Growing-season variability and environmental controls of CH_4 and N_2O fluxes across high-latitude terrestrial ecosystems

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Abstract. Rapid warming and accelerating ecological change across high-latitude ($> 66°$ N) terrestrial environments are reshaping greenhouse gas dynamics. However, spatiotemporal patterns and environmental drivers of CH_4 and N_2O fluxes in high-latitude ecosystems remain poorly understood.

In this study we describe the seasonal and spatial variation in CH_4 and N_2O fluxes across key subarctic and boreal environmental gradients. A secondary aim is to examine the environmental drivers underlying these flux patterns. We measured fluxes during the snow-free season of 2024 and 2025 in northern Finland, starting early June (spring snowmelt) and ending late September (first snowfall). We conducted 1272 chamber measurements at 144 locations spanning eight high-latitude land cover classes: bare tundra, dwarf-shrub tundra, deciduous forest, evergreen forest, fen, bog, permafrost bog and tundra wetland.

Bare tundra, dwarf-shrub tundra, deciduous forest and evergreen forest sites acted as net CH_4 sinks (mean flux of -0.16 to $-0.05\,\mathrm{mg\,m^{-2}\,h^{-1}}$, depending on land cover class) and as both N_2O sources and sinks (-0.69 to $1.69\,\mu\mathrm{g\,m^{-2}\,h^{-1}}$). Fen, bog, permafrost bog and tundra wetland sites were net CH_4 sources (mean flux of 0.93 to $27.81\,\mathrm{mg\,m^{-2}\,h^{-1}}$) and net N_2O sinks (-0.89 to $-6.53\,\mu\mathrm{g\,m^{-2}\,h^{-1}}$). Highest mean CH_4 fluxes occurred in fens ($23.4\,\mathrm{mg\,m^{-2}\,h^{-1}}$) and permafrost bogs ($27.8\,\mathrm{mg\,m^{-2}\,h^{-1}}$) which often coincided with substantial N_2O sinks (mean flux of -6.53 in fens and $-5.40\,\mu\mathrm{g\,m^{-2}\,h^{-1}}$ in permafrost bogs). Seasonal trends were present for CH_4 but were land cover-dependent, with both sinks and sources typically peaking during July–August. Land cover class, soil temperature, and soil pH had the strongest effects on CH_4 fluxes while variation in N_2O was best explained by soil pH and land cover class. However, a significant portion of the variability in N_2O remained unexplained.

These results demonstrate that strong spatial heterogeneity and pronounced seasonal dynamics during the snow-free period generate substantial variability in CH_4 and N_2O fluxes across high-latitude ecosystems. Our findings suggest that monthly measurements of insitu gas fluxes and soil conditions in combination with accurate land cover mapping are essential for modelling and upscaling of GHG fluxes. This is especially important in high-latitude regions, where rapid warming is altering both ecosystem distributions and soil conditions with significant consequences for current and future GHG budgets.

1 Introduction

High-latitude ($> 66°$ N) ecosystems exhibit highly dynamic greenhouse gas (GHG) fluxes shaped by the interactions of soil, microbial, and vegetation processes (Whalen, 2005; Riutta et al., 2007; Olefeldt et al., 2013; Martins et al., 2017). Climate change can alter all these factors by reshaping abiotic conditions such as precipitation, temperature, and snow cover. Changing climate also drives biological shifts, including changes in microbial communities (Deslippe et al., 2012) and broader ecosystem transitions such as shrubifica-

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tion, paludification, and permafrost thaw (Myers-Smith et al., 2011; Biskaborn et al., 2019). Arctic regions store a major fraction of global soil carbon reserves (Batjes et al., 2016 TS5; Hugelius et al., 2020) and are warming faster than the global average (Rantanen et al., 2022), underscoring their importance in the global carbon cycle. However, the interplay of complex GHG processes and the heterogeneous landscape makes both measurements and modelling particularly challenging.

Soil carbon storage and GHG dynamics vary widely across high-latitude ecosystems (Hugelius et al., 2024). Peatlands (wetlands with peat soils), for example, function as major long-term carbon sinks due to ongoing peat accumulation, yet they are also the largest natural source of atmospheric methane (CH_4) because of their persistently anoxic soils (Saunois et al., 2025; Kübert et al., 2026). Most high-latitude wetlands can be further categorized into bogs, permafrost bogs and fens based on their hydrology and permafrost extent (Olefeldt et al., 2021). We have additionally included a tundra wetland category as both their ecology and hydrology significantly differ from the other categories (see Sect. 2.3 for detailed land cover classifications). In addition to wetlands, we also studied dry tundra and forest soils which were further categorized into bare tundra, shrub tundra, evergreen forests and deciduous forests (see Sect. 2.3). In this study, dry tundra soils specifically refer to non-wetland upland mineral soils with no tree cover. Although dry tundra and forest ecosystems store less carbon per unit area than peatlands, their vast extent, up to 80 % of the boreal-Arctic area (Raynolds et al., 2019; Olefeldt et al., 2022 TS6), means they exert a major influence on high-latitude carbon cycling through CH_4 uptake, biomass and soil carbon storage, and gross-primary production (GPP) (Virkkala et al., 2021; Voigt et al., 2023; Li et al., 2025; Jiang et al., 2025). In addition to CH_4 dynamics, nitrous oxide (N_2O) sources and sinks have been reported from high-latitude ecosystems, but their magnitudes, spatial patterns, and environmental drivers remain poorly understood (Chapuis-Lardy et al., 2007; Maruschak et al., 2011 TS7; Voigt et al., 2020). CH_4 and N_2O are the second and third most influential greenhouse gases after CO_2. Over a 100-year time horizon, CH_4 has a global warming potential (GWP100) that is 27–29 times greater than that of CO_2, whereas N_2O has a GWP100 of 273 (IPCC, 2021 TS8).

High latitude peatlands have a disproportionately strong effect on carbon dynamics due to their intense CH_4 emissions and large soil carbon stocks and because of that, they are overrepresented in many flux databases (Kuhn et al., 2021; Virkkala et al., 2025 TS9). Yet, high-latitude landscapes are heterogenous with a mosaic composition of wetlands, forests and tundra of which tundra soils are especially underrepresented in literature. Furthermore, it is typical to focus on peak growing season fluxes (Olefeldt, 2013 TS10, Kuhn et al., 2021) as it makes measurements and modelling more straightforward. However, this ignores the strong seasonal dynamics of the environmental drivers in high-latitude ecosystems which control GHG fluxes (Rautakoski et al., 2024; Kübert et al., 2026). Multi-month measurements across ecosystems are therefore required to resolve the seasonal progression of GHG fluxes and their underlying controls, which can shift substantially across environmental gradients and between spring, summer peak productivity, and autumn transition.

This study aims to characterize the seasonal and spatial variability of CH_4 and N_2O fluxes across key subarctic and boreal land cover classes, focusing on representative forest, wetland, and tundra ecosystems. A secondary aim is to examine the key environmental drivers and their role in shaping these flux patterns. To capture ecosystem-level heterogeneity, we classify these biomes into eight land cover classes following the BAWLD dataset framework (Olefeldt et al., 2021; Kuhn et al., 2021): deciduous and evergreen forests, bare tundra, dwarf-shrub tundra, fens, bogs, permafrost bogs and tundra wetlands (see Sect. 2.3 for details). Specifically, we address three research questions: (i) How do CH_4 and N_2O fluxes differ among the land cover classes? (ii) How do these fluxes develop seasonally during the snow-free period (June–September)? (iii) How do various environmental variables explain the variance in GHG fluxes? The analysis is based on 1273 in situ chamber measurements collected at 144 locations spanning eight high-latitude land cover classes at two research sites in Northern Finland (Kilpisjärvi and Pallas). The measurements cover the full duration of growing seasons of years 2024 and 2025, from spring snowmelt to late autumn.

2 Materials and methods

2.1 Location and area descriptions

The two study locations are in the Pallas-Yllästunturi national park in the Kittilä and Muonio municipalities in western Finnish Lapland, and the Kilpisjärvi region in the Enontekiö municipality, in northwestern Finnish Lapland. The Pallas study sites (67.9° N, 24.0° E, 250–570 m a.s.l.) are located at the edge of the northern-boreal and subarctic climatic zones with a long-term mean annual temperature of −1.0 °C and annual precipitation of 644 mm (Pirinen et al., 2012). It is characterized by a mosaic of northern boreal ecosystems consisting of mixed and coniferous forests (*Picea abies*, *Pinus sylvestris*, *Betula pubescens*), open fen wetlands and open tundra on the fells. The evergreen forest sites are only found in the Pallas area (Fig. 1, map 2) as the Kilpisjärvi sites are located above the coniferous forest limit. The Kilpisjärvi (Gilbbesjávri in Northern Sámi language) study sites are located in the subarctic climate zone (69.0° N, 20.8° E, 474–737 m a.s.l.) with a long-term mean annual temperature of −1.9 °C and annual precipitation of 487 mm (Pirinen et al., 2012). The Kilpisjärvi region lies at the ecotone between the boreal and tundra biomes and is

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characterized by dwarf-shrub dominated tundra highlands, *Carex* spp. dominated open fens and short stature mountain birch (*Betula pubescens* var *pumila*) forests at lower altitudes. The permafrost bog, tundra wetland, and deciduous forest sites occur only in the Kilpisjärvi research area (Fig. 1a and b). The Pallas area lies too far south and at too low an elevation to support permafrost bogs and extensive mountain birch forests, and comparable tundra wetlands are absent from the vicinity of the Pallas research area. The Pallas area is underlain by acidic Precambrian bedrock, resulting in low soil pH, whereas Kilpisjärvi is characterized by a geological mosaic of Paleozoic sedimentary rocks and acidic Precambrian bedrock (Simonen, 1990), yielding intermediate soil pH conditions. This geological contrast provides a robust framework for assessing the influence of pH on CH_4 and N_2O fluxes. Snowmelt and associated flooding and nutrient pulses typically occur in June, with rapidly increasing daylight hours progressing into a ~ 2-month period of continuous daylight from late-May to late-July. Afterwards, a senescing period with occasional sub-zero temperatures occurs already during September. All measurement areas are relatively undisturbed with no ditching, recent logging or other major human influences.

2.2 Field campaign

The measurement campaign was conducted in two parts (years 2024 and 2025). The first part took place between 9 June and 1 October (2024) and the second part between 3 June and 21 September (2025). A total of 144 measurement plots were repeatedly sampled once a month (June–July–August–September) each year. Additional repeated measurements were taken from roughly a third of the research plots. In total, we conducted 1273 chamber measurements (n = 1273 for each GHG) while continuously measuring microclimate and green vegetation cover at each measurement plot. Total chamber measurements were distributed between land cover classes as follows: shrub tundra (n = 266, 20.9 %), fen (n = 248, 19.5 %), evergreen forest (n = 186, 14.6 %), deciduous forest (n = 180, 14.2 %), bare tundra (n = 106, 8.3 %), tundra wetland (n = 106, 8.3 %), bog (n = 100, 7.9 %) and permafrost bog (n = 80, 6.3 %).

2.3 Land cover classification and site descriptions

The measurement plots were equally distributed between the Pallas (boreal) and Kilpisjärvi (subarctic) regions with three main ecosystem types of dry tundra, forests and wetlands, which were further divided between eight different key land covers: bare tundra, shrub tundra, evergreen forest, deciduous forest, tundra wetland, bog, fen and permafrost bog. These land cover categorizations are based on the BAWLD dataset (Olefeldt et al., 2021), with some minor changes to more accurately describe our research sites. Specifically, the differences to the BAWLD categorization: our tundra wetland sites are not permafrost affected, and we separated dry upland soils into bare tundra and shrub tundra as we expected them to have different GHG flux profiles due to the lack of vegetation cover and shallower organic layer thickness in bare tundra plots.

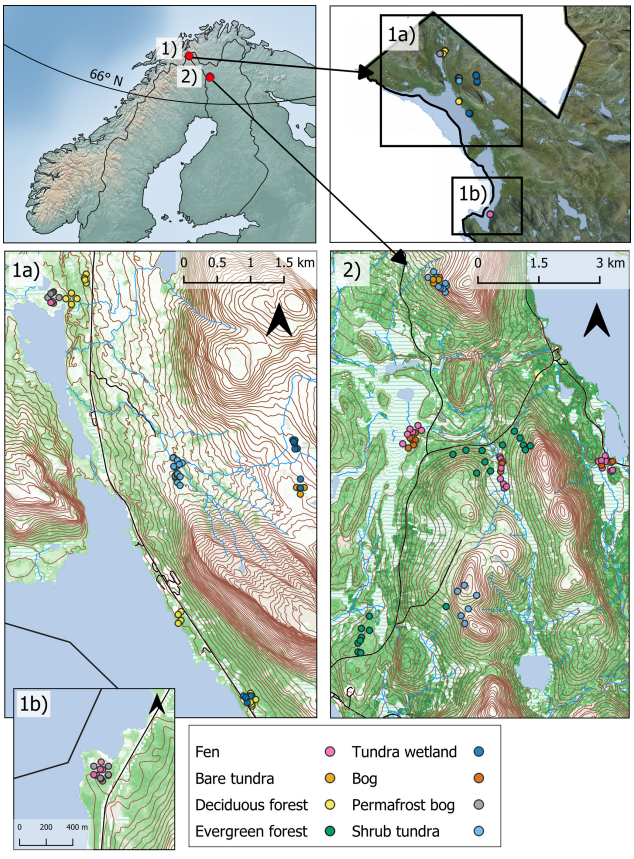


Figure 1. Maps of the two research sites. Subarctic Kilpisjärvi site (1a and 1b) and boreal Pallas site (2). Colored dots indicate the measurement plots (n = 144) and their eight different land cover classifications.

The following descriptions are specifically for our research sites and based on our in situ observations. Shrub tundra plots (n = 28) are located above the tree line and typically have a fairly thin dry organic soil layer (mean organic layer thickness, OLT = 5.0 cm). Vegetation is generally dominated by *Betula nana*, *Empetrum nigrum* and *Vaccinium myrtillus*, although vegetation can occasionally be more diverse in wind protected valleys and creeks. Bare tundra plots (n = 11) lack vascular plants and typically result from repeating freeze-melt cycles (cryoturbation) and persistent winds at higher altitudes. These plots may occasionally support small amounts of lichen and mosses and have a thin OLT (mean OLT = 3.2 cm).

Evergreen forest plots (n = 21) are located at the northern edge of the boreal zone with a tree cover dominated by *Picea abies* or *Pinus sylvestris* but can also have some deciduous trees mixed in, i.e. *Betula pubescens or Pop-*

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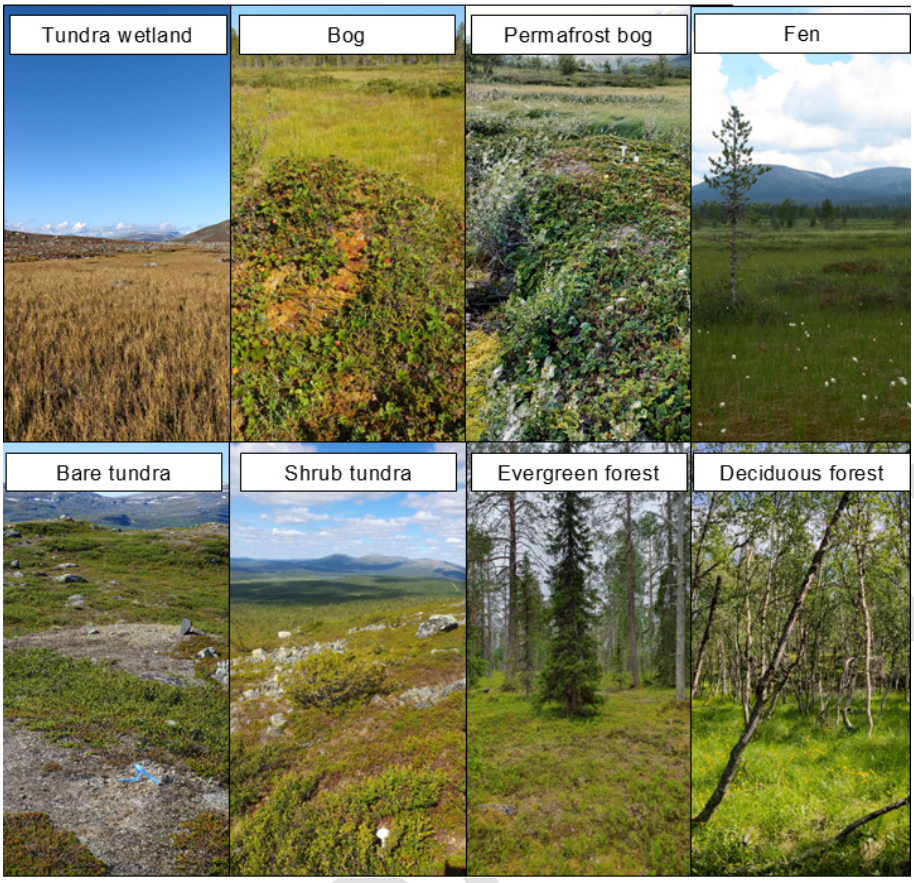


Figure 2. Representative in situ photographs of the eight land-cover classes included in the study, illustrating the range of environments sampled across the research areas (Photos: Johannes TS11 Niemi, S. TS12 Mäki-Ikola). TS13

ulus tremula. Field layer is dominated by dwarf-shrubs (e.g. *Vaccinium myrtillus*, *Vaccinium vitis-ideae*) and various graminoids (e.g. *Avenella* spp. and *Poa* spp.). Evergreen forests had the highest OLT of the dry land cover classes (mean OLT = 9.82 cm). The deciduous forest plots ($n = 20$) are primarily located at the edges of the forest limit, and the tree cover consists solely of short growing *Betula pubescens.* var. *pumila* (mountain birch). The field layer is mixed with various dwarf-shrubs, graminoids (*Avenella* spp.) and forbs (e.g. *Cornus suecica*, *Viola palustris*). Most deciduous forests occur on well-drained glaciofluvial eskers and therefore have shallow organic layers (mean OLT = 2.7 cm).

Fen plots ($n = 31$) are minerotrophic, wet and treeless wetlands with peat depths up to ~ 4.4 m. Field layer is dominated by *Carex* spp. and *Sphagnum* spp. Bog plots ($n = 11$) are all of the dry and "hummocky" (permafrost free) wetland forms, i.e. if a wetland as a whole would be considered a fen, it might still have elevated dry Sphagnum hummocks with bog-type shrub vegetation which have considerably different greenhouse gas profiles compared to the wetter fen-type areas. Permafrost bog plots ($n = 10$) are similar to the previously mentioned bogs but with permafrost cores that are in various stages of degradation. The permafrost bog research sites include two different peatlands representing varying degrees of permafrost degradation: one with largely intact but degrading palsas, the other with only remnants of old permafrost left. Permafrost extent was confirmed using soil cores to distinguish permafrost bogs from bogs without permafrost. Lastly, tundra wetland plots ($n = 12$) are *Carex* spp., *Trichoforum* spp. and moss dominated wetlands (note: generally, no *Sphagnum* spp.). The tundra wetland sites are situated on hillsides, with small creeks running through them. They have a shallow organic soil layer compared to the other wetland types (mean organic layer thickness, OLT = 11.2 cm). The upper soil layer consists of a mixture of fluvial minerogenic and organic deposits, including mud, gyttja, litter, peat, and roots. They have variable seasonal hydrology, i.e. water table was typically above the soil surface during spring snowmelt and autumn rains but below the surface during summer. Despite the low summer water table, soil moisture remained high throughout the measurement campaigns.

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2.4 Environmental variables

The field layer vegetation cover of each plot was estimated by image analysis with Canopeo (Patrigani and Ochsner, 2015TS15). Canopeo measures the relative number of green pixels from a photo and converts them into percentages, i.e. it acts as a measurement for green vegetation cover ("vegetation cover"). A photo was taken of each measurement plot every time chamber measurements were conducted. The pictures covered roughly a 1 m^2 square around the measurement plot and thus included vegetation from outside the chamber collar.

Microclimatic variables were measured with TOMST TMS-4 sensors (Wild et al., 2019), which are located 0.5–1 m from the chamber collar and measure continuously throughout the year. Loggers record microclimatic variables in 15 min intervals: soil temperature at a depth of −6 cm and soil moisture at depth of −14 cm. Each logger operated with a temperature resolution of 0.0625 °C and an accuracy of ±0.5 °C. For the soil temperature, the data quality control and processing procedures followed the approach outlined in Aalto et al. (2022). First the temperature time series were visually inspected to remove readings that clearly reflected sensor malfunction, such as abrupt shifts in soil moisture or temperature indicative of the logger being displaced. After the first screening of the data, the temperature sensors were then calibrated based on their individual response differences when stored under stable conditions (i.e. data logged while in storage indoors). Finally, any abnormal spikes in the data were filtered out, for example, a data point was flagged as erroneous if soil temperature change between consecutive time steps exceeded 5 °C or change in air temperature exceeded 10 °C.

For the soil moisture data, the quality control and preprocessing of the data followed the same outline as in Kemppinen et al. (2023). Similarly to the temperature time series, the raw soil moisture data was screened manually for clearly erroneous data where values were removed if the logger had been displaced, or the raw soil moisture value was < 200, as it is outside the range of the sensor and usually indicates of poorly placed logger. The TOMST logger records raw soil moisture data based on electromagnetic pulses, producing time-domain transmission measurements (see Wild et al., 2019, for details). These were subsequently converted to volumetric water content (VWC, %) using the generic calibration function for unknown soils by Kopecký et al. (2021). The function for unknown soils was chosen, as it provided more realistic ranges of VWC values compared to soil-specific calibrations, however it does limit the maximum possible VWC to 60 %. This method has already been validated in Finland by Kemppinen et al. (2023), supporting its applicability to this research. After quality control and preprocessing, microclimate data was coupled with flux measurements by selecting the nearest 15 min logger reading corresponding to the flux measurement time. In cases where the closest time point contained erroneous or missing data, the next closest valid reading was used.

Both water table depth and soil moisture have been identified as key drivers for GHG fluxes (Moore and Dalva, 1993; Olefedt et al., 2013TS16; Rinne et al., 2018; Voigt et al., 2020; Virkkala et al., 2024). We chose to use soil moisture over water table depth for three reasons: (a) determining water table depth in mineral soils typically requires invasive methods that disturb soil structure and alter hydrological conditions (b) water table in dry tundra and forest sites is generally very deep, thus it has minimal impact on biogeochemistry occurring at the surface or at intermediate depths (c) we wanted model parameters that can be measured at all sites (i.e. both mineral soils and wetlands) for robust modelling across ecosystem gradients.

Soil samples for pH data were collected once during the measurement campaign from the organic layer underneath the vegetation (outside the chamber collars). The laboratory analysis was conducted by the Finnish Natural Resources Institute (LUKE) according to the ISO 10390 soil Quality-Determination of pH standards, in a suspension of soil in water (pH in H_2O).

2.5 Chamber measurements and flux calculations

The gas measurements were conducted as closed opaque chamber measurements with a duration of 3 min, taken with two portable gas analyzers (LI-COR-LI7810 and LI-COR-LI7820) that measure gas concentrations in 1 s intervals. Gas fluxes for CH_4 and N_2O were both measured simultaneously from the same chamber.

The chamber had a built-in fan and a wireless Bluetooth temperature sensor (Ruuvi tag pro) for measuring chamber temperatures. The volume of the chamber was ∼ 7.85 L (cylindrical, 20 cm × 25 cmTS17) and was placed on top of a steel collar. As the collar height changed between measuring plots (depending on how soft and deep the soil was), the height was always measured separately for each gas measurement to get accurate chamber volumes. For wetlands, the collars were installed into the soil roughly 7–14 d before the measurement campaign started to avoid disturbances in the peat that could cause ebullition or other artefacts in the flux. For mineral soils, the collar was removed between measurements to minimize changes in microclimatic conditions such as sheltering from wind. If unusually high fluxes were measured, they were repeated multiple times in situ to make sure the emissions were not caused by any influence of the measurer (i.e. ebullition due to the pressure on surrounding soil). If the repeat measurement showed similar fluxes, then the initial measurement was used (assuming linear change of gas concentration).

Fluxes were calculated as shown in Eq. (1), based on a linear fit between starting gas concentration at the start of the measurement and at the end of the measurement, utilizing the

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ideal gas law

$$\text{flux} = \frac{(ct - c0) \cdot V \cdot M \cdot p}{t \cdot R \cdot (T + 273.15) \cdot A} \quad (1)$$

where $(ct - c0)/t$ is the slope (expressed as ppm s^{-1}), V is chamber volume, M is molar mass, p is atmospheric pressure, R is the ideal gas constant, T is chamber temperature and A is chamber area. Output of the formula without conversions is µg m^{-2} s^{-1}. Fluxes were converted from s^{-1} to h^{-1} (multiply by seconds in an hour; 3600 s) and from micrograms to milligrams for CH_4 (by dividing by 1000). Final units used for fluxes are mg m^{-2} h^{-1} for CH_4 and µg m^{-2} h^{-1} for N_2O.

Each chamber measurement was visually graphed to make sure the gas concentrations were changing linearly and to remove any possible measurement errors, such as anomalies at the start and end of the measurements that come from pressure changes or from handling the chamber (Pavelka et al., 2018). The same flux calculation method has been used in previous studies by Rissanen et al. (2023), Karjalainen et al. (2025) and Suominen et al. (2026).

Python script for flux calculations is also available: https://github.com/janivaltteri/djp-mcds/blob/main/djpmcds/mcds/fluxcalc.py TS18 (last access: TS19).

We followed Nickerson (2016) to calculate a minimum detectable flux (MDF) for chamber measurements which is an extended formula from Christiansen et al. (2015) TS20 specifically designed for continuous high-resolution laser-based measurements. The calculation is based on the LI-COR sampling frequency (1 Hz) and measurement precision (0.6 ppb for CH_4 and 0.4 ppb for N_2O as reported by LI-COR), chamber volume, closure time and ideal gas law. For CH_4, 1.18 % of measurements were under the MDF (i.e. not confidently distinguishable from zero) while 10.6 % were below the MDF for N_2O.

Clear measurement errors (i.e. flat zero measurements that are sometimes caused by LI-COR malfunction) were removed, but all other measurements which had linear gas concentration change were kept in the data. Following quality control, 15 N_2O measurements (0.12 % of all N_2O measurements) and 9 CH_4 measurements (0.01 % of all CH_4 measurements) were excluded from the analyses. Non-linear fluxes are typically caused by the increasing concentration gradient at high gas concentrations, which slows down the gas diffusion rates from the soil to the chamber causing a log-curve (Pihlatie et al., 2013). In such cases, only the linear part of the measurement was used for the calculation to avoid underestimating fluxes. Chamber measurements that showed exponential concentration growth were typically repeated in situ until a linear trend was observed, as this kind of flux behavior is usually caused by influence of the measurer (Pavelka et al., 2018).

2.6 Statistical analysis and modelling

We used the *lme4* package (Bates et al., 2015) in R (R Core Team, 2024 TS21) to fit linear mixed-effects models examining the relationships between GHG fluxes and environmental variables. We used a joint model for both wetlands and dry land cover classes to investigate useful parameters for modelling across broad environmental gradients. The final model for CH_4 consisted of four fixed effects, three two-way interactions terms and a random effect. As the main effects, we included soil temperature (continuous), soil moisture (continuous), pH (continuous) and land cover (categorical, 8 levels). Interaction terms were land cover paired with each of the other fixed variables. The land cover interaction term was used to give the model flexibility for the direction of the GHG flux response, e.g. increasing temperature can have both positive and a negative effect on GHG flux depending on the land cover class. Measurement point identification was used as a random effect (random intercept) to account for the repeated measurements design (e.g. pseudoreplication). Green vegetation cover (continuous) and measurement year (categorical, 2 levels) were also tested as fixed effects but were dropped due to worse model performance.

The final model for N_2Oconsisted of five fixed effects, two two-way interactions and a random effect. As main effects we included soil temperature (continuous), pH (continuous), Green vegetation cover (continuous), year (categorical, 2 levels) and land cover (categorical, 8 levels). Interaction terms were land cover paired with pH and vegetation cover. Measurement point identification was used as a random effect (random intercept). Soil moisture, soil moisture-land cover interaction and soil temperature-land cover interaction were also tested as fixed effects but were dropped due to worse model performance.

Organic layer thickness (OLT) was considered as a potential explanatory variable but was not included in the final models as it was strongly correlated with land-cover class and soil moisture. Because these variables already captured much of the variation associated with OLT, its inclusion would have increased collinearity and model complexity without providing substantial additional explanatory power.

Model validation was done through Akaike information criterion (AIC) and model parameter effects sizes were described with partial eta-squared (η^2p). The GHG flux data was skewed and included both negative and positive values and as such was normalized with a YeoJohnson-transformation (Yeo and Johnson, 2000) for CH_4 and with arcsinh-transformation for N_2O. Optimal data transformation options were investigated with the *bestNormalize* R-package (Peterson, 2021). Statistical differences between months and between land cover classes were tested with Aligned Rank Transformed Anova, (ART-Anova, Wobbrock et al., 2011) using the *ARTool* R-package (Kay et al., 2025). This allows testing for differences between groups (months or land cover class) while accounting for repeated measurements and un-

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even group sizes. Post-Hoc tests were done following the ART-C contrast testing procedure (Elkin et al., 2021) with Tukey's HSD using the *art-contrasts* R-package (Elkin et al., 2021). Differences between groups were considered statistically significant at $p < 0.05$.

3 Results

3.1 CH_4 between land cover classes

TS22 All dry land cover classes (shrub and bare tundra, deciduous and evergreen forest) were small net sinks (mean $-0.12\,mg\,m^{-2}\,h^{-1}$) for CH_4 and conversely all wetland classes were net sources (mean $17.1\,mg\,m^{-2}\,h^{-1}$) of CH_4 (Table A1), fluxes across all measurements varied from -0.55 to $1008\,mg\,m^{-2}\,h^{-1}$ (Table A1). Highest positive mean CH_4 fluxes were measured from permafrost bogs (mean $27.8\,mg\,m^{-2}\,h^{-1}$; range -0.24 to 602), followed by fens ($23.4\,mg\,m^{-2}\,h^{-1}$; 0.13 to 1008), bogs ($10.1\,mg\,m^{-2}\,h^{-1}$; -0.11 to 668) and tundra wetlands ($0.93\,mg\,m^{-2}\,h^{-1}$; -0.04 to 7.59). Largest negative mean CH_4 fluxes (sinks) were measured from shrub tundra ($-0.16\,mg\,m^{-2}\,h^{-1}$; -0.55 to 0.001) followed by bare tundra ($-0.13\,mg\,m^{-2}\,h^{-1}$; -0.52 to 0.01), deciduous forests ($-0.12\,mg\,m^{-2}\,h^{-1}$; -0.50 to 3.02) and lastly evergreen forests ($-0.045\,mg\,m^{-2}\,h^{-1}$; -0.40 to 0.57). Although both deciduous and evergreen forest were net sinks, they also included some small positive fluxes which all came from locations in ecotones adjacent to wetlands or streams.Pairwise comparisons showed that 21 out of 28 land cover classes differed significantly in CH_4 fluxes (Table A7).

3.2 CH_4 temporal trends

Across all land cover types, both negative and positive fluxes were small in early June, increased to their highest values during July–August, and declined again in September. However, the magnitude of monthly differences was dependent on land cover type. Monthly mean fluxes can be found in Table A2 and statistical differences between months in Table A5. Wetland sites remained net sources throughout the measurement period, with positive fluxes persisting until the end of the growing season. (Fig. 3a, Table A2). No significant effect of measurement year on CH_4 fluxes was detected. Within-plot variability across months was often high, but extreme values typically originated from the same few measurement locations. For example, one permafrost bog plot, across multiple months, produced repeated CH_4 outliers of 50, 116, 366, and $602\,mg\,m^{-2}\,h^{-1}$, resulting in a within-plot range of over $500\,mg\,m^{-2}\,h^{-1}$.

3.3 N_2O between land cover classes

N_2O showed high variance within land cover classes, with fluxes ranging from -141 to $59.5\,\mu g\,m^{-2}\,h^{-1}$ across all sites. Notably, all wetland types were on average, net sinks of N_2O, although all land covers exhibited both positive and negative fluxes (Fig. 3b). Positive mean N_2O fluxes were observed in deciduous forests (mean $1.70\,\mu g\,m^{-2}\,h^{-1}$; range -15.6 to 39.2) and shrub tundra ($0.44\,\mu g\,m^{-2}\,h^{-1}$; -36.9 to 23.0). The strongest negative mean fluxes (net sinks) occurred in fens ($-6.53\,\mu g\,m^{-2}\,h^{-1}$; -142 to 5.13) and permafrost bogs ($-5.40\,\mu g\,m^{-2}\,h^{-1}$; -92.8 to 16.6). N_2O sinks were also measured in bogs ($-3.51\,\mu g\,m^{-2}\,h^{-1}$), tundra wetlands ($-0.89\,\mu g\,m^{-2}\,h^{-1}$), bare tundra ($-0.69\,\mu g\,m^{-2}\,h^{-1}$), and evergreen forests ($-0.52\,\mu g\,m^{-2}\,h^{-1}$). Pairwise comparisons showed that 15 out of 28 land cover classes differed significantly in N_2O fluxes (Table A8).

Mean CH_4 and N_2O fluxes were significantly inversely correlated (Spearman $R = -0.70$, $p < 0.001$), with positive CH_4 fluxes generally associated with strong negative N_2O fluxes (sinks) (Fig. A2).

3.4 N_2O Temporal trends

Unlike CH_4, most land cover classes showed no significant trends between months for N_2O (Table A6). However, there was a significant difference in N_2O fluxes between years, where year 2025 had smaller positive N_2O fluxes and larger negative fluxes (Fig. 7c for model results and Fig. A1 for observed yearly fluxes). Furthermore, within plot variability was often very high and some plots would measure both negative and positive fluxes in monthly repeated measurements.

3.5 Environmental variables

Soil temperature during the conducted chamber measurements varied between -0.02 and 22.18 °C (mean 10.23 °C). September was the coldest month (soil temperature) with the exceptions of evergreen forests, deciduous forests and permafrost bogs, where the coldest month was June. Mean soil temperatures peaked in July or August depending on the land cover (Fig. 4b), with the highest single measurements generally occurring during August. Soil moisture varied between 4.7 % and 60.7 % across all plots. Mean soil moisture was highest in fens (56.5 %), followed by tundra wetlands (55.4 %), bogs (40.8 %), permafrost bogs (34.2 %), evergreen forests (32.5 %), deciduous forests (31.9 %), shrub tundra (26.9 %) and bare tundra (25.2 %). Generally, soil moisture was highest in June due to snow melt waters and then steadily dropped towards August. Depending on the land cover, August or September was generally the driest month (Fig. 4a). Green vegetation cover ranged from 0.01 % to 85.6 %, with mean values highest in deciduous forest (37.5 %) and lowest in bare tundra (3.2 %); other land covers were between 12 %–25 %. Green vegetation cover peaked in July or August depending on the land cover while the lowest values were recorded in June and September (Fig. 4c). Soil pH was acidic across all plots except for a few locations with dolomite runoff. Mean pH was highest in tundra wetlands

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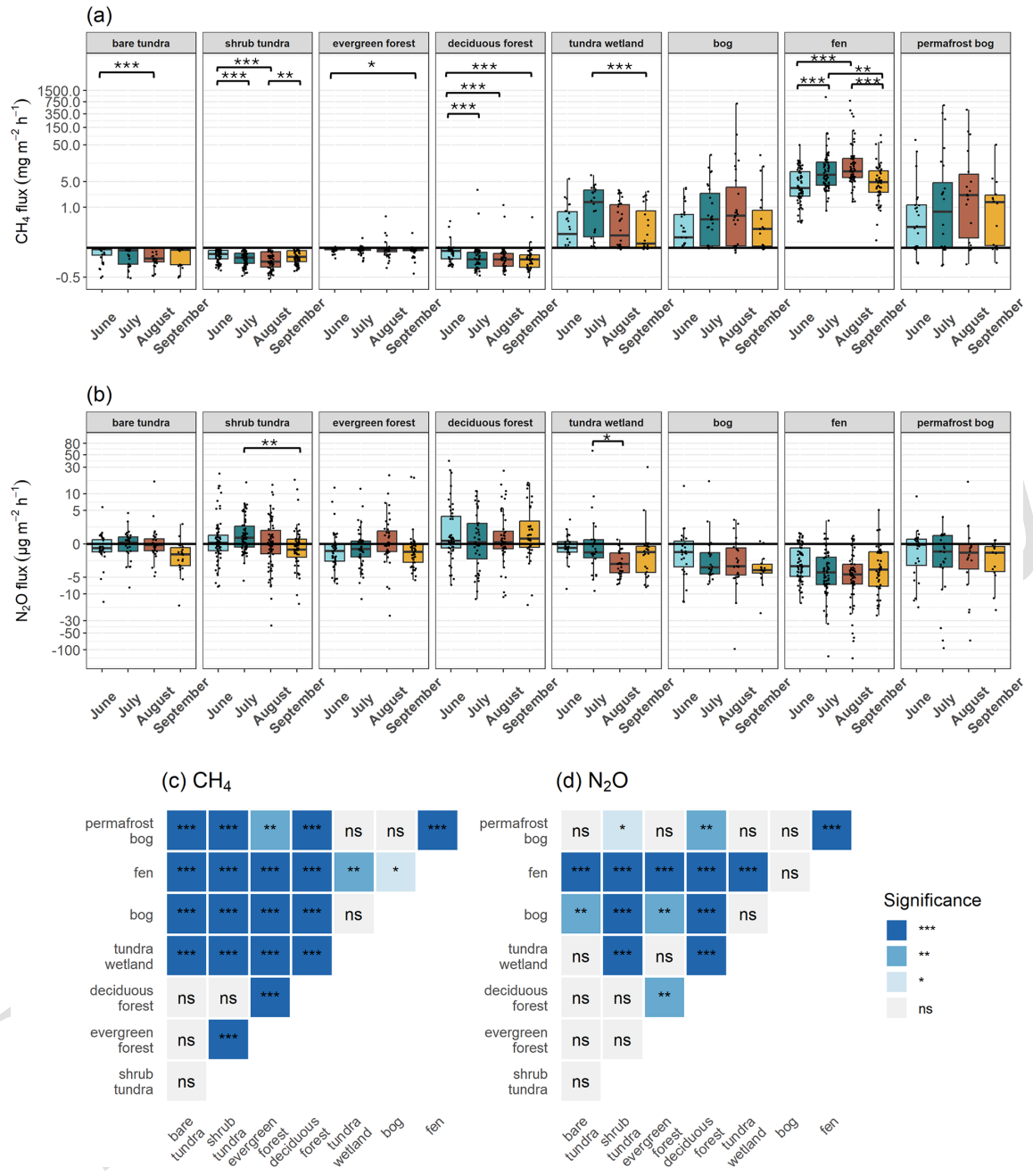


Figure 3. Growing season variability of **(a)** CH_4 and **(b)** N_2O (both $\log_{10}$-transformed) with significant monthly differences in the eight land cover classes. Positive values are emissions and negative values are sinks. Boxplots have 25th and 75th quantiles, median (horizontal line), 1.5 IQR from the lower and higher quantiles (whiskers) and individual data points (black dots). In addition, significant differences between land cover classes for **(c)** CH_4, and **(b)** N_2O (*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$).

(5.53), likely reflecting the influence of dolomite runoff at these sites, followed by bare tundra (5.21), deciduous forests (5.07), fens (4.85), shrub tundra (4.53), bogs (4.45), evergreen forests (4.38) and permafrost bogs (4.12) (Fig. 4d).

3.6 Multivariate models

3.6.1 CH_4

The mixed-effects model for CH_4 showed high explanatory power (R2 cond. = 0.96, R2 marg. 0.81). Moisture, temperature and pH all influenced CH_4 fluxes, however the direction of the response was dependent on the land cover (Figs. 5



Figure 4. Observed variation in soil and vegetation characteristics. Boxplots of **(a)** soil moisture, **(b)** soil temperature and **(c)** green vegetation cover between different land covers and months and **(d)** pH between land covers. Boxplots have 25th and 75th quantiles, median (horizontal line), mean (white dot), 1.5 IQR from the lower and higher quantiles (whiskers).

and 6). Highest effect variables based on partial eta-squared (η^2p) were soil temperature (interaction) and pH (interaction) (Fig. 5, final model). For land cover main effect, the CH_4 fluxes were highest in fens, followed by bogs, permafrost bogs, tundra wetlands, evergreen forests, bare tundra, shrub tundra and deciduous forests (Table A3). All dry land cover classes modelled as net sinks (negative fluxes).

3.6.2 N_2O

The mixed-effects model for N_2O showed modest explanatory power (R2 cond. = 0.38, R2 marg. = 0.30, Fig. 7). Modelled N_2O fluxes were highest in deciduous forests, followed by shrub tundra, bare tundra, evergreen forests, tundra wetlands, bogs, fens and permafrost bogs (Fig. 8a). Based on η^2p, land cover and its interaction with pH were the strongest predictors while both soil moisture and temperature had noticeably less effect (Fig. 7). Unlike CH_4, years 2024 and 2025 also showed a noticeable difference for N_2O with 2025 generally having smaller positive fluxes and larger sinks (Fig. 8, Table A7). In addition, vegetation cover had a small but significant effect on N_2O fluxes depending on the land cover type (Figs. 7 and 8). The point identification random effect also had a weaker effect compared to the CH_4 model.

Baseline CH_4 model → Interaction CH_4 model → Final CH_4 model

AIC	R2 cond.	R2 marg.
529.53	0.94	0.77

Parameter	Eta2 part.
moisture	0.03
temperature	0.02
pH	0.00
land cover	0.72
vegetation cover	0.00
year	0.00

AIC	R2 cond.	R2 marg.
358.90	0.96	0.81

Parameter	Eta2 part.
moisture	0.00
temperature	0.03
pH	0.03
vegetation cover	0.00
land cover	0.16
year	0.00
moisture:land cover	0.07
temperature:land cover	0.12
pH:land cover	0.15
vegetation cover:land cover	0.01

AIC	R2 cond.	R2 marg.
349.83	0.96	0.81

Parameter	Eta2 part.
moisture	0.00
temperature	0.06
pH	0.03
land cover	0.16
moisture:land cover	0.07
temperature:land cover	0.17
pH:land cover	0.16

Figure 5. Model performance indicators and effect sizes for the CH_4 linear mixed-effects models. The baseline model, which included only main effects, showed lower explanatory power compared to the interaction model that incorporated interactions with land cover class. The final model was obtained after removing the least influential predictors based on AIC to achieve a more parsimonious model structure. Partial eta-squared (Eta2 part.) describes the effect size of each model parameter.

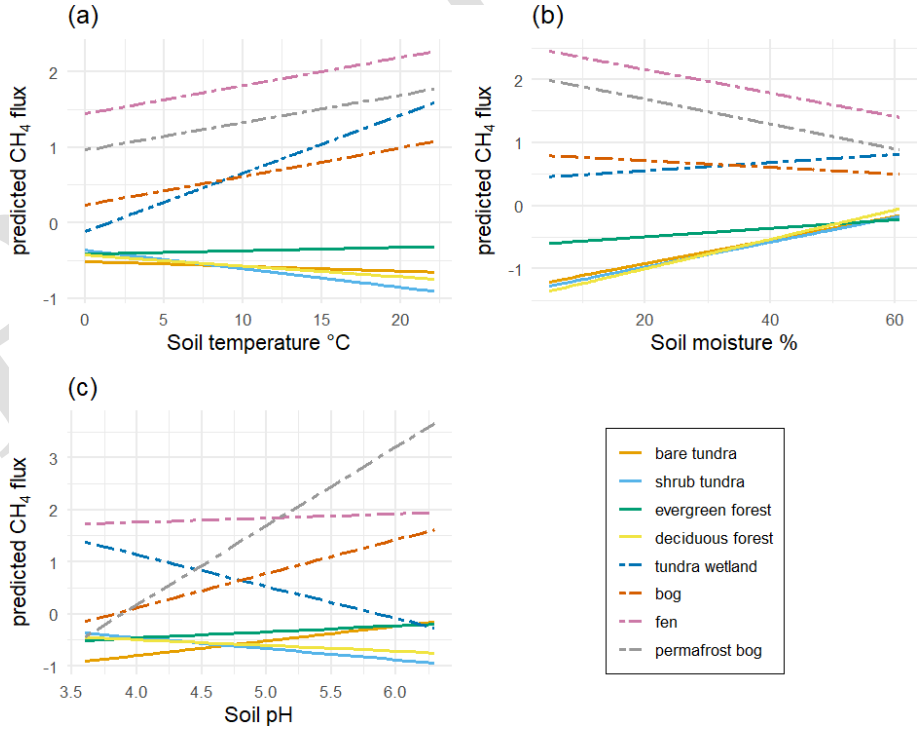
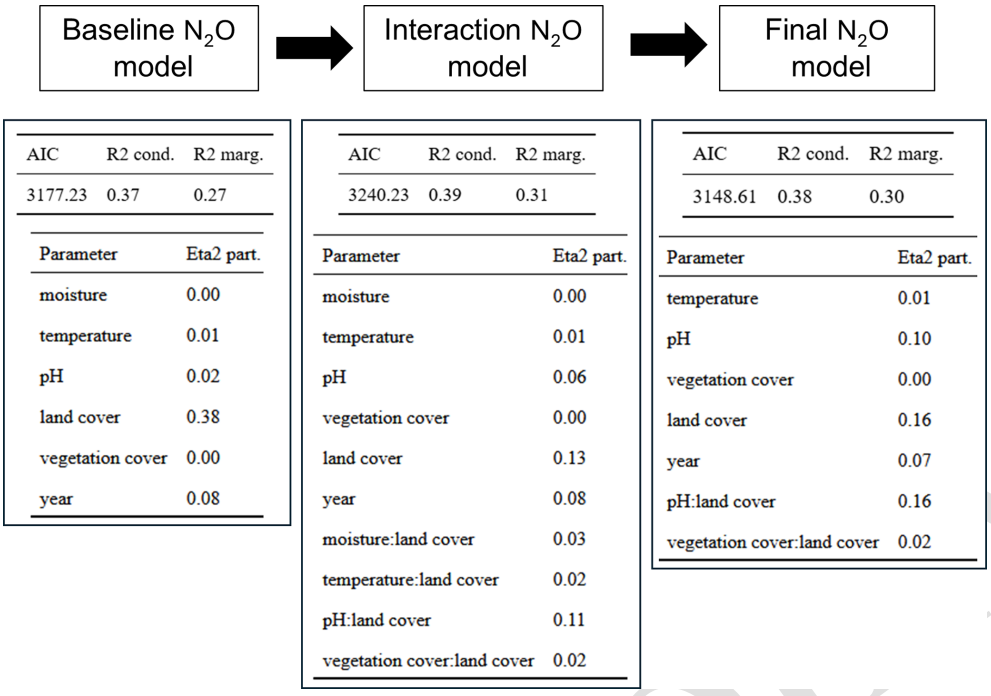


Figure 6. Land cover specific modelled (linear mixed-effect model) responses of CH_4 to **(a)** soil temperature, **(b)** soil moisture, **(c)** pH. All CH_4 model variables included interaction terms, so no simple main effects are shown. Dashed lines represent wetlands. The figure includes only predicted (normalized) values for GHG fluxes.



AIC	R2 cond.	R2 marg.
3177.23	0.37	0.27

Parameter	Eta2 part.
moisture	0.00
temperature	0.01
pH	0.02
land cover	0.38
vegetation cover	0.00
year	0.08

AIC	R2 cond.	R2 marg.
3240.23	0.39	0.31

Parameter	Eta2 part.
moisture	0.00
temperature	0.01
pH	0.06
vegetation cover	0.00
land cover	0.13
year	0.08
moisture:land cover	0.03
temperature:land cover	0.02
pH:land cover	0.11
vegetation cover:land cover	0.02

AIC	R2 cond.	R2 marg.
3148.61	0.38	0.30

Parameter	Eta2 part.
temperature	0.01
pH	0.10
vegetation cover	0.00
land cover	0.16
year	0.07
pH:land cover	0.16
vegetation cover:land cover	0.02

Figure 7. Model performance indicators and effect sizes for the N_2O linear mixed-effects models. The baseline model, which included only main effects, showed lower explanatory power compared to the interaction model that incorporated interactions with land cover class. The final model was obtained after removing the least influential predictors based on AIC to achieve a more parsimonious model structure. Partial eta-squared (Eta2 part.) describes the effect size of each model parameter.

4 Discussion

Our findings on the seasonal and spatial patterns of CH_4 and N_2O fluxes provide several insights into how environmental drivers shape gas exchange across high-latitude subarctic and boreal landscapes. Both GHGs showed strong variation among land-cover classes (research question i), with 21 of 28 pairwise comparisons being significant for CH_4 (Table A7) and 15 of 28 for N_2O fluxes (Table A8). The seasonal development of CH_4 fluxes (research question ii) was strongly dependent on land-cover class, with significant differences between months observed mainly in fen, deciduous forest, and shrub tundra sites (Table A5). In contrast, N_2O showed no clear seasonal trends (Table A6). Regarding environmental controls (research question iii), the strongest predictors of CH_4 fluxes were soil temperature and pH, with a weaker effect of soil moisture. For N_2O, pH was the strongest predictor, with weaker effects of year and green vegetation cover. For both GHGs, the direction and strength of these relationships often varied among land-cover classes. This suggests that the influence of environmental drivers is strongly context dependent and that the same change in environmental conditions can lead to contrasting flux responses across ecosystems. Consequently, broad-scale models based on uniform driver-response relationships may fail to capture important spatial variation in GHG exchange across subarctic and boreal landscapes.

4.1 GHG flux magnitudes

Dry land cover classes acted as net CH_4 sinks (-0.16 to $-0.05\,mg\,m^{-2}\,h^{-1}$, depending on land cover class) and as both N_2O sources and sinks (-0.69 to $1.69\,\mu g\,m^{-2}\,h^{-1}$). Wetlands were net CH_4 sources (0.93 to $27.81\,mg\,m^{-2}\,h^{-1}$) and net N_2O sinks (-0.89 to $-6.53\,\mu g\,m^{-2}\,h^{-1}$), with fens and permafrost bogs having the highest CH_4 fluxes and largest N_2O sinks (see full descriptions of GHG magnitudes in Tables A1 and A2).

Overall, the wetland CH_4 emissions were quite high but within maximum ranges reported in published literature (Turetsky et al., 2014; Tupek et al., 2015; Bao et al., 2021; Kuhn et al., 2021; Virkkala et al., 2025). Similar to our measurements, the ABCFlux CH_4 database (Virkkala et al., 2025) also had peak fluxes in August. Our forest and tundra CH_4 sinks were larger (more negative) than many previously reported values (Emmerton et al., 2014; Voigt et al., 2023; Virkkala et al., 2025), where some studies have even reported positive fluxes from upland tundra soils (Kuhn et al., 2021). Also, our mean net CH_4 sinks peaked in August, whereas in the ABCFlux database CH_4 sinks peaked in July (Virkkala et al., 2025). Our N_2O sources were generally at the lower end of the reported ranges, while the N_2O sinks were at the higher end (more negative) (Repo et al., 2009; Lohila et al., 2010; Maruschack et al., 2011; Tupek et al., 2015; Voigt et al., 2020; Virkkala et al., 2024; Triches et al., 2025).

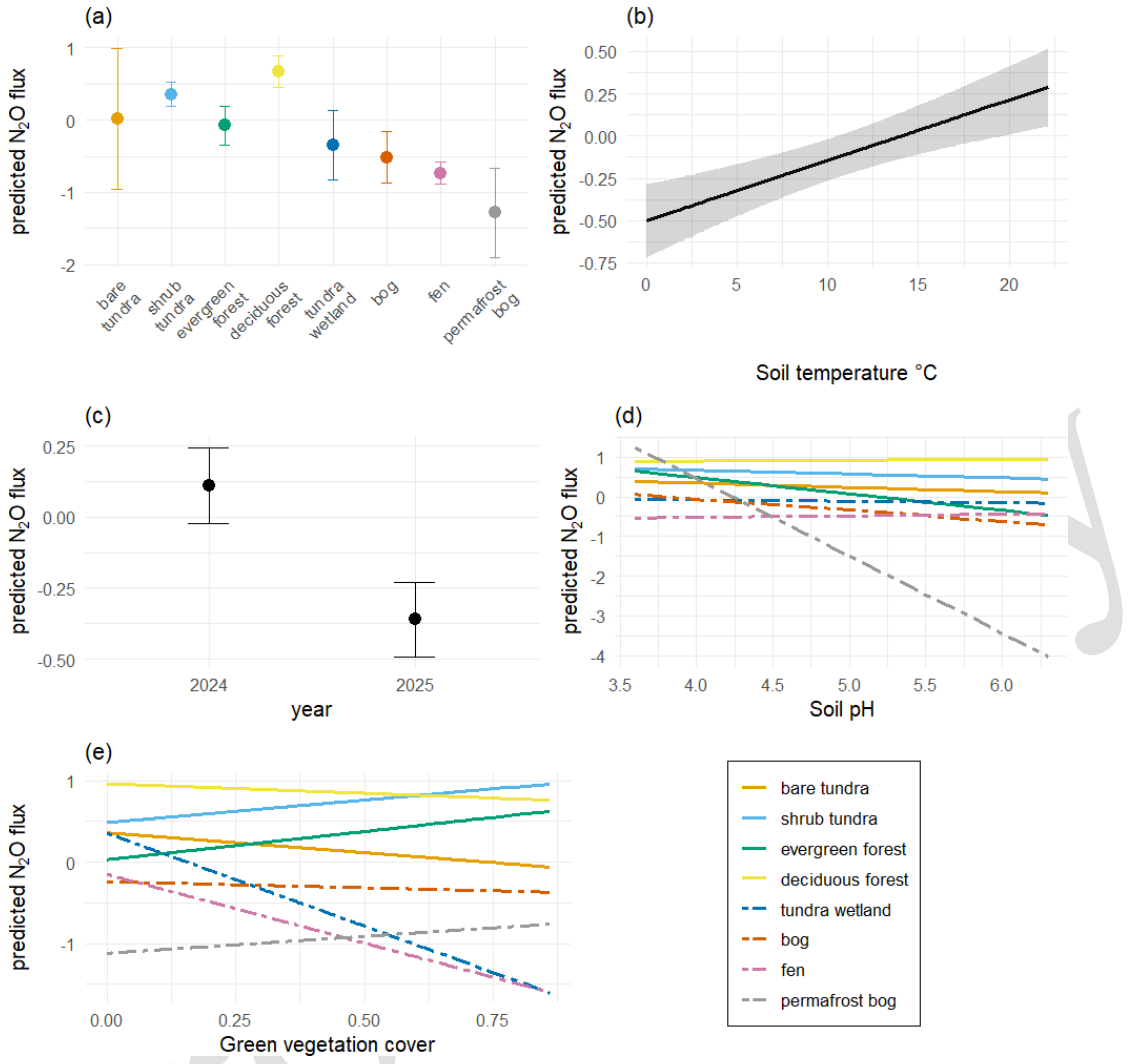


Figure 8. Modelled (linear mixed-effects model) responses of N_2O to **(a)** land cover, **(b)** soil temperature, **(c)** year, **(d)** pH-land cover interaction and **(e)** green vegetation cover-land cover interaction. Note differences in the scale of y axis between subfigures (use η^2p for comparison of effect size between parameters). Dashed lines represent wetlands. The figure includes only predicted (normalized) values for GHG fluxes.

4.2 Environmental drivers of CH_4

We observed that CH_4 emissions increased with higher soil moisture across most of the study sites. This pattern is consistent with the underlying microbial processes: methanogenesis occurs primarily in anaerobic soils (Le Mer and Roger, 2001; Lai, 2009), whereas methanotrophy dominates under aerobic conditions (Whalen, 2005; Lai, 2009). Soil oxygen content follows soil water content (Mainiero and Kazda, 2005) and as such soil moisture should have a direct causal effect on CH_4 emissions. Our results show that in dry land cover classes, soil moisture was a good indicator for CH_4. The few positive CH_4 measurements in dry land cover classes (Fig. 3a) were always from ecotones close to wetlands or streams or because of spring snow melt waters saturating the soils for extended periods of time. Conversely, CH_4 sinks in dry soils also grew stronger (more negative) with decreasing soil moisture. Notably, all mean and median fluxes for CH_4 in dry land cover classes were negative. Christiansen et al. (2015), Voigt et al. (2023) and Virkkala et al. (2024) reported similar findings, with soil moisture being one of the major drivers for CH_4 sinks at high-latitude dry upland soils and Lohila et al. (2016) found that a prolonged wet autumn period changed a net CH_4 sink forest into a net source for several months. Although the CH_4 sinks are relatively small when compared to wetland emissions, dry tundra and boreal forest soils cover extensive areas of the northern hemisphere which leads to a globally significant CH_4 sink (Jiang et al., 2025), and as previously mentioned (see Sect. 4.1 TS23), the sink magnitudes reported here are larger than in many pre-

vious studies. The soil CH_4 sink is still poorly constrained in global CH_4 budgets (Saunois et al., 2025) and should be considered in further studies.

In wetland environments, soil moisture has a more complex relationship to CH_4. Often the wetland types with highest soil moisture (i.e. fens and marshes) are also the highest CH_4 emitters (Kuhn et al., 2021), however within those wetland types, increased soil moisture does not necessarily increase CH_4 flux (Knox et al., 2019). Our results show that, within the different fen sites, the wettest soils were not the largest emitters (Fig. 5b). Soil moisture also did not increase CH_4 emissions in bogs or permafrost bogs. In permafrost bogs, intact permafrost bogs exhibited higher soil moisture but acted as substantially smaller CH_4 sources compared to fully degraded ones. This suggests that additional factors, such as soil pH, vegetation, or active layer thickness, may play a more important role in regulating wetland CH_4 fluxes. Similar results were reported by Turetsky et al. (2014) and Rinne et al. (2018), who found that CH_4 fluxes were smaller in fully flooded wetland soils and that higher water-table levels alone did not predict higher fluxes in boreal fens. Our findings suggest that there is an upper threshold of soil moisture for methanogenesis in wetlands, beyond which further increases in soil moisture no longer increase CH_4 fluxes. Reduced CH_4 emissions at sites with extreme soil moisture are likely the result of differences in substrate availability (i.e. less labile organic material) and temperature regime (Strack et al., 2004).

Soil temperature and land cover interaction was the strongest predictor for CH_4 fluxes (Fig. 5) and soil temperature has been identified as a key driver of CH_4 fluxes in many previous studies (Turetsky et al., 2014; Knox et al., 2019; Kuhn et al., 2021; Kübert et al., 2026). Soil temperature directly affects both methanogenesis and methanotrophy as they are microbial processes and therefore causally linked to soil temperature through the temperature-dependence of microbial activity (Moore and Dalva, 1993; Biasi et al., 2005; Deslippe et al., 2012). This also indicates that rising temperatures can intensify CH_4 dynamics in opposite directions where the net effect on emissions is dependent on the ecosystem's temperature sensitivity (Zhang et al., 2021). This effect can also be seen in our modelled CH_4 flux responses (Fig. 6a), where wetland CH_4 fluxes have a positive response to higher temperatures and dry land cover classes have a negative response, with higher temperatures leading to larger sinks (i.e. net-sink environments become stronger sinks and net-source environments emit more CH_4). Additionally, warming can affect the amount of organic material available for decomposition through increased GPP as higher mean temperatures generally increase biomass which could further enhance CH_4 fluxes (Knox et al., 2021). Thus, rising temperatures under climate change can strengthen both CH_4 sinks and CH_4 sources depending on land cover type, meaning that long-term changes in net CH_4 emissions will depend largely on shifts in land cover and ecosystem distributions and their effects on soil conditions.

Soil pH had a strong but mixed influence on CH_4 fluxes depending on land cover type (Figs. 5 and 6c). This response appeared to be related to the overall pH of these environments. Low-pH systems, such as permafrost bogs, bogs, and evergreen forests (Fig. 4d), showed a positive response to increasing pH, meaning higher CH_4 fluxes. In contrast, higher-pH environments, such as tundra wetlands and deciduous forests, showed the opposite pattern, meaning smaller fluxes with increased pH. It is worth noting the overall pH range was between 3.6–6.3 and as such, most vegetation and soil biota, including microbes, are adapted to acidic conditions (Kip et al., 2011; Yao et al., 2023). At the lower end of that pH range, many plant species and microbes can exist at the far end of their acidity tolerance (Yao et al., 2023) and even small increases in pH could increase their overall activity. This may explain the strong response seen in bogs and permafrost bogs as pH increases. At the opposite end, some of the tundra wetland and deciduous forest sites are affected by dolomite runoff from the surrounding watershed which significantly increases the overall soil pH. At these sites a higher pH decreased the CH_4 flux (Fig. 6c), which could indicate that methanogenesis is inhibited by higher pH while methanotrophs benefit from it. Conversely, it could also be a result of methanogenetic Archaea being more tolerant of acidic conditions than methanotrophs (Weslien et al., 2009). Other factors could also be of importance, specifically nutrient availability and vegetation are partly controlled by soil pH and directly affect CH_4 dynamics. Nonetheless, pH can exert a strong effect on CH_4 fluxes and should be considered in further studies.

4.3 Environmental drivers of N_2O

Similarly to CH_4 production and consumption, N_2O dynamics are also governed by microbial processes and are therefore influenced by factors such as soil temperature, moisture, pH, and vegetation. N_2O is mostly produced in the denitrification process where nitrates are reduced back into N_2 (Butterbach-Bahl et al., 2013). Other possible, but less important pathways include nitrification and nitrate/nitrite reduction to ammonium (Butterbach-Bahl et al., 2013). Soil oxygen content has a causal link to N_2O flux as nitrification generally only occurs in aerobic conditions and denitrification and reduction to ammonium mostly occur in anaerobic conditions (Shan et al., 2021). However, N_2O showed a much weaker response to the environmental variables compared to CH_4, and the explanatory power of the models was correspondingly smaller (Fig. 7). In addition, N_2O showed a significant difference between measurement years which remained unexplained by the other factors, suggesting that some key explanatory variables were missing from the model. Year 2024, which had the higher mean N_2O fluxes, was considerably warmer during the measurement

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period than 2025. Difference in growing season mean temperature for Pallas was 0.90 and 1.35 °C for Kilpisjärvi (Table 1). It is possible that these temperature differences contributed to the observed patterns. However, because soil temperature showed only a weak relationship with N_2O fluxes, any such effect was likely indirect, for example through the influence of air temperature on vegetation. Previous studies have identified sources of N_2O from many high-latitude environments such as unvegetated permafrost peat plateaus and palsas (Maruschak et al., 2011) and both vegetated and unvegetated permafrost tundra soils (Voigt et al., 2020) but also sinks from nutrient-poor permafrost peatlands (Triches et al., 2025) and grasslands and forests (Chapuis-Lardy et al., 2007). In our measurements, we found strong sinks from wetlands, while dry land cover classes had medians close to zero fluxes, however there was a noticeable amount of both negative and positive extreme values (Fig. 3b) and many land cover classes and single measurement plots exhibited both negative and positive fluxes. This type of flux behaviour is not uncommon (Lohila et al., 2010; Tupek et al., 2015) and highlights the spatiotemporally sporadic behaviour of N_2O, where many plots have zero or small positive N_2O emissions, yet a small number of plots can act as significant hotspots, both for sinks and sources (Maruschak et al., 2011; Triches et al., 2025).

Soil temperature and moisture generally had modest effects on N_2O fluxes and much of the variability was explained by pH and land cover (particularly wetlands). Our highest measured mean N_2O fluxes came from deciduous (mountain birch) forests, which also had comparatively higher soil pH (Fig. 4d). Cryoturbation and the absence of vegetation have been occasionally reported as having an increasing effect on N_2O fluxes, especially in permafrost soils (Repo et al., 2009; Maruschak et al., 2011; Voigt et al., 2020). However, we found no strong evidence for such an effect in our measurements from peat soils or tundra cryoturbation plots. Both bare tundra and permafrost bogs acted as small net sinks for N_2O (Fig. 3b), and vegetation cover only had a small effect in the model results.

It is possible that low soil nitrogen availability is the primary reason for the small N_2O fluxes, while also being a key missing variable in the model. Nitrate (NO_3^-) and nitrite (NO_2^-) are the preferred electron acceptors in denitrification, thus in the presence of excess soil nitrogen content, N_2O would not be reduced and would likely result in larger positive fluxes. Furthermore, there is evidence of NO_3^- reliant pathways of methanotrophy that can produce N_2O in low oxygen conditions such as wetlands (Kits et al., 2015).

4.4 N_2O sinks

The largest N_2O sinks were found in wetlands, specifically in fen plots (Fig. 3b). This may appear counterintuitive as denitrification, which is the main pathway for N_2O production (Butterbach-Bahl et al., 2013), requires anoxic conditions which wetlands typically have. The reason is likely in the last step of the denitrification chain where N_2O is reduced. In anoxic conditions with high C/N ratios or low NO_3^- concentrations, the denitrification process is fully completed, N_2O is produced but then reduced into N_2 gas (Kolb and Horn, 2012; Brummell et al., 2012; Pessi et al., 2022). In these low-nitrate conditions, any excess N_2O would be rapidly reduced, including atmospheric N_2O, resulting in a net sink.

Other less studied pathways for N_2O sinks and the inverse N_2O and CH_4 correlation involve anaerobic methane oxidation driven by denitrification (Valenzuela, 2020TS24; Tan et al., 2025). It follows that in soils rich in CH_4, such as wetlands, substantial methanotrophic activity would occur throughout the soil profile, provided that environmental conditions permit it (Whalen et al., 1990). However, chamber measurements only capture the net CH_4 flux, so if net oxidation is smaller than net CH_4 production, the oxidation would not be detected. Anerobic methane oxidation (AOM) is a well-known process in marine environments (Hinrichs and Boetius, 2002TS25; Reeburgh, 2007), but less so in wetlands, yet there is growing evidence for its importance (Smemo and Yavitt, 2011; Cai et al., 2019TS26; Cheng et al., 2021). In AOM, CH_4 can be used as electron donor for reducing nitrite and N_2O, which would act as an additional sink of N_2O (Valenzuela et al., 2020; Cheng et al., 2021; Tan et al., 2025). However, the magnitude of these pathways in natural systems is unknown and somewhat hypothetical at this point and needs more research to be confirmed. Furthermore, the negative relationship between CH_4 and N_2O fluxes was restricted to wetland ecosystems (Fig. A2b). This suggests that the coupling between CH_4 and N_2O dynamics depends on environmental conditions and may be strongest under the wet and anaerobic conditions characteristic of wetlands. Regardless of the underlying mechanisms, based on our measurements, wetlands can act as notable N_2O sinks in the right conditions.

5 Uncertainties

The monthly repeated measurements and two-year design resulted in a robust overall dataset, even though some land cover classes, such as permafrost bogs, were represented by fewer plots. Generic environmental variables that can be measured from anywhere make the models easily applicable everywhere, but it also limits the number of usable explanatory variables. Modelling across diverse environments (i.e. not solely focusing on specific ecosystems such as wetlands) provides a valuable opportunity to capture a wide range of ecosystem behaviors, though it naturally results in more complex data structures. In particular, wetland GHG fluxes often show substantial variability, including occasional high fluxes far above median values (Kuhn et al., 2021; Saunois et al., 2025). These contrasts become pronounced when wetlands and non-wetland systems are analyzed together due to their inherently different flux magnitudes. Furthermore,

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Table 1. Long-term (1991–2020) and study-period (2024–2025) mean temperatures and precipitation during the growing season (June–September). Data were obtained from the nearest weather stations (Kilpisjärvi Kyläkeskus and Muonio Kirkonkylä).

Location	years 1991–2020 °C (Jun–Sep)	year 2024 °C (Jun–Sep)	year 2025 °C (Jun–Sep)	1991–2020 precipitation (mm)	2024 precipitation (mm)	2025 precipitation (mm)
Kilpisjärvi	8.78 °C	11.80 °C	10.45 °C	52.50 mm	56.33 mm	56.15 mm
Pallas	10.93 °C	12.93 °C	12.03 °C	65.00 mm	77.35 mm	67.63 mm

capturing seasonal changes requires repeated measurements which limit the available statistical methods due to pseudoreplication. These complexities were effectively addressed in our statistical analyses through the use of robust and flexible approaches, including mixed multivariate models, appropriate data transformations, and non-parametric tests.

The multivariate and correlative nature of our models makes it challenging to directly disentangle causal pathways. Also, some variables, such as land cover, function as a composite descriptor of multiple co-varying environmental conditions rather than a direct mechanistic driver. Because land cover classes ultimately reflect underlying temperature, moisture, and other soil properties, these environmental factors are the more direct drivers of GHG fluxes than land cover per se.

Opaque ("dark") chambers are used to reduce uncertainties caused by changes in chamber temperature or possible interactions between photosynthesis and GHG fluxes and are typically the default method for measuring CH_4 and N_2O fluxes (Livingston and Hutchinson, 2009; Pavelka et al., 2018; Clough et al., 2020). However, there is recent evidence of CH_4 and N_2O fluxes being affected by varying light conditions (Voigt et al., 2023; Triches et al., 2025). Thus, it is possible that GHG fluxes measured with transparent and opaque chambers are not directly comparable and the additional use of transparent chambers should be considered in future research.

6 Conclusions

Across all ecosystems, we found consistent contrasts in greenhouse gas dynamics, with tundra and forest sites acting as modest CH_4 sinks and wetlands functioning as strong CH_4 sources, whereas N_2O flux patterns were more subtle. Most notably wetlands acted as consistent N_2O sinks.

Land cover, pH, soil temperature and soil moisture were strong predictors of CH_4 fluxes and explained most of the observed variance, whereas green vegetation cover had only a marginal effect. CH_4 fluxes exhibited clear seasonal patterns, with the smallest sources and sinks occurring at the beginning and end of the growing season while peak fluxes were observed in July–August. These findings highlight the importance of longitudinal studies for accurately estimating annual CH_4 budgets rather than relying solely on peak growing-season measurements but also demonstrates that CH_4 fluxes can be modelled with relatively few key environmental variables.

In comparison to CH_4, N_2O exhibited relatively moderate responses to the measured environmental variables and considerably weaker seasonal patterns, emphasizing the inherent challenges in explaining N_2O dynamics across heterogeneous ecosystems. In a multivariate context, land cover and soil pH emerged as the most influential factors shaping N_2O flux variability, while the effects of soil temperature and moisture were negligible. The substantial proportion of unexplained variability in N_2O fluxes indicates that key controlling factors were not fully captured, pointing to the importance of additional drivers, particularly those related to soil biogeochemistry and microbial nitrogen transformation pathways. Furthermore, the high variability of N_2O emissions suggests a notable influence of spatiotemporal dynamics, underscoring the need for high resolution measurements to adequately characterize N_2O behavior. These findings highlight the complexity of N_2O dynamics and identify important avenues for future research to improve the understanding and modeling of N_2O fluxes.

All in all, our results demonstrate that high-latitude GHG fluxes arise from interactions between broad land cover patterns and fine-scale environmental variability. From an applied perspective, our findings suggest that accurate land cover mapping and in situ measurements of soil conditions are essential for robust modelling and upscaling of GHG fluxes. Because land cover classes exhibit such different GHG dynamics due to the underlying soil and vegetation characteristics, improved classification greatly enhances predictive performance across heterogenous landscapes. This is especially important in high-latitude regions, where rapid warming is expected to alter both ecosystem distributions and soil conditions with significant implications for current and future GHG budgets.

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Appendix A

Table A1. CH_4 and N_2O fluxes in different land cover classes.

	CH_4 ($mg\,m^{-2}\,h^{-1}$)					N_2O ($\mu g\,m^{-2}\,h^{-1}$)				
land cover	Mean	SD	Median	Min	Max	Mean	SD	Median	Min	Max
bare tundra ($n = 106$)	−0.126	0.143	−0.075	−0.525	0.008	−0.691	3.354	−0.559	−16.101	16.721
shrub tundra ($n = 266$)	−0.160	0.097	−0.141	−0.553	0.001	0.435	4.774	0.191	−36.857	23.038
evergreen forest ($n = 186$)	−0.045	0.084	−0.042	−0.398	0.567	−0.523	4.491	−0.885	−24.271	21.549
deciduous forest ($n = 180$)	−0.123	0.298	−0.148	−0.499	3.020	1.697	6.484	0.574	−15.609	39.175
tundra wetland ($n = 106$)	0.925	1.354	0.245	−0.039	7.587	−0.893	7.241	−1.557	−8.772	59.463
bog ($n = 100$)	10.073	67.715	0.321	−0.108	668.329	−3.512	10.488	−2.942	−96.405	16.802
fen ($n = 248$)	23.368	90.651	6.561	0.128	1007.896	−6.527	14.321	−3.867	−141.854	5.134
permafrost bog ($n = 80$)	27.808	96.405	0.960	−0.242	602.251	−5.404	15.920	−1.034	−92.814	16.559

Table A2. Monthly CH_4 and N_2O fluxes in different land cover classes.

			CH_4 ($mg\,m^{-2}\,h^{-1}$)					N_2O ($\mu g\,m^{-2}\,h^{-1}$)				
	land cover	n	Mean	SD	Median	Min	Max	Mean	SD	Median	Min	Max
June	bare tundra	28	−0.102	0.143	−0.040	−0.519	0.005	−0.900	3.251	−0.777	−13.855	5.774
	shrub tundra	70	−0.127	0.072	−0.119	−0.326	0.001	0.934	4.804	0.174	−7.323	23.038
	evergreen forest	48	−0.035	0.043	−0.022	−0.154	0.028	−0.966	3.249	−1.289	−6.440	12.971
	deciduous forest	48	−0.079	0.129	−0.087	−0.261	0.419	2.863	8.201	0.547	−6.835	39.175
	tundra wetland	26	0.643	1.213	0.197	0.001	5.940	−0.628	2.130	−0.725	−8.103	3.456
	bog	27	0.594	0.940	0.153	0.001	3.373	−1.734	5.109	−1.468	−13.851	13.886
	fen	69	6.433	7.366	3.321	0.385	49.585	−3.202	3.198	−3.198	−14.518	1.940
	permafrost bog	24	5.165	15.139	0.291	−0.211	68.058	−2.045	5.583	−0.170	−17.242	9.023
July	bare tundra	28	−0.133	0.150	−0.071	−0.525	0.005	−0.125	2.395	0.246	−7.636	4.548
	shrub tundra	70	−0.169	0.092	−0.149	−0.461	−0.001	1.441	3.062	1.068	−5.782	16.018
	evergreen forest	48	−0.048	0.058	−0.043	−0.188	0.147	−0.766	3.443	−0.906	−12.323	12.315
	deciduous forest	43	−0.104	0.510	−0.166	−0.453	3.020	0.603	5.080	0.274	−12.319	11.152
	tundra wetland	26	1.784	1.877	1.392	−0.012	7.587	1.300	12.380	−1.552	−8.772	59.463
	bog	26	2.877	5.801	0.470	−0.018	26.812	−2.074	4.438	−3.353	−6.650	16.802
	fen	69	28.840	121.145	7.639	0.812	1007.896	−7.364	16.234	−4.116	−130.475	3.297
	permafrost bog	24	54.310	145.913	1.009	−0.242	602.251	−9.334	23.795	−1.404	−92.814	3.859
August	bare tundra	28	−0.151	0.132	−0.156	−0.457	0.003	0.254	3.742	−0.207	−4.784	16.721
	shrub tundra	70	−0.197	0.119	−0.189	−0.553	−0.003	−0.274	6.011	−0.062	−36.857	14.503
	evergreen forest	39	−0.037	0.133	−0.058	−0.280	0.567	0.401	6.216	−0.011	−24.271	21.549
	deciduous forest	48	−0.145	0.240	−0.164	−0.425	1.160	1.244	6.053	0.260	−11.271	25.924
	tundra wetland	30	0.700	0.921	0.180	−0.039	2.948	−3.012	1.845	−2.885	−6.702	0.794
	bog	27	31.463	128.687	0.594	−0.108	668.329	−6.077	18.389	−3.189	−96.405	3.418
	fen	63	46.033	123.948	9.556	1.432	794.017	−10.279	21.438	−4.460	−141.854	1.425
	permafrost bog	18	40.396	107.838	2.140	−0.223	450.080	−6.032	17.471	−1.609	−68.213	16.559
September	bare tundra	22	−0.118	0.154	−0.053	−0.507	0.008	−2.350	3.566	−1.868	−16.101	2.884
	shrub tundra	54	−0.142	0.081	−0.140	−0.341	−0.003	−0.597	4.515	−1.028	−15.010	17.925
	evergreen forest	42	−0.060	0.084	−0.048	−0.398	0.200	−0.598	4.897	−1.417	−5.606	19.983
	deciduous forest	40	−0.170	0.160	−0.165	−0.499	0.540	2.019	5.964	0.929	−15.609	15.651
	tundra wetland	24	0.579	0.870	0.102	−0.035	2.700	−0.907	7.221	−1.499	−7.505	30.325
	bog	18	3.003	6.932	0.259	−0.038	26.535	−4.406	4.702	−3.730	−21.947	0.475
	fen	46	9.814	16.613	4.863	0.128	92.036	−5.119	5.880	−3.675	−24.178	5.134
	permafrost bog	14	5.008	13.228	1.375	−0.206	50.499	−3.617	5.547	−1.582	−19.359	0.707

Proof only

Table A3. Linear mixed-effects model summary for CH_4 flux. Model estimates are standardized.

	Est.	S.E.	*t*	*p*
Intercept	−0.582	0.209	−2.786	0.005
moisture	0.284	0.092	3.102	0.002
temperature	−0.020	0.016	−1.232	0.218
pH	0.168	0.188	0.895	0.371
shrub tundra	−0.033	0.229	−0.143	0.886
evergreen forest	0.211	0.245	0.861	0.389
deciduous forest	0.004	0.232	0.018	0.986
tundra wetland	1.258	0.318	3.955	< 0.001
bog	1.198	0.273	4.384	< 0.001
fen	2.405	0.242	9.950	< 0.001
permafrost bog	1.906	0.385	4.956	< 0.001
moisture × shrub tundra	0.014	0.108	0.132	0.895
moisture × evergreen forest	−0.182	0.107	−1.704	0.089
moisture × deciduous forest	0.070	0.103	0.681	0.496
moisture × tundra wetland	−0.186	0.125	−1.489	0.137
moisture × bog	−0.365	0.106	−3.437	< 0.001
moisture × fen	−0.568	0.121	−4.701	< 0.001
moisture × permafrost bog	−0.585	0.160	−3.645	< 0.001
temperature × shrub tundra	−0.057	0.022	−2.629	0.009
temperature × evergreen forest	0.034	0.025	1.358	0.175
temperature × deciduous forest	−0.025	0.029	−0.864	0.388
temperature × tundra wetland	0.264	0.028	9.462	< 0.001
temperature × bog	0.140	0.027	5.154	< 0.001
temperature × fen	0.138	0.022	6.229	< 0.001
temperature × permafrost bog	0.135	0.032	4.268	< 0.001
pH × shrub tundra	−0.295	0.220	−1.343	0.179
pH × evergreen forest	−0.099	0.240	−0.415	0.679
pH × deciduous forest	−0.234	0.205	−1.141	0.254
pH × tundra wetland	−0.535	0.239	−2.234	0.026
pH × bog	0.222	0.311	0.712	0.476
pH × fen	−0.120	0.204	−0.588	0.556
pH × permafrost bog	0.741	0.325	2.284	0.023

Proof only

Table A4. Linear mixed-effects model summary for N_2O flux. Model estimates are standardized.

	Est.	S.E.	*t*	*p*
Intercept	0.265	0.499	0.531	0.595
temperature	0.113	0.027	4.176	< 0.001
pH	−0.065	0.184	−0.354	0.723
vegetation cover	−0.078	0.457	−0.170	0.865
shrub tundra	0.334	0.506	0.661	0.509
evergreen forest	−0.091	0.518	−0.176	0.860
deciduous forest	0.653	0.512	1.275	0.202
tundra wetland	−0.366	0.555	−0.660	0.510
bog	−0.532	0.530	−1.004	0.316
fen	−0.753	0.504	−1.496	0.135
permafrost bog	−1.283	0.586	−2.189	0.029
year 2025	−0.471	0.048	−9.790	< 0.001
pH × shrub tundra	0.007	0.215	0.035	0.972
pH × evergreen forest	−0.182	0.238	−0.765	0.444
pH × deciduous forest	0.078	0.200	0.388	0.698
pH × tundra wetland	0.043	0.235	0.185	0.853
pH × bog	−0.102	0.306	−0.334	0.738
pH × fen	0.086	0.202	0.427	0.669
pH × permafrost bog	−1.106	0.312	−3.544	< 0.001
vegetation cover × shrub tundra	0.166	0.464	0.358	0.720
vegetation cover × evergreen forest	0.188	0.463	0.407	0.684
vegetation cover × deciduous forest	0.041	0.460	0.088	0.930
vegetation cover × tundra wetland	−0.286	0.486	−0.589	0.556
vegetation cover × bog	0.054	0.493	0.109	0.913
vegetation cover × fen	−0.191	0.461	−0.415	0.678
vegetation cover × permafrost bog	0.144	0.497	0.290	0.772

Proof only

Table A5. ART-ANOVA summary for monthly differences of CH_4 flux in different land cover types. Significance intervals $P < 0.001$ ***, $P < 0.01$ **, $P < 0.05$ TS27 *.

land cover	month	estimate	SE	df	t.ratio	p.value	sig
bare tundra	June–September	49.90	23.58	1095.29	2.12	0.97	
bare tundra	July–June	−65.93	22.07	1095.01	−2.99	0.42	
bare tundra	July–September	−16.03	23.58	1095.29	−0.68	1.00	
bare tundra	August–June	−130.79	22.07	1095.01	−5.93	0.00	***
bare tundra	August–July	−64.86	22.07	1095.01	−2.94	0.46	
bare tundra	August–September	−80.89	23.58	1095.29	−3.43	0.15	
bog	June–September	−47.75	25.24	1095.65	−1.89	0.99	
bog	July–June	67.22	22.48	1095.01	2.99	0.42	
bog	July–September	19.47	25.24	1095.65	0.77	1.00	
bog	August–June	65.04	22.48	1095.01	2.89	0.50	
bog	August–July	−2.19	22.48	1095.01	−0.10	1.00	
bog	August–September	17.29	25.24	1095.65	0.68	1.00	
deciduous forest	June–September	174.65	17.72	1095.32	9.85	0.00	***
deciduous forest	July–June	−164.36	17.35	1095.83	−9.47	0.00	***
deciduous forest	July–September	10.29	18.08	1095.25	0.57	1.00	
deciduous forest	August–June	−156.62	17.01	1096.15	−9.21	0.00	***
deciduous forest	August–July	7.74	17.35	1095.83	0.45	1.00	
deciduous forest	August–September	18.03	17.72	1095.32	1.02	1.00	
evergreen forest	June–September	70.00	17.49	1095.28	4.00	0.02	*
evergreen forest	July–June	−41.08	16.86	1095.01	−2.44	0.85	
evergreen forest	July–September	28.91	17.49	1095.28	1.65	1.00	
evergreen forest	August–June	−65.78	17.90	1096.71	−3.67	0.07	
evergreen forest	August–July	−24.70	17.90	1096.71	−1.38	1.00	
evergreen forest	August–September	4.22	18.45	1096.59	0.23	1.00	
fen	June–September	−12.46	15.67	1095.61	−0.80	1.00	
fen	July–June	73.96	14.06	1095.01	5.26	0.00	***
fen	July–September	61.49	15.67	1095.61	3.92	0.03	*
fen	August–June	97.25	14.44	1095.57	6.74	0.00	***
fen	August–July	23.29	14.44	1095.57	1.61	1.00	
fen	August–September	84.79	15.96	1095.47	5.31	0.00	***
permafrost bog	June–September	−5.25	27.80	1095.16	−0.19	1.00	
permafrost bog	July–June	35.50	23.84	1095.01	1.49	1.00	
permafrost bog	July–September	30.25	27.80	1095.16	1.09	1.00	
permafrost bog	August–June	42.93	25.93	1095.98	1.66	1.00	
permafrost bog	August–July	7.43	25.93	1095.98	0.29	1.00	
permafrost bog	August–September	37.67	29.48	1095.25	1.28	1.00	
shrub tundra	June–September	37.73	14.84	1095.29	2.54	0.78	
shrub tundra	July–June	−78.77	13.96	1095.01	−5.64	0.00	***
shrub tundra	July–September	−41.04	14.84	1095.29	−2.77	0.61	
shrub tundra	August–June	−107.63	13.96	1095.01	−7.71	0.00	***
shrub tundra	August–July	−28.86	13.96	1095.01	−2.07	0.98	
shrub tundra	August–September	−69.89	14.84	1095.29	−4.71	0.00	**
tundra wetland	June–September	32.21	23.41	1095.15	1.38	1.00	
tundra wetland	July–June	81.58	22.90	1095.01	3.56	0.10	
tundra wetland	July–September	113.79	23.41	1095.15	4.86	0.00	***
tundra wetland	August–June	14.87	22.17	1095.24	0.67	1.00	
tundra wetland	August–July	−66.71	22.17	1095.24	−3.01	0.41	
tundra wetland	August–September	47.08	22.73	1095.56	2.07	0.97	

Proof only

Table A6. ART-ANOVA summary table for monthly differences of N_2O flux in different land cover types. Significance intervals $P < 0.001$ ***, $P < 0.01$ **, $P < 0.05$ *.

land cover	month	estimate	SE	df	t.ratio	p.value	sig
bare tundra	June–September	169.20	85.28	1096.41	1.98	0.99	
bare tundra	July–June	88.39	79.90	1091.13	1.11	1.00	
bare tundra	July–September	257.59	85.28	1096.41	3.02	0.40	
bare tundra	August–June	64.21	79.90	1091.13	0.80	1.00	
bare tundra	August–July	−24.18	79.90	1091.13	−0.30	1.00	
bare tundra	August–September	233.42	85.28	1096.41	2.74	0.63	
bog	June–September	269.47	91.17	1101.04	2.96	0.45	
bog	July–June	−144.95	82.19	1093.07	−1.76	1.00	
bog	July–September	124.53	91.91	1102.57	1.35	1.00	
bog	August–June	−123.70	81.37	1091.13	−1.52	1.00	
bog	August–July	21.24	82.19	1093.07	0.26	1.00	
bog	August–September	145.77	91.17	1101.04	1.60	1.00	
deciduous forest	June–September	−26.69	64.09	1096.88	−0.42	1.00	
deciduous forest	July–June	−83.51	62.96	1104.15	−1.33	1.00	
deciduous forest	July–September	−110.20	65.73	1094.72	−1.68	1.00	
deciduous forest	August–June	−40.82	61.32	1111.63	−0.67	1.00	
deciduous forest	August–July	42.69	62.96	1104.15	0.68	1.00	
deciduous forest	August–September	−67.50	64.09	1096.88	−1.05	1.00	
evergreen forest	June–September	20.48	63.24	1096.27	0.32	1.00	
evergreen forest	July–June	53.06	61.03	1091.13	0.87	1.00	
evergreen forest	July–September	73.54	63.24	1096.27	1.16	1.00	
evergreen forest	August–June	196.00	64.92	1123.01	3.02	0.40	
evergreen forest	August–July	142.94	64.92	1123.01	2.20	0.95	
evergreen forest	August–September	216.48	66.87	1117.63	3.24	0.25	
fen	June–September	61.12	57.00	1099.78	1.07	1.00	
fen	July–June	−124.04	50.90	1091.13	−2.44	0.85	
fen	July–September	−62.92	57.00	1099.78	−1.10	1.00	
fen	August–June	−141.57	52.17	1098.29	−2.71	0.65	
fen	August–July	−17.53	52.17	1098.29	−0.34	1.00	
fen	August–September	−80.45	58.07	1099.08	−1.39	1.00	
permafrost bog	June–September	144.03	100.60	1093.73	1.43	1.00	
permafrost bog	July–June	−61.83	86.31	1091.13	−0.72	1.00	
permafrost bog	July–September	82.19	100.60	1093.73	0.82	1.00	
permafrost bog	August–June	−61.84	93.55	1107.18	−0.66	1.00	
permafrost bog	August–July	−0.01	93.55	1107.18	−0.00	1.00	
permafrost bog	August–September	82.19	106.63	1095.23	0.77	1.00	
shrub tundra	June–September	122.27	54.22	1097.25	2.25	0.93	
shrub tundra	July–June	109.57	50.54	1091.13	2.17	0.95	
shrub tundra	July–September	231.85	54.22	1097.25	4.28	0.01	**
shrub tundra	August–June	−36.89	50.54	1091.13	−0.73	1.00	
shrub tundra	August–July	−146.46	50.54	1091.13	−2.90	0.50	
shrub tundra	August–September	85.39	54.22	1097.25	1.57	1.00	
tundra wetland	June–September	175.04	84.70	1094.23	2.07	0.98	
tundra wetland	July–June	−68.19	82.92	1091.13	−0.82	1.00	
tundra wetland	July–September	106.85	84.70	1094.23	1.26	1.00	
tundra wetland	August–June	−317.59	80.19	1095.48	−3.96	0.03	*
tundra wetland	August–July	−249.40	80.19	1095.48	−3.11	0.33	
tundra wetland	August–September	−142.55	82.10	1102.33	−1.74	1.00	

Proof only

Table A7. ART-ANOVA summary table for differences of CH_4 flux between land cover classes. Significance intervals $P < 0.001$ ***, $P < 0.01$ **, $P < 0.05$ *.

contrast	estimate	SE	df	t.ratio	p.value	sig
bare tundra – shrub tundra	138.87	59.09	134.78	2.35	0.27	
bare tundra – evergreen forest	−111.7	61.85	135.16	−1.81	0.62	
bare tundra – deciduous forest	127.93	62.35	134.89	2.05	0.45	
bare tundra – tundra wetland	−421.61	69.36	135.13	−6.08	0	***
bare tundra – bog	−451.18	70.89	135.36	−6.36	0	***
bare tundra – fen	−661.32	58.36	135.52	−11.33	0	***
bare tundra – permafrost bog	−364.35	72.73	136.07	−5.01	0	***
shrub tundra – evergreen forest	−250.57	48	135.5	−5.22	0	***
shrub tundra – deciduous forest	−10.94	48.64	135.04	−0.22	1	
shrub tundra – tundra wetland	−560.48	57.36	135.35	−9.77	0	***
shrub tundra – bog	−590.05	59.19	135.67	−9.97	0	***
shrub tundra – fen	−800.19	43.41	136.22	−18.43	0	***
shrub tundra – permafrost bog	−503.21	61.39	136.65	−8.2	0	***
evergreen forest – deciduous forest	239.63	51.96	135.56	4.61	0	***
evergreen forest – tundra wetland	−309.91	60.2	135.71	−5.15	0	***
evergreen forest – bog	−339.48	61.95	135.98	−5.48	0	***
evergreen forest – fen	−549.62	47.1	136.68	−11.67	0	***
evergreen forest – permafrost bog	−252.64	64.05	136.87	−3.94	0	**
deciduous forest – tundra wetland	−549.54	60.71	135.41	−9.05	0	***
deciduous forest – bog	−579.11	62.45	135.69	−9.27	0	***
deciduous forest – fen	−789.25	47.75	136.17	−16.53	0	***
deciduous forest – permafrost bog	−492.27	64.53	136.58	−7.63	0	***
tundra wetland – bog	−29.57	69.45	135.78	−0.43	1	
tundra wetland – fen	−239.71	56.61	136.16	−4.23	0	**
tundra wetland – permafrost bog	57.27	71.33	136.5	0.8	0.99	
bog – fen	−210.14	58.47	136.44	−3.59	0.01	*
bog – permafrost bog	86.83	72.82	136.67	1.19	0.93	
fen – permafrost bog	296.98	60.69	137.4	4.89	0	***

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Table A8. ART-ANOVA summary table for differences of N_2O flux between land cover classes. Significance intervals $P < 0.001$ ***, $P < 0.01$ **, $P < 0.05$ *.

contrast	estimate	SE	df	t.ratio	p.value	sig
bare tundra – shrub tundra	−89.94	48.65	121.36	−1.85	0.59	
bare tundra – evergreen forest	25.55	51.45	126.92	0.5	1	
bare tundra – deciduous forest	−135.84	51.54	124.17	−2.64	0.15	
bare tundra – tundra wetland	134.55	57.62	126.26	2.33	0.28	
bare tundra – bog	222.82	58.99	125.11	3.78	0.01	**
bare tundra – fen	349.7	48.76	126.89	7.17	0	***
bare tundra – permafrost bog	93.43	61.59	133.88	1.52	0.8	
shrub tundra – evergreen forest	115.49	40.34	132.42	2.86	0.09	
shrub tundra – deciduous forest	−45.9	40.46	127.68	−1.13	0.95	
shrub tundra – tundra wetland	224.49	47.97	129.77	4.68	0	***
shrub tundra – bog	312.76	49.6	127.87	6.31	0	***
shrub tundra – fen	439.64	36.85	133.49	11.93	0	***
shrub tundra – permafrost bog	183.37	52.66	140.13	3.48	0.01	*
evergreen forest – deciduous forest	−161.39	43.78	135.01	−3.69	0.01	**
evergreen forest – tundra wetland	108.99	50.8	135.05	2.15	0.39	
evergreen forest – bog	197.26	52.35	132.92	3.77	0.01	**
evergreen forest – fen	324.15	40.47	141.6	8.01	0	***
evergreen forest – permafrost bog	67.88	55.25	144.04	1.23	0.92	
deciduous forest – tundra wetland	270.39	50.89	131.93	5.31	0	***
deciduous forest – bog	358.66	52.44	130.06	6.84	0	***
deciduous forest – fen	485.54	40.59	136.36	11.96	0	***
deciduous forest – permafrost bog	229.27	55.34	141.11	4.14	0	**
tundra wetland – bog	88.27	58.43	131.02	1.51	0.8	
tundra wetland – fen	215.15	48.07	136.01	4.48	0	***
tundra wetland – permafrost bog	−41.12	61.04	140.02	−0.67	1	
bog – fen	126.88	49.71	133.57	2.55	0.18	
bog – permafrost bog	−129.39	62.34	138.23	−2.08	0.44	
fen – permafrost bog	−256.27	52.76	145.84	−4.86	0	***

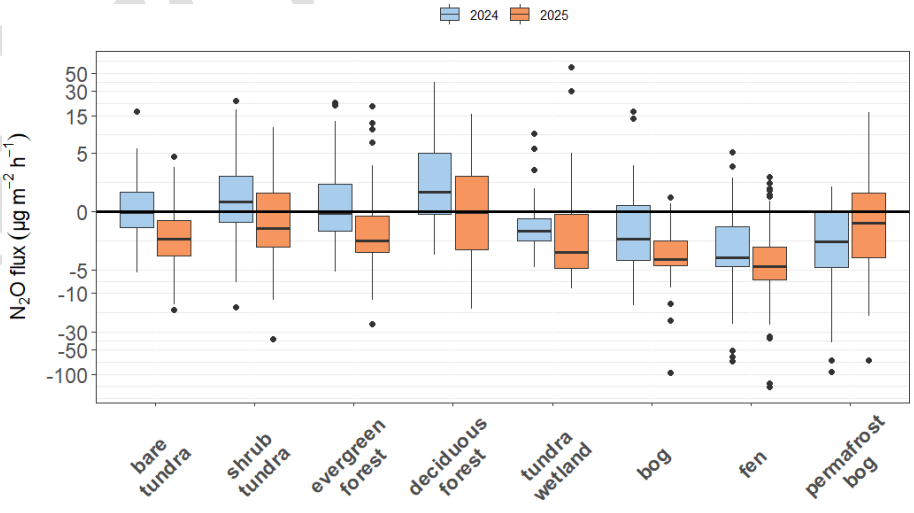


Figure A1. Yearly differences in N_2O fluxes between land cover classes. Colors indicate years, positive values are emissions and negative values are sinks. Note that permafrost bog category is not directly comparable between years due to additional plots being added in 2025. Boxplots have 25th and 75th quantiles, median (horizontal line), 1.5 IQR from the lower and higher quantiles (whiskers) and outliers (black dots). *Y* axis is log10 transformed.

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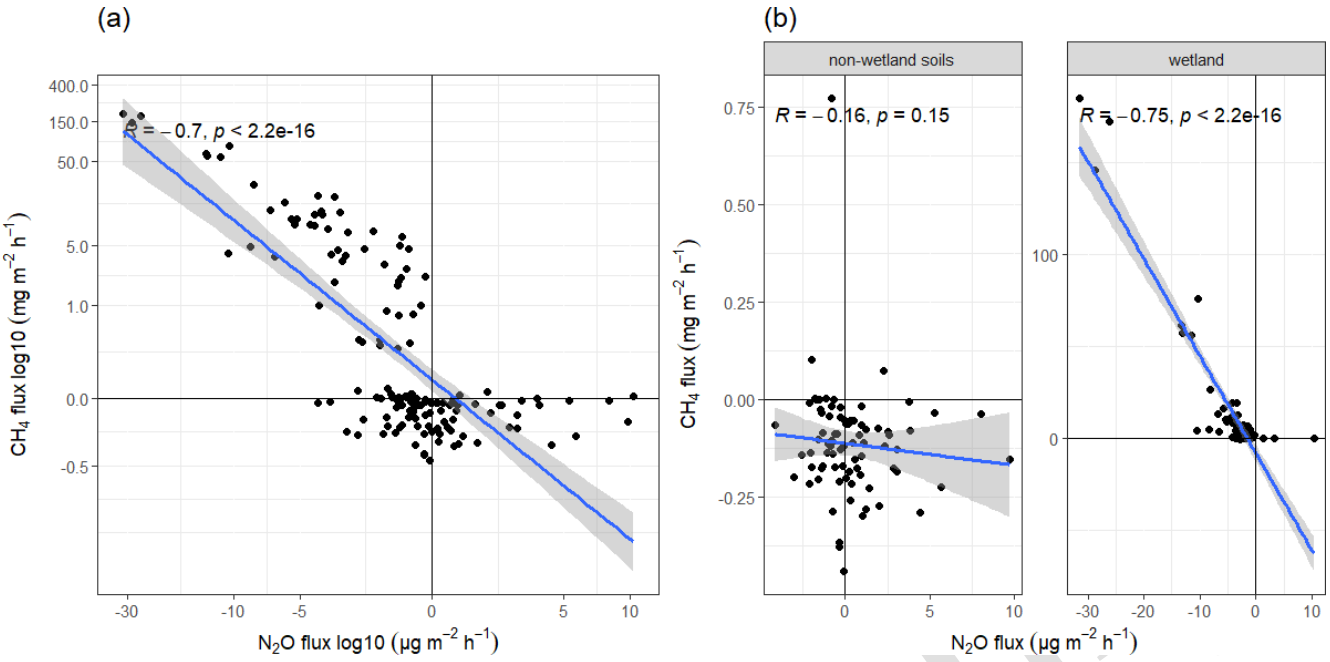


Figure A2. (a) Scatterplot of mean CH_4 flux (log10) and mean N_2O flux (log10) with spearman correlation **(b)** Scatterplot of mean CH_4 flux and mean N_2O flux between non-wetland soils (shrub tundra, bare tundra, deciduous forest, evergreen forest) and wetland soils (bog, fen, tundra wetland, permafrost bog) with spearman correlation.

Code and data availability. Data and R scripts used for the data visualizations and analysis can be provided by the corresponding authors upon request.

Python script for flux calculations available at https://github.com/janivaltteri/manual-chamber-dataserver/blob/main/dataserver/application/fluxcalc.pyTS28.

Author contributions. Conceptualization: ML, JH, JN, AMV; Methodology: ML, JN, JH, AMV; Formal analysis: JN; Investigation: JN, JL; Resources: JH, ML, AMV; Data curation: JN, JL; Writing – Original draft: JN; Writing – Review and Editing: JN, ML, JH, AMV, JL; Visualization: JN, JL; Supervision: ML, JH, AMV; Project administration: ML, JH, JN; Funding acquisition: JH, ML.

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