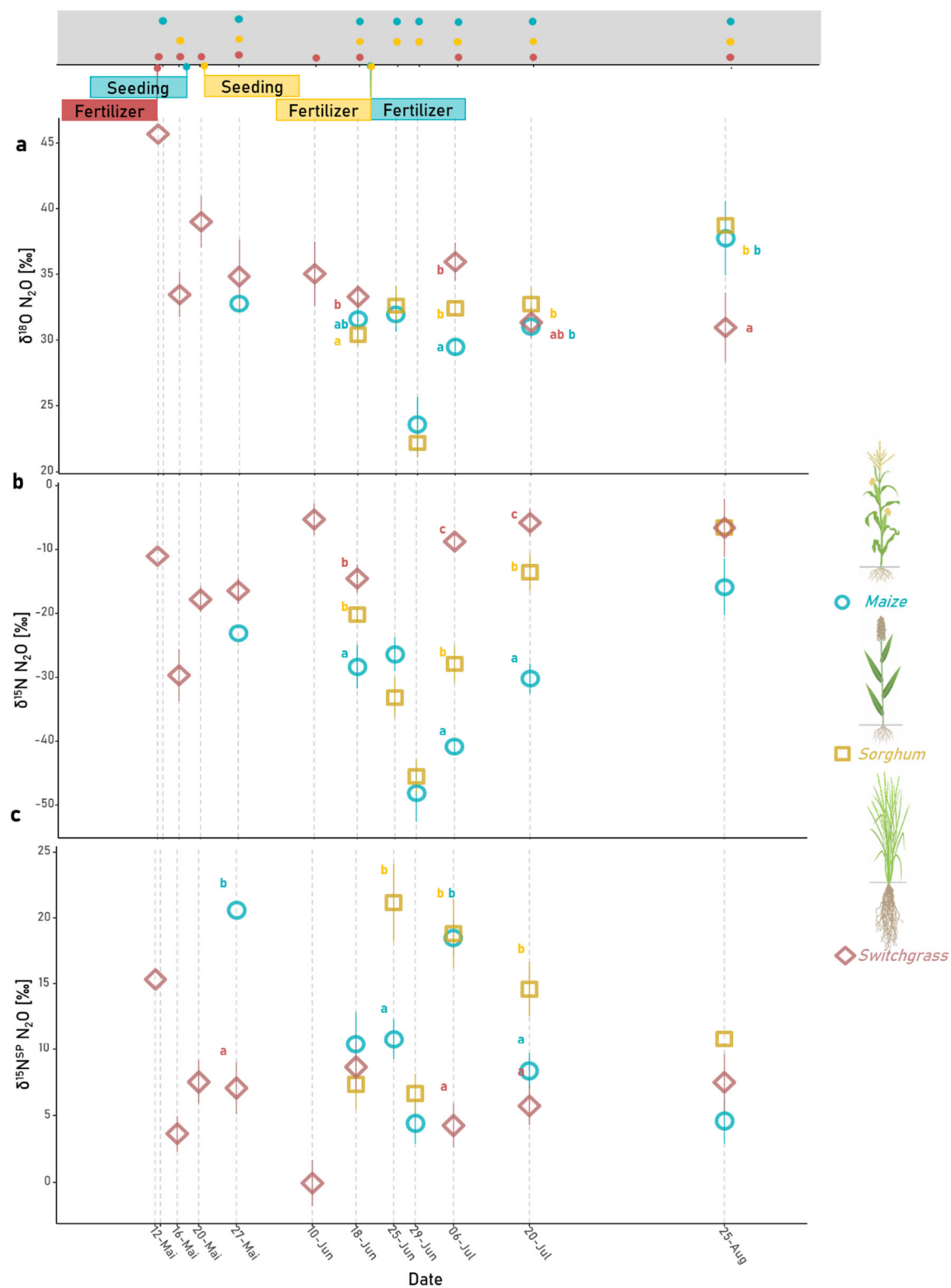


Figure S 1: Aggregated mean values and standard errors of $\delta^{18}\text{O}$, $\delta^{15}\text{N}_{\text{bulk}}$, $\delta^{15}\text{N}_{\text{sp}}$ from N_2O collected at different dates throughout the season 2021. Different letters indicate significant difference between the three crops.

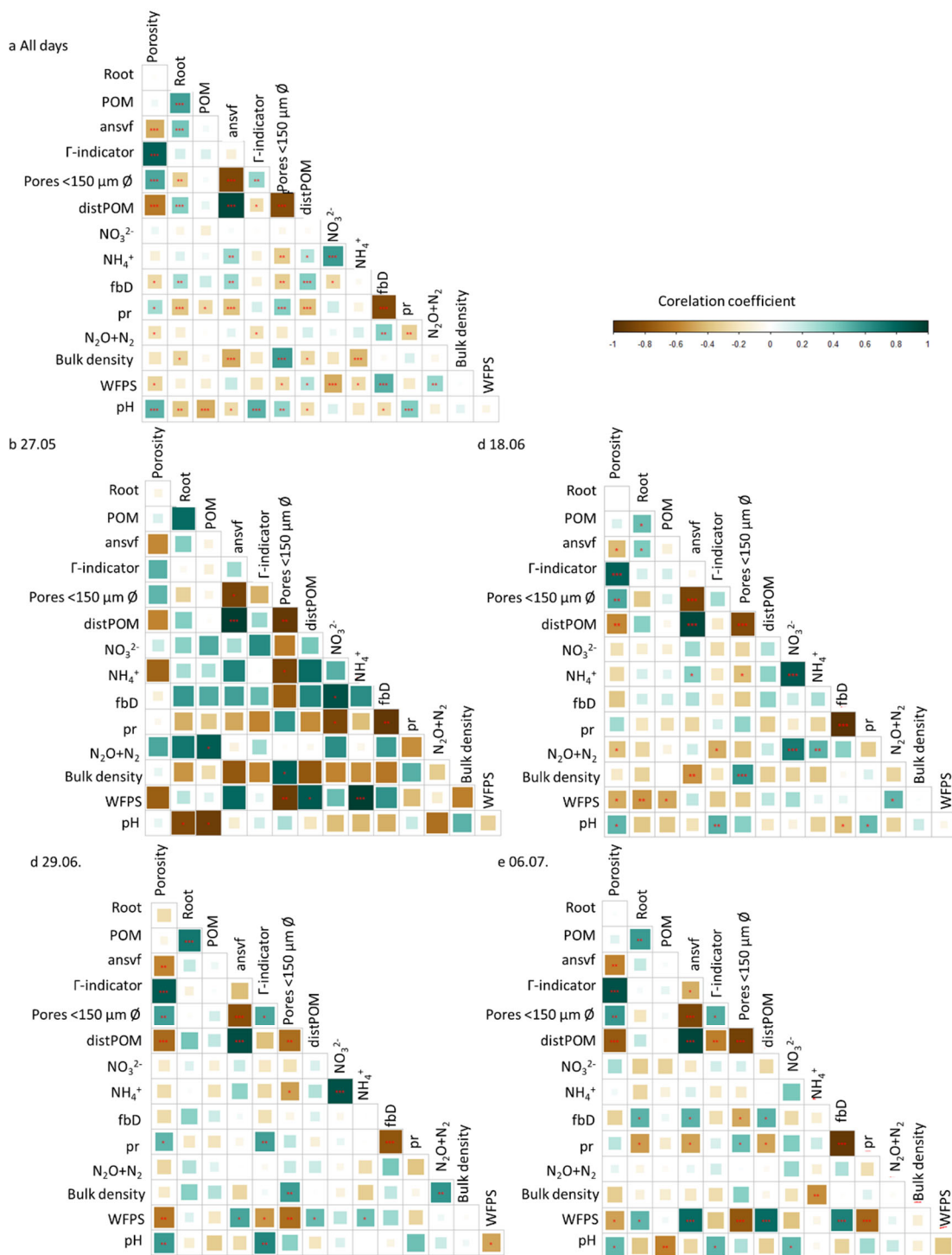


Figure S 2: Pearson correlation coefficients for CT properties and mean values of N parameters throughout the season (a) and at different sampling days (b-e). Red stars indicate significant differences, $p < 0.05$ = *, $p < 0.01$ = **, $p < 0.001$ = ***. The ansvf is the anaerobic soil volume fraction, the distPOM is the mean distance of POM to the next macropore.

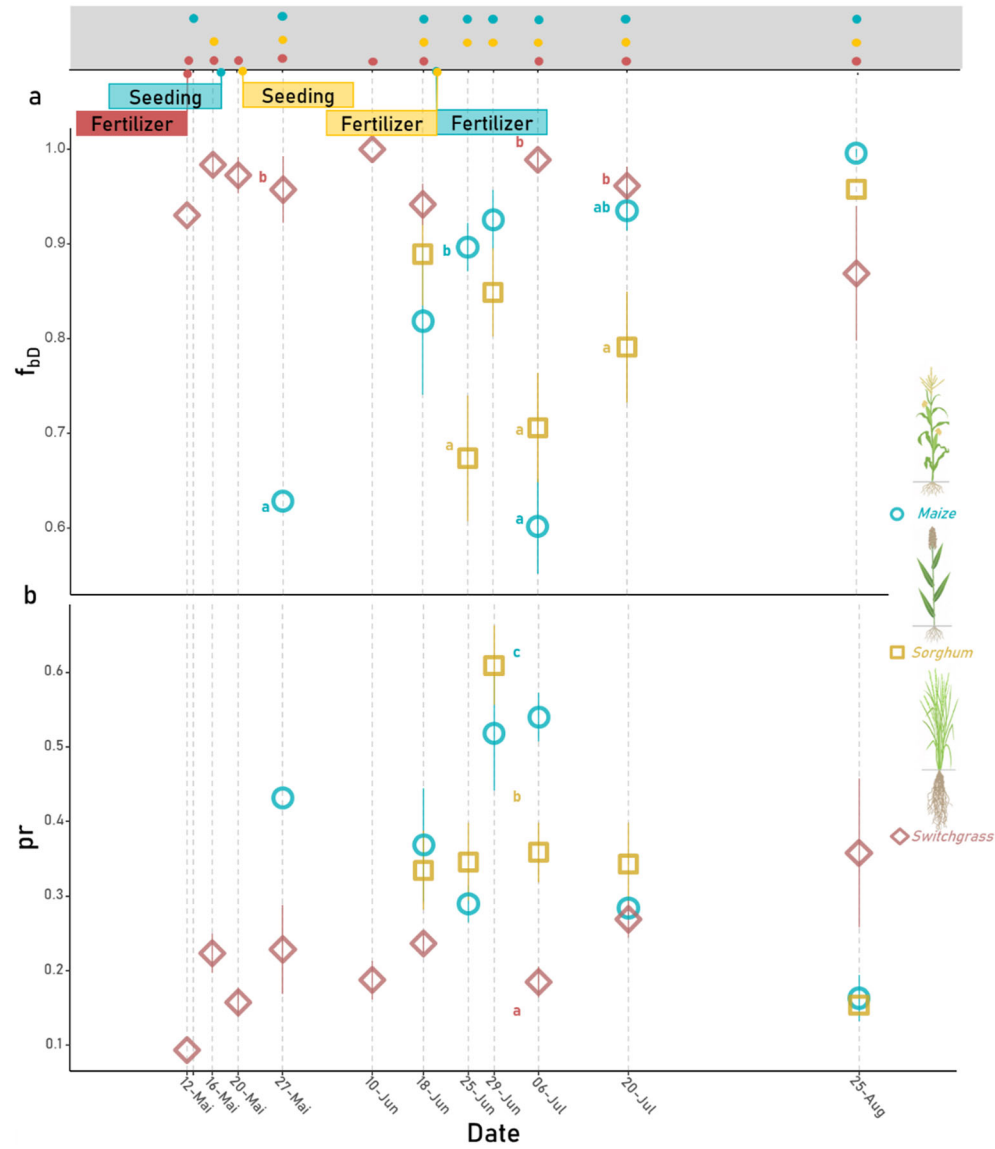


Figure S 3: Fraction of N_2O originating from bacterial denitrification of the total emitted N_2O (f_{BD}) and the denitrification product ratio (pr). The calculation assumes a mixing between bacterial denitrification and Nitrification. Shown are mean values and standard errors of the mean. Different letter indicate significant difference between the investigated plants on the specific day.

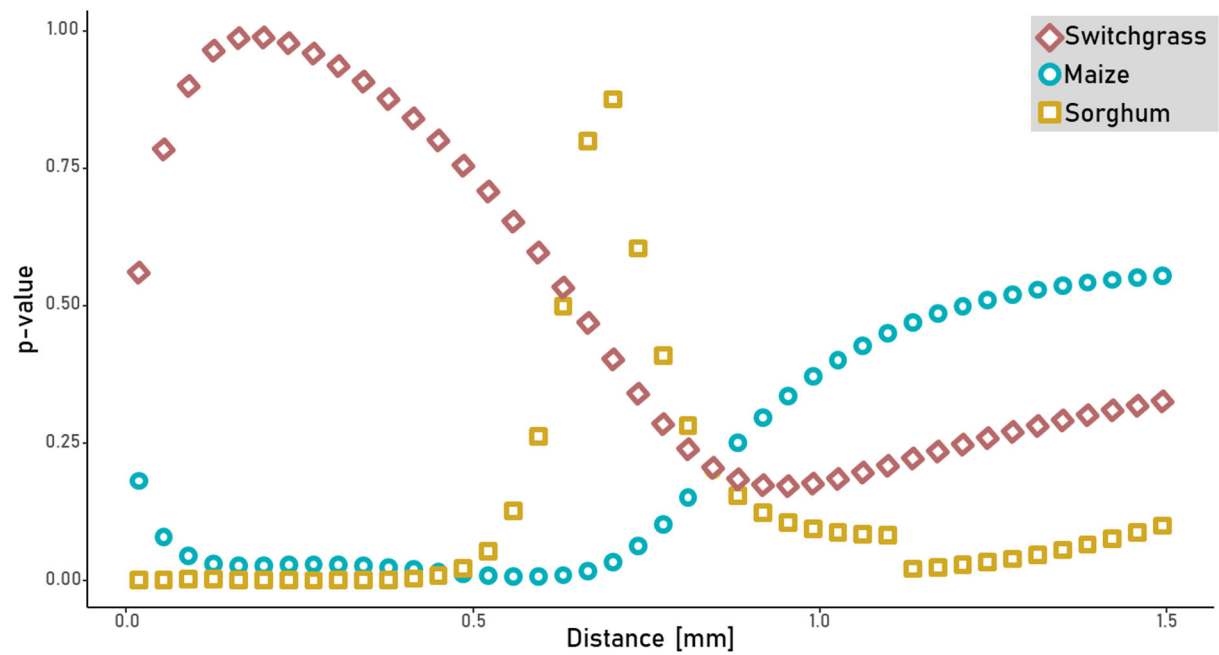


Figure S 4: P-values resulting from the linear correlation of differently computed ansvf with the total N_2O+N_2 emissions. The distance corresponds to the threshold at which all voxel larger than the distance are attributed to the ansvf.

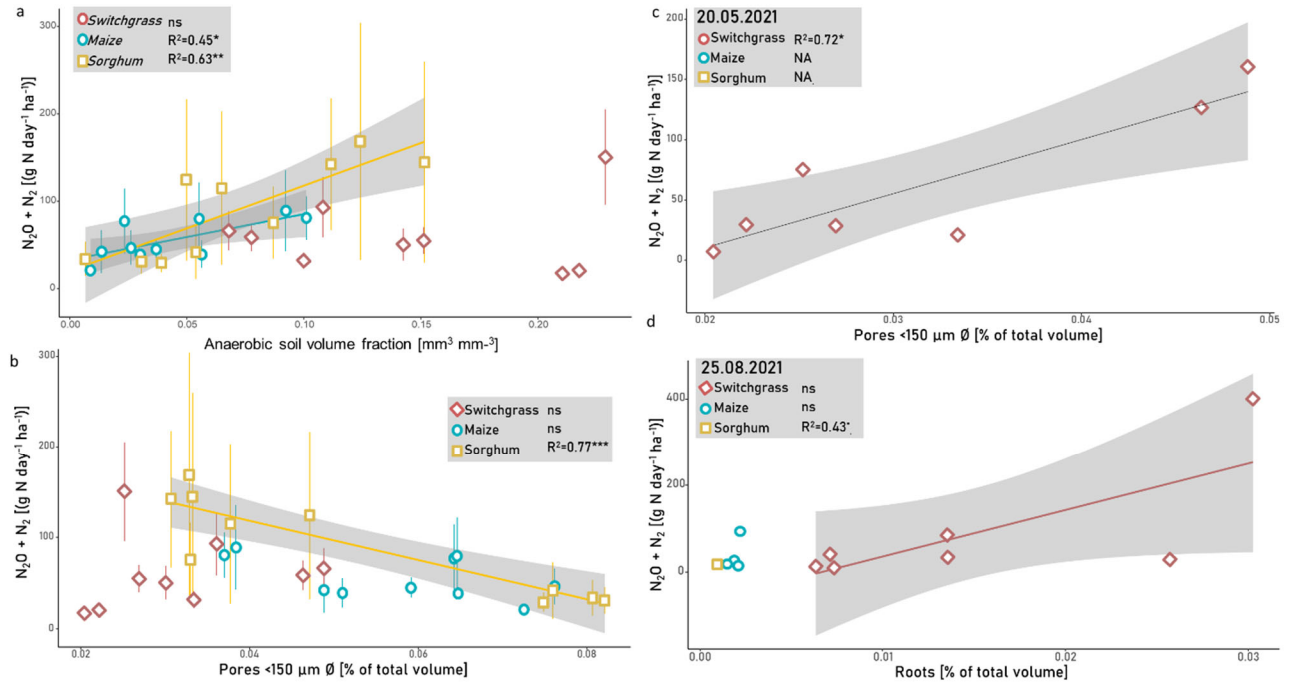


Figure S 5: a) Associations of the anaerobic soil volume fraction (ansvf) and the mean N_2+N_2O emissions, b) associations of the pores $<150\ \mu m$ and the mean N_2+N_2O emissions, c) associations of the pores $<150\ \mu m$ and the total N_2O+N_2 emissions on the 20.05.2021 (7 days after fertilization of switchgrass) and d) association of the root volume with the $N_2O + N_2$ emissions in late August (during senescence of switchgrass). Stars indicate the significant association between the two parameters; $p<0.05 = *$, $p<0.01 = **$, ns = non-significant. Shadows show 95% confidence level interval for the predictions of the linear model.

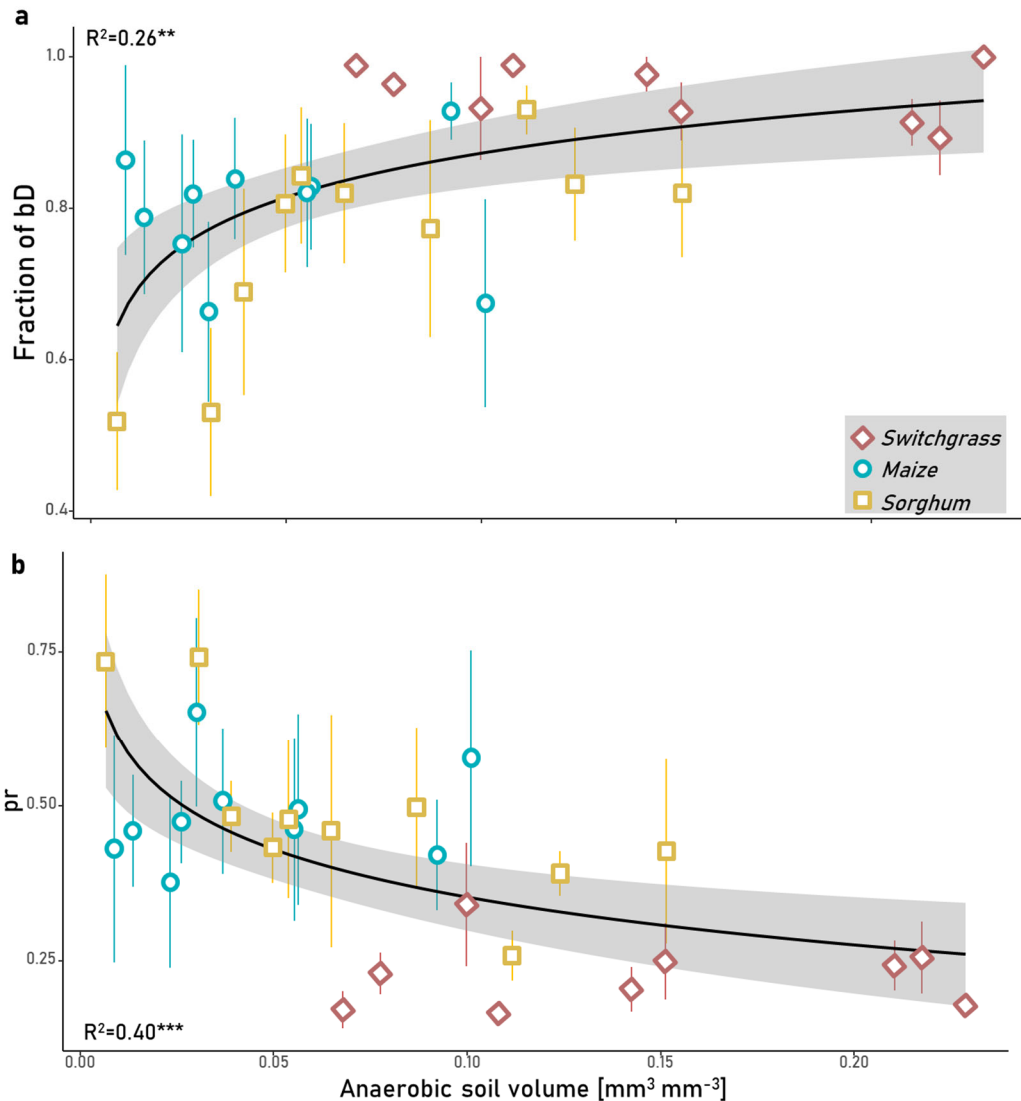


Figure S6: Influence of ansvf on the fraction of bacterial denitrification (bD) and the denitrification product ratio (pr) . Each point represents mean values of fluxes throughout the season for one static flux chamber.

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