Methane exchange at the peatland forest floor – automatic chamber system exposes the dynamics of small fluxes

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Abstract. We measured methane (CH4) exchange rates with automatic chambers at the forest floor of a nutrient-rich drained peatland in 2011–2013. The fen, located in southern Finland, was drained for forestry in 1969 and the tree stand is now a mixture of Scots pine, Norway spruce, and pubescent birch. Our measurement system consisted of six transparent chambers and stainless steel frames, positioned on a number of different field and moss layer compositions. Gas concentrations were measured with an online cavity ring-down spectroscopy gas analyzer. Fluxes were calculated with both linear and exponential regression. The use of linear regression resulted in systematically smaller CH4 fluxes by 10–45 % as compared to exponential regression. However, the use of exponential regression with small fluxes (<2.5 µg CH4 m⁻² h⁻¹) typically resulted in anomalously large absolute fluxes and high hour-to-hour deviations. Therefore, we recommend that fluxes are initially calculated with linear regression to determine the threshold for “low” fluxes and that higher fluxes are then recalculated using exponential regression. The exponential flux was clearly affected by the length of the fitting period when this period was <190 s, but stabilized with longer periods. Thus, we also recommend the use of a fitting period of several minutes to stabilize the results and decrease the flux detection limit. There were clear seasonal dynamics in the CH4 flux: the forest floor acted as a CH4 sink particularly from early summer until the end of the year, while in late winter the flux was very small and fluctuated around zero. However, the magnitude of fluxes was relatively small throughout the year, ranging mainly from −130 to +100 µg CH4 m⁻² h⁻¹. CH4 emission peaks were observed occasionally, mostly in summer during heavy rainfall events. Diurnal variation, showing a lower CH4 uptake rate during the daytime, was observed in all of the chambers, mainly in the summer and late spring, particularly in dry conditions. It was attributed more to changes in wind speed than air or soil temperature, which suggest that physical rather than biological phenomena are responsible for the observed variation. The annual net CH4 exchange varied from −104 ± 30 to −505 ± 39 mg CH4 m⁻² yr⁻¹ among the six chambers, with an average of −219 mg CH4 m⁻² yr⁻¹ over the 2-year measurement period.

1 Introduction

Methane (CH4) is one of the most important atmospheric greenhouse gases due to its capability to absorb thermal radiation and warm the climate (IPCC, 2014). One of the main sources of CH4 globally is peatland (e.g., Denman et al., 2007), where CH4 is produced by the decomposition of organic matter in anaerobic conditions. Around 3 % (ca. 4 000 000 km²) of the Earth’s land surface is covered by peatlands (Clarke and Rieley, 2010) and the majority of these are located in the boreal region (Fischlin et al., 2007). About one third (104 000 km²) of European mire and peat resources are located in Finland (Montanarella et al., 2006) and more than half (55 000 km²) of this area has been drained for forestry (Päivänen and Hånell, 2012).

Methane can be both produced and consumed in soil so that the net CH4 flux depends on the rate of CH4 produc-
tion in anoxic soil layers and on the rate of CH$_4$ oxidation in the oxic soil layers. In peatlands, the thickness and depth of CH$_4$ producing and oxidizing layers are largely determined by the water table (Bubel and Moore, 1994), which controls the vertical distribution of oxygen in the soil profile. CH$_4$ is produced under anaerobic conditions by microbes known as methanogens. The production rate is dependent on the availability of organic substrates at low redox potential (Eh) values and is controlled by soil temperature and pH (Dunfield et al., 1993; Wang et al., 1993; Segers, 1998; Kotsyurbenko et al., 2004). In contrast, oxidation of CH$_4$ occurs in the oxic soil layer closer to the surface and potentially also in the moss layer (Larmola et al., 2010). The production rate, the oxidation rate is controlled by soil temperature and pH (Dunfield et al., 1993; Scheutz and Kjeldsen, 2004; Boeckx and Van Cleemput, 1996), but many other factors also affect oxidation processes, such as soil water content, soil texture, nutrients, and CH$_4$ and oxygen concentration (Boeckx and Van Cleemput, 1996; Ridgwell et al., 1999; Scheutz and Kjeldsen, 2004). In addition to the direct control of production and oxidation rates, there are other phenomena which may affect the observed net CH$_4$ flux above the soil surface, including lateral CH$_4$ transport in the soil (Christophersen and Kjeldsen, 2001) and subsurface storage (Hutchinson et al., 2000).

In environments with low soil CH$_4$ production, such as upland forest soils, grasslands, and tundra, uptake of atmospheric CH$_4$ by the methanotrophic microbes dominates (Dutaur and Verchot, 2007). This is also what commonly happens after the drainage of peatlands, which results in water level drawdown and increased oxic layer thickness. Thereby, CH$_4$ production is decreased and the fraction of oxidized CH$_4$ increased (e.g., Moore and Knowles, 1989; Roulet et al., 1992). Consequently, the CH$_4$ oxidation rate in the aerated surface soil and mosses typically exceeds CH$_4$ production that occurs deeper in the soil, thus turning in particular well-drained peatlands into net CH$_4$ sinks (Martikainen et al., 1995; Minkkinen et al., 2007; Ojanen et al., 2010; Lohila et al., 2011). However, poorly drained sites may remain to act as CH$_4$ sources (Ojanen et al., 2010). In addition, the drainage ditches even at well-drained sites typically continue to emit CH$_4$ at rates similar to pristine boreal peatlands (Minkkinen et al., 1997; Minkkinen and Laine, 2006; Luan and Wu, 2015).

Closed chambers are commonly used in the measurement of greenhouse gas exchange between the forest floor and the atmosphere (e.g., Denmead, 2008; Forbrich et al., 2010; Koskinen et al., 2014). Unlike the eddy covariance (EC) method, which is more suitable for measuring fluxes at the ecosystem level, the chamber method permits the investigation of small-scale processes, such as the gas exchange of different microtopographic surfaces, and enables the quantification of spatial variation (Keller et al., 1990; Singh et al., 1997). However, there are various details related to the chamber design and the deployment of this measurement technique in practice that may have a significant impact on the flux that is estimated from the observed concentration change in the chamber headspace. For example, the flux estimate seems to depend on the dimensions of the chamber (Pihlatie et al., 2013). In addition, chambers should include a fan to evenly distribute the air in the chamber headspace (Pumpanen et al., 2004; Christiansen et al., 2011) although the rotational speed of the fan should be kept low to avoid excessive speed (Koskinen et al., 2014). A major source of uncertainty is the impact of the chamber itself on the gas concentration gradient in the soil (Healy et al., 1996; Hutchinson et al., 2000; Conen and Smith, 2000; Davidson et al., 2002; Livingston et al., 2005) and in the boundary layer just above it. The concentration gradient is critical as it drives the soil–atmosphere gas exchange and thus any aerodynamic disturbance may impact the observed flux.

The gradient between the soil and the air inside the chamber changes when the gas concentration inside the chamber changes during the measurement. This changes the flux, which makes the concentration change nonlinear in time. However, nonlinearity of the concentration during the chamber closure may also result from chamber leaking. For example, Pirk et al. (2016) demonstrated that the degree of convex curvature in the increasing methane concentration correlated positively with wind speed (WS) outside the chamber. Furthermore, in the case of soil acting as a methane sink, the methane consumption by soil methanotrophs obeys the first-order reaction kinetics, which should lead to curvilinear concentration dynamics in the chamber (e.g., Sabrekov et al., 2016). However, the different processes responsible for the curvature in the concentration time series may be difficult to separate from each other (Kutzbach et al., 2007).

There are many studies that have recognized that the use of linear regression in flux calculation can cause significant underestimation of the flux (e.g., Healy et al., 1996; Hutchinson et al., 2000; Nakano et al., 2004; Livingston et al., 2005, 2006; Kutzbach et al., 2007; Kroon et al., 2008; Pedersen et al., 2010; Pihlatie et al., 2013). However, many studies have used linear regression (e.g., Laine et al., 2006; Alm et al., 2007; Jones et al., 2011; Bergier et al., 2013; Fassbinder et al., 2013), because under field conditions it is more robust to random measurement errors than nonlinear methods. Moreover, the use of linear regression is preferred when comparing measurement sites as it is not as sensitive as nonlinear models to small differences in soil properties (Venterea et al., 2009). The selection of the optimal fitting method is important as it can be a large source of uncertainty in flux calculations (Levy et al., 2011; Venterea, 2013).

Although several studies have examined the different fitting methods for calculating fluxes from chamber data, there exist only a few papers exploring the dynamics of CH$_4$ flux data that mainly consist of small uptake fluxes and are measured with automatic chambers coupled to a high-resolution gas analyzer (e.g., Savage et al., 2014; Ueyama et al., 2015). In this study, we measured the CH$_4$ flux between a forest.
Table 1. CN ratio, bulk density, and ash content (±SD) of the peat at Lettosuo (n = 4).

<table>
<thead>
<tr>
<th>Layer</th>
<th>CN ratio (±SD)</th>
<th>Bulk density (g cm⁻³)</th>
<th>Ash content (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Humus</td>
<td>29.2 ± 1.8</td>
<td>0.005 ± 0.003</td>
<td>3.1 ± 0.4</td>
</tr>
<tr>
<td>0–10 cm</td>
<td>23.9 ± 1.0</td>
<td>0.107 ± 0.014</td>
<td>6.5 ± 1.5</td>
</tr>
<tr>
<td>10–20 cm</td>
<td>24.3 ± 0.7</td>
<td>0.170 ± 0.011</td>
<td>3.4 ± 0.4</td>
</tr>
</tbody>
</table>

floor and the atmosphere continuously throughout 2 years at a boreal nutrient-rich forestry-drained peatland site with typically small CH₄ exchange rates. We used six automatic soil chambers and a cavity ring-down spectroscopy analyzer, which allowed us to measure with a relatively high sampling rate during each chamber closure and to perform multiple daily measurements with each chamber. Our particular aims with this setup were to answer the following questions:

1. What is the optimal fitting method for calculating the CH₄ flux?

2. How large are the diurnal, seasonal, and interannual variations in the CH₄ flux?

3. What is the annual CH₄ balance of the study site?

2 Materials and methods

2.1 Site description

The measurements were made in southern Finland at Lettosuo (60°38’N, 23°57’E) (Fig. S1a in the Supplement), which is a nutrient-rich peatland forest that was drained in 1969 and fertilized with phosphorus and potassium soon after. The open ditches, located in approximately 45 m intervals (Fig. S1b), were originally about 1 m deep but have since been partly filled with new vegetation. Before drainage, the tree stand was dominated by Scots pine (Pinus sylvestris) with some pubescent birch (Betula pubescens). After drainage, the stand has developed to a mixture of Scots pine and pubescent birch in the dominant canopy layer, with an understorey of Norway spruce (Picea abies) with some scattered small-sized pubescent birch. The vessel volumes at the time of this study equaled 174, 46, and 28 m³ ha⁻¹ for Scots pine, pubescent birch, and Norway spruce respectively. The tree stand is quite dense, which results in irregular shading and, consequently, patchy and variable ground vegetation layer. For example, herbs such as Dryopteris carthusiana and Trifolium pratense and dwarf shrubs such as Vaccinium vitis-idaea are common in the ground vegetation (Bhuiyan et al., 2017). In addition, the moss layer is patchy and is dominated by Pleurozium schreberi and Dicranum polysetum with some Sphagna (Sphagnum girgensohnii, Sphagnum angustifolium, and Sphagnum russowii) appearing in moist patches.

CN ratio of the surface peat, sampled at four points located at a 20–40 m distance from the chamber plots, averaged 24 for the 0–20 cm layer (Table 1). The relatively low CN ratio is typical for fertile peatland forests and reflects the fen history of the site. The bulk density of these samples was 0.11 and 0.17 g cm⁻³ for the 0–10 and 10–20 cm layers, respectively, while the average bulk density of the 0–20 cm layer below each chamber varied from 0.03 to 0.13 g cm⁻³ (Koskinen et al., 2014). The ash content of the peat varied from 3.4 to 6.5%.

The vascular green area (VGA) was estimated for each chamber and vascular plant species every 2 weeks during the growing seasons 2011 and 2012 (Ojanen, unpublished data). This was done by estimating the number and dimensions of leaves within each chamber and calculating green area by species-specific regression models between leaf dimensions and green area. For Vaccinium myrtillus, the surface area of the green stems was also included into VGA. The coverage of the mosses was estimated visually. The maximum VGA and the coverages for each chamber are shown in Table 2.

<table>
<thead>
<tr>
<th>Chamber</th>
<th>Vegetation</th>
<th>VGAmax</th>
<th>CFM</th>
<th>CSp</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Pleurozium schreberi</td>
<td>2.04</td>
<td>56</td>
<td>0</td>
</tr>
<tr>
<td>2</td>
<td>Dicranum polysetum</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>Vaccinium myrtillus</td>
<td>0.85</td>
<td>60</td>
<td>0</td>
</tr>
<tr>
<td>4</td>
<td>Maianthemum bifolium</td>
<td>0.01</td>
<td>3</td>
<td>0</td>
</tr>
<tr>
<td>5</td>
<td>Dryopteris carthusiana</td>
<td>2.34</td>
<td>26</td>
<td>0</td>
</tr>
<tr>
<td>6</td>
<td>Vaccinium vitis-idaea</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Pleurozium schreberi</td>
<td>0.11</td>
<td>30</td>
<td>0</td>
</tr>
<tr>
<td>7</td>
<td>Dicranum polysetum</td>
<td>0.85</td>
<td>60</td>
<td>0</td>
</tr>
<tr>
<td>8</td>
<td>Sphagnum girgensohnii</td>
<td>–</td>
<td>0</td>
<td>90</td>
</tr>
</tbody>
</table>

from the 25.5 m tall EC mast (Koskinen et al., 2014) from which the WS above the canopy was measured. The locations of the chambers were selected to maximize the number of different ground vegetation compositions (Table 2) within a circle of ca. 15 m radius around the cabin.

The details of the chamber system can be found in Koskinen et al. (2014), and thus here we only describe the main features of the system. The size of the chamber boxes were 57 cm × 57 cm × 30 cm (length × width × height). We used a permanently installed steel collar (height 5 cm, inserted at a depth of 2 cm) below each chamber to minimize the disturbance to the soil and to enhance the sealing between the soil and the chamber. There was a U profile at the bottom of the chamber edges, insulated with a foam tape, to further improve the sealing. In winter, the whole chamber frame was raised above the snow level by placing one or two extension collars (height 16 cm) between the frame and soil.

A 24 V fan (Maglev KDE2408PTV1, Sunon Ltd, Kaohsiung, Taiwan) (size 8 cm × 8 cm) was used to mix the air inside the chamber headspace. The voltage of the fan was regulated to keep the mixing steady, but as low as possible (Koskinen et al., 2014). Sample gas was drawn from the chamber typically once an hour (with some exceptions explained below) at a flow rate of about 1 L min⁻¹ and returned back to the chamber from the gas analyzers. CH₄, CO₂, and water vapor concentrations were measured approximately every 4 s with a Picarro G1130 cavity ring-down spectroscopy gas analyzer (Picarro Inc., Santa Clara, CA, USA). The inlet and outlet gas tubes (Festo Oy, Vantaa, Finland) were made of polyurethane and were 15 m in length and had an inner and outer diameter of 4 and 6 mm respectively.

The tubes were flushed with ambient air just before the chamber was closed. When all the chambers were open, ambient air was sampled. The delay in the analyzer response caused by the long tubing was taken into account using a flagging system in the computer program collecting the data, which labeled each data point with the respective chamber number using a 20 s lag. However, as the flow rate varied slightly in time, some points were removed from the data before the flux calculation (Sect. 2.4).

Air and soil temperature data were collected every 10 s using Pt100 probes (PT4T, Nokeval Oy, Nokia, Finland) and Nokeval 680 loggers (Nokeval Oy, Nokia, Finland). One probe was located inside each chamber at a height of 30 cm and positioned next to the fan under a metal heat shield to prevent direct solar radiation from affecting the measurements. Furthermore, soil surface temperature was monitored inside each chamber just below the surface of the moss or litter layer. In addition, soil temperature probes were placed at depths of 2, 5, 10, 20, and 30 cm at one location near the chambers. Water table level (WTL) was monitored every hour at four different points at the site (TruTrack WT-HR data loggers, Intech Instruments Ltd, Auckland, New Zealand). The air pressure, precipitation, and snow depth data were acquired from the nearby Finnish Meteorological Institute observatory at Jokioinen (~35 km northwest of Lettosuo).

2.3 Meteorological conditions

The climate at the site has both continental and maritime influences. The annual mean temperature and precipitation at the nearby weather station in 1981–2010 were 4.6 °C and 627 mm respectively (Pirinen et al., 2012). During this study, the first measurement year (April 2011–March 2012) was significantly warmer (annual mean temperature 5.8 °C) than the second measurement year (April 2012–March 2013) (1.4 °C) (Fig. 1). The first year was slightly warmer and the second year was significantly colder than the long-term mean recorded at the nearby weather station (4.6 °C). Both the summer (JJA) (17.6 °C) and winter (DJF) (2.7 °C) temperatures in 2011 were warmer than those of 2012 (12.1 and −2.3 °C). In particular, the beginning of summer 2012 was much colder than the same period in 2011.

Annual precipitation during the first (976 mm) and second (780 mm) measurement years was higher than the long-term mean (627 mm). Summertime precipitation was 9% higher in the first (309 mm) year as compared to the second (284 mm) year, while in winter the difference was 18% (577 and 490 mm in the first and second winters respectively). The first snow appeared on 5 December in 2011 and 25 October in 2012, and the first permanent snow was recorded on 7 January in 2012 and 28 November in 2012. In spring (MAM) 2011, the snow had melted by 13 April. For spring 2012, we do not know the exact day of snowmelt due to missing data, although the snow had melted at latest by 4 April. From the temperature data we estimate that the snow cover disappeared sometime in mid-March.

WTL varied from −8 to −59 cm from the soil surface (negative sign denotes WTL below the surface) and was highest in the spring and late autumn (SON). The lowest (i.e., deepest) values were recorded at the end of summer. The average WTL in summer 2011 was −47.2 ± 7.4 cm (±SD) and −49.1 ± 7.1 in summer 2012. Occasional sudden increases in WTL were observed after rainfall events and it usually took 1–2 weeks to reach the WTL observed prior to the event.

2.4 Flux calculation

During the study period, the chambers were operated with varying closure times ranging from 2 to 16 min. In 2011, 2 min closures were used with the exception of 6 min measurements made four times per day. After mid-March 2012, the minimum closure time was 6 min. Thus, each chamber was typically sampled once an hour, with the exception of summer 2012 (JJA) when a longer closure time of 16 min was tested and each chamber was sampled every 2 h. For the analysis of CH₄ exchange dynamics (Sect. 3.3–3.4), we used the fluxes calculated with a 6 min closure time (as justified in Sect. 3.2). In addition to removing 20 s from the start of the
measurement due to lag caused by long tubing (Sect. 2.2), 18 s was discarded from the start of a measurement to ensure that the air inside the chamber was properly mixed. Dilution and spectral corrected CH₄ concentrations reported by Picarro G1130 were used to calculate the fluxes.

Two different regression types were fitted to the data: linear and exponential. The linear function describing the change in the concentration, C, as a function of time was

$$C(t) = a_{\text{lin}} + b_{\text{lin}}t,$$

where $a_{\text{lin}}$ and $b_{\text{lin}}$ are parameters and $t$ is the time from the start of the closure. In this model, the slope $b_{\text{lin}}$ equals the concentration change in time.

The exponential function we fitted was

$$C(t) = a_{\text{exp}} + b_{\text{exp}} \exp(c_{\text{exp}}t),$$

where $a_{\text{exp}}$, $b_{\text{exp}}$, and $c_{\text{exp}}$ are parameters. When differentiating Eq. (2) with respect to time and inspecting the moment when the chamber closes ($t = 0$), it follows that the concentration change with time is the product of parameters $b_{\text{exp}}$ and $c_{\text{exp}}$. It is generally considered that this initial rate of concentration change best represents the flux at that time. However, when fitting the exponential function to the data using the least-squares approach, the fitting frequently fails due to local minima. To overcome this and to avoid over-parameterization, a Taylor power series expansion (Kutzbach et al., 2007) was fitted to the data to determine initial estimates of the parameters of the exponential regression.

The exponential regression should capture the flux better than the linear regression as it takes into account the change in the gradient between soil and chamber headspace during chamber closure, which is evident when diffusion flux is decreasing the concentration difference. However, exponential regression is very sensitive to possible disturbances to the data at the beginning of chamber closure. In our study, we attempted to minimize these disturbances by closing the chamber slowly and smoothly, which seemed to prevent pressure fluctuations related to chamber closing. For the analysis of CH₄ exchange dynamics (Sect. 3.3–3.4), we used flux data that are based on a combination of linear and exponential fits: first all fluxes were calculated using the linear regression, and below and above a limit of 2.5 µg CH₄ m⁻² h⁻¹ the fluxes were calculated with the linear and exponential method respectively (for justification see Sect. 3.1).

The CH₄ flux ($F$, µg CH₄ m⁻² h⁻¹) was calculated according to Eq. (3), which is based on the ideal gas law:

$$F = \left( \frac{dC(t)}{dt} \right)_{t=0} \frac{MPV}{RTA} \frac{3600 s}{h},$$

where $\left( \frac{dC(t)}{dt} \right)_{t=0}$ is the time derivative (ppm s⁻¹) of a linear ($b_{\text{lin}}$) or exponential ($b_{\text{exp}} \times c_{\text{exp}}$) regression at the beginning of the closure, $M$ is the molecular mass of CH₄ (16.042 g mol⁻¹), $P$ is air pressure (Pa), $R$ is the universal gas constant (8.31446 J mol⁻¹ K⁻¹), $T$ is the mean chamber headspace temperature during closure (K), and $V$ and $A$ are the volume (m³) and the base area (m²) of the chamber headspace respectively. Here, a micrometeorological sign convention is used: a positive flux indicates a flux from the ecosystem to the atmosphere (emission) and a negative flux indicates a flux from the atmosphere into the ecosystem (uptake).
When estimating the volume of the chamber headspace, the height of the moss and snow surfaces was assumed to represent the interface between the soil and air. In other words, the pore space in the soil and snow was ignored from the headspace volume. The error caused by this in flux calculations was estimated to be only a few percent (Koskinen et al., 2014). To create a continuous data set of snow depth, the manual measurements carried out irregularly at the site were combined with those measured daily at the Jokioinen observatory. In addition to snow depth, the height of the chamber headspace was measured at the start and end of the growing season from 16 points inside each collar by gently placing the end of a tape measure on top of the surface mosses (Koskinen et al., 2014). The height of the chamber headspace between these manual measurements was determined with linear interpolation.

All the calculations and analyses were made with the Python programming language (Python Software Foundation, version 2.7, https://www.python.org) using the following libraries: NumPy (http://www.numpy.org/), SciPy (http://www.scipy.org/), Pandas (http://pandas.pydata.org/), and matplotlib (http://www.matplotlib.org). All the Python scripts were developed specifically for this study. For the fits, the least-squares method was used through the “polyfit” function of NumPy library for the linear regression and the “curve_fit” function of SciPy library for the nonlinear fits.

2.5 Filtering of the flux data

After the fluxes were calculated, several filters were applied to remove cases where the measurement system did not work adequately. The most common reason for discarding data was due to the problems with the chamber operation, for example, for the improper functioning of a linear actuator, which caused the chambers to remain stuck either open or closed. These cases were detected by monitoring the simultaneously measured CO₂ concentration data during the closure. The goodness of fit was checked by calculating the normalized root mean square error (NRMSE) (e.g., Christiansen et al., 2011; Pihlatie et al., 2013) for each fit:

\[ \text{NRMSE} = \sqrt{\frac{1}{n} \sum_{i=1}^{n} \left( \frac{C_{\text{fit},i} - C_i}{C_{\text{max}} - C_{\text{min}}} \right)^2} \]

where \( n \) is the number of measurement points, \( C_{\text{fit},i} \) is the CO₂ concentration calculated from the fit, \( C_i \) is the measured CO₂ concentration, and \( C_{\text{max}} \) and \( C_{\text{min}} \) are the highest and lowest concentrations measured during closure. If the NRMSE was larger than 0.05, the CH₄ data from that closure were discarded. It should be noted that the application of this criterion removes closures with no change in CO₂ concentration, which may result when photosynthesis rate equals respiration rate. Here we found < 20 of such cases, meaning that this criterion could be applied without removing a significant amount of potentially suitable data.

In addition to NRMSE filtering, the running mean of CH₄ flux (\( F_{\text{CH}_4} \)) with a time window of 14 days (shifting one day at a time) and the corresponding standard deviation (\( \sigma \)) were calculated to remove random spiking in the data. The data points that failed to fall within \( F_{\text{CH}_4} \pm 10\sigma \) were removed iteratively. In total, 71229 closures were recorded from which 14 % (\( n = 9987 \)) were discarded due to large NRMSE values (problems with the chambers) and < 0.001 % (\( n = 40 \)) were removed with the \( \sigma \) filter as outliers.

2.6 Detection limit

The minimum detectable flux (MDF) was estimated by using the metric originally developed by Christiansen et al. (2015), which was modified by Nickerson (2016) to make it more suitable for high-frequency measurements:

\[ \text{MDF} = \frac{\text{NRMSE}}{\sqrt{\frac{1}{n} \sum_{i=1}^{n} \left( C_{\text{fit},i} - C_i \right)^2}} \]

Figure 2. Concentration data during one chamber closure for a case with a higher (a linear and exponential: −90 and 104 μg CH₄ m⁻² h⁻¹ respectively) and lower (b lin −3.5, exp −4.3 μg CH₄ m⁻² h⁻¹) flux.
MDF = \left( \frac{P_I}{t_c/\sqrt{p_s}} \right) \left( \frac{V P}{ART} \right) M, \quad (5)

where \( P_I \) is the analytical precision of the instrument (ppm), \( t_c \) is the closure time of the chamber (h), and \( p_s \) is the sampling periodicity (h). The \( P_I \) for the Picarro G1130 analyzer, tested and reported by the manufacturer for the specific instrument used in this study, was 0.256 ppb and \( p_s \) was 5 s. On a typical summer day (\( T = 20^\circ C \)), the MDF of the system was about 0.06 \( \mu g \) CH\(_4\) m\(^{-2}\) h\(^{-1}\). However, during winter the MDF was higher due to lower temperatures and the use of the extension collars, which together about double the headspace volume (without snow) and therefore also the MDF.

2.7 The annual balance and its uncertainty

The annual balance of CH\(_4\) was estimated for each chamber by first calculating the daily flux sums from the hourly fluxes and then summing these over a year. The gaps in the data were filled by using linear interpolation between the existing hourly and daily fluxes. As most of the fluxes in 2011 and in the first quarter of 2012 were measured with a 2 min closure time, which was considered too short for the exponential regression (Sect. 3.2), we corrected the fluxes calculated with linear regression from the 2 min closures to correspond to those measured using a 6 min closure available four times a day during this period as a reference. This correction was implemented by calculating the daily median ratios between the fluxes from 6 and 2 min closure times, which were smoothed by a running median with a moving window of 14 days. Finally, the 2 min data from 2011 to March 2012 were multiplied by this ratio (Fig. S2).

The uncertainty of the CH\(_4\) balance estimate derived from the measurements was evaluated by identifying three key error sources: (1) the random error of regression, (2) the error caused by gap filling, and (3) the error caused by the correction of the fluxes measured using the 2 min closure time during the first measurement year. First, because the annual balance of each chamber was calculated from the mean daily fluxes, we estimated the daily random error as the squared sum of the uncertainties of the hourly flux data of each day. Assuming that the goodness of fit reflects all the uncertainties related to a single flux measurement, the standard deviation of the slope estimate obtained \( \left( \frac{dC(t)}{dt} \right)_{t=0}; \text{ Eq. 3} \) provides a measure of this uncertainty.

Next, the error caused by the gap-filling procedure was estimated by removing 1 month of flux data from different parts of the whole data set and inspecting how this affected the annual balance of the different chambers. The average value of the effect of these monthly gaps was calculated and downscaled to represent the effect of one missing day. Multiplying this value by the number of missing days during the year gives an estimate of the gap-filling error. It must be noted that the length of the removed period was similar to the longest gap observed in our data.

Last, the error estimate related to the ratio used to transform the fluxes calculated from 2 min closure to represent the 6 min closure was estimated from the median absolute deviation assuming normally distributed medians. Finally, these three error estimates were added together by using the standard accumulation principle of independent errors.
As our measurements started in April 2011 and ended in March 2013, after exactly 2 years, from now on in this paper the expressions “first year” and “second year” denote the time periods of April 2011–March 2012 and April 2012–March 2013 respectively.

3 Results

3.1 Flux calculation method

Examples of typical concentration development inside a chamber during one measurement are shown in Fig. 2, for both “high flux” case in summer (Fig. 2a) and a “low flux” case in winter (Fig. 2b). In summer and autumn, when the fluxes were the highest, the concentration development inside a chamber usually was not adequately approximated by a linear function and thus the slope calculated with the linear regression (Eq. 1) did not properly represent the initial “undisturbed” slope from which the flux should be calculated. As a result, linear regression resulted in lower flux estimates for these cases than exponential regression (Eq. 2). However, during the periods when the flux approached the detection limit and the concentration data became noisier, the use of exponential regression resulted in noisier flux data. Often, exponential regression created a sharp slope at the beginning of the fit in the concentration time series that resulted in unphysically high fluxes. To be able to reliably estimate the CH$_4$ exchange for the whole range of fluxes, we determined the flux limit below which the exponential regression resulted in unreliable flux estimates and the linear fit should be preferred. This limit was estimated by comparing bin ($n = 500$) averages of linear and exponential fluxes for the whole data set (using a 6 min closure time) (Fig. 3). When the linearly calculated fluxes fell below ca. 2.5 µg CH$_4$ m$^{-2}$ h$^{-1}$, the noise in the flux calculated using the exponential regression increased steeply and the shape of the relationship changed (Fig. 3). Therefore, we decided to first calculate the flux with the linear regression and to recalculate all the fluxes exceeding the limit of 2.5 µg CH$_4$ m$^{-2}$ h$^{-1}$ with the exponential regression. Henceforth, all the data shown in this paper have been calculated in this way unless stated otherwise.

The whole 2-year data set showed that the CH$_4$ fluxes calculated with linear regression (Eq. 1) were systematically and significantly lower than those calculated with exponential regression (Eq. 2) (Table 3). The seasonal average flux
Figure 6. Hourly CH$_4$ fluxes from April 2011 to March 2013 measured in each chamber. Negative values indicate uptake by the soil, and positive values indicate emission to the atmosphere. Fluxes have been calculated using the exponential fit unless the value of the flux obtained from the linear fit was below 2.5 µg CH$_4$ m$^{-2}$ h$^{-1}$.

difference between the linear and exponential regressions varied within 10.9–44.4 % (average over 2 year 27.5 ± 0.3 %, ±95 % confidence interval). The mean relative difference was dependent on the time of the year: it was largest during the winter and spring (24.9–44.4 %) when the soil CH$_4$ sink was at its lowest and smallest in summer and autumn (10.9–14.4 %) when the sink was at its highest. When comparing individual measurements, the average relative difference between the linear and exponential regression was slightly smaller in 2012 compared to 2011. Also, the uncertainties associated with the fluxes were slightly larger in 2011 due probably to the fewer measurements available with 6 min closure time.

3.2 Effect of closure time on fluxes

The effect of the different fitting time windows was tested by both increasing the fitting period from the beginning of the closure with 10 s steps and by keeping the fitting window constant but moving its starting point. For these tests, we used the data from summer 2012, when the measurements were made with a 16 min closure time. The flux from the exponential fit was clearly affected by the length of the fit when the fitting period was < 190 s (Fig. 4b). After that, the mean difference was mostly statistically insignificant ($p > 0.05$), as compared to the flux calculated with the 900 s period. However, the estimated linear flux stayed about the same for the first 140 s resulting in 16.2 ± 0.6 % higher fluxes than obtained with the 900 s fitting window (Fig. 4a). However, further increase of the fitting period systematically decreased the estimated flux by about 1.3 % per 60 s. A decrease of 17.3 ± 3.0 % was also observed when the starting point of the fit was delayed by 530 s, but the fitting period was kept constant at 6 min (Fig. 5).

Even though the results above might support the selection of a fitting period of 190 s, a 6 min fitting period was applied in further analysis. This was selected based on three arguments: (1) it made the exponential regression results more stable; (2) we wanted to use the same fitting period in both linear and exponential regressions; and (3) a longer fitting period decreases the detection limit (Eq. 5). The last point was mainly related to winter measurements when the detection limit was increased by lower temperatures and the use of extension collars (increasing the effective volume before the collars were filled with snow). However, in 2011 and in the first quarter of 2012, a 2 min closure time was mostly used, which proved to be too short for accurate estimates with the exponential fit. As a result, the results from these shorter measurements were corrected to correspond to those obtained with the 6 min closure time (Sect. 2.7).
Table 3. Seasonal (summer is JJA, autumn is SON, winter is DJF, spring is MAM) averages of CH$_4$ flux (µg CH$_4$ m$^{-2}$ h$^{-1}$) calculated with linear (“linear flux”) and exponential regression (“exponential flux”) with 95 % confidence intervals (±). The difference shows how much smaller the linear flux was on average when compared to the exponential flux. The data include the fluxes measured by all six chambers with a 6 min closure time.

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<tr>
<td>Meanlinear flux</td>
<td>$-38.9 \pm 1.3$</td>
<td>$-39.0 \pm 1.2$</td>
<td>$-32.2 \pm 0.7$</td>
<td>$-31.3 \pm 0.4$</td>
<td>$-45.1 \pm 1.8$</td>
<td>$-44.9 \pm 1.3$</td>
<td>$-34.9 \pm 0.4$</td>
<td>$-21.0 \pm 0.3$</td>
<td>$-22.7 \pm 0.4$</td>
<td>$-27.5 \pm 0.3$</td>
</tr>
<tr>
<td>Noclosures</td>
<td>1558</td>
<td>3685</td>
<td>1172</td>
<td>10642</td>
<td>3685</td>
<td>1172</td>
<td>942</td>
<td>1029</td>
<td>1272</td>
<td>1018</td>
</tr>
<tr>
<td>Difference (%)</td>
<td>14.4 ± 0.5</td>
<td>11.7 ± 0.3</td>
<td>14.2 ± 0.5</td>
<td>10.9 ± 0.2</td>
<td>24.9 ± 1.0</td>
<td>29.8 ± 0.8</td>
<td>24.9 ± 1.0</td>
<td>29.8 ± 0.8</td>
<td>24.9 ± 1.0</td>
<td>29.8 ± 0.8</td>
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3.3 Seasonal dynamics of CH$_4$ flux and comparison between the chambers

During the 2-year measurement period, CH$_4$ fluxes varied mainly between $-120$ and $+20$ µg CH$_4$ m$^{-2}$ h$^{-1}$ (Figs. 6 and 7). Higher uptake rates (70 µg CH$_4$ m$^{-2}$ h$^{-1}$) were measured more often during the first year (Fig. 7a) and during summer 2011 (Fig. 7b) than in the second year and in summer 2012. The number of emission cases was low in both years and the soil acted as a CH$_4$ sink for most of the time in all chambers, although a few emission peaks of up to 200 µg CH$_4$ m$^{-2}$ h$^{-1}$ were recorded during and following heavy rainfall events. The data filtering (Sect. 2.5) was designed to be flexible to prevent the removal of these short-lasting CH$_4$ bursts from the accepted data set. While some emission peaks were ob-
CH$_4$ fluxes showed a clear seasonal variation. In spring, when the snow melted and thawing of the soil surface had started, CH$_4$ emissions fluctuated around 2 µg CH$_4$ m$^{-2}$ h$^{-1}$ in all chambers (Fig. 6). As the temperature rise continued, increasing CH$_4$ uptake was observed. In both years, CH$_4$ uptake was largest in August, when fluxes varied between −30 and −130 µg CH$_4$ m$^{-2}$ h$^{-1}$. In September, the uptake decreased and by the end of November it had dropped to half of that in the summer. However, the soil acted as a sink until the soil surface froze, after which CH$_4$ fluxes fluctuated around zero.

Considerable systematic differences in fluxes between the chambers were detected. The largest 2-year average sinks were measured in chambers 5 and 2 (Fig. 8), which were dominated by forest mosses Pleurozium schreberi and Dicranum polysetum (Table 2). The third largest sink was observed in chamber 6 (Fig. 8) with a Sphagnum sp. carpet, while the remainder of the chambers showed similar fluxes. Annual net CH$_4$ exchange rates were on average −267 ± 55 and −172 ± 31 mg CH$_4$ m$^{-2}$ yr$^{-1}$ (± standard error of the mean) in 2011–2012 and 2012–2013 respectively. The forest floor sink was 10–44 % lower during the second than first annual period and this difference was statistically significant (p < 0.05) for all the chambers except chamber 4. The largest absolute year-to-year reduction in annual CH$_4$ exchange was observed in the two chambers dominated by Pleurozium schreberi and Dicranum polysetum (5 and 2), followed by the chamber dominated by Sphagnum sp. (6). However, the largest relative decrease in net CH$_4$ exchange happened in chamber 6 (44 %) while in the rest of the chambers the decrease varied within 10–37 %.

The sink period, i.e., the period when all chambers acted as CH$_4$ sinks (daily mean flux < −2 µg CH$_4$ m$^{-2}$ h$^{-1}$), was slightly longer in the first than in the second year. In the first year, all chambers acted as sinks from 14 May 2011 to 20 February 2012, a total of 282 days. During the second year, the sink period lasted for 269 days from 10 May 2012 until 3 February 2013. In addition, chamber 6 acted as a sink over the whole winter and spring in 2012, while from chambers 2 and 5 daily mean fluxes of < −2 µg CH$_4$ m$^{-2}$ h$^{-1}$ were measured already at the start of the study in April 2011.

3.4 Factors controlling the short- and long-term variations in CH$_4$ exchange

3.4.1 Seasonal cycle

There was an observable, exponential relationship between the mean daily CH$_4$ fluxes and the deeper soil temperatures when the data from the spring and summer of 2011 were pooled (Fig. 9a). Splitting these data into shorter periods showed that the relationship was rather strong in April–May and in June–early July, but after that, when the soil temperature at 30 cm depth exceeded 12 °C, the relationship was absent. It is evident that there is some covariation between soil temperature and WTL with typically higher CH$_4$ uptake taking place at lower WTL. The plot between the residuals of the temperature response against WTL (Fig. 9b) suggests that the variation in the CH$_4$ uptake during the latter half of July and August was better explained by the WTL than temperature. All the chambers recorded similar behavior in 2011 and 2012, although in 2012 the data were noisier and the relationships observed in 2011 were not as clear.
3.4.2 Diurnal cycle

In addition to seasonal dynamics, diurnal variation was observed in CH₄ fluxes at least occasionally in all chambers, mostly in May and in the first half of June. Such variation was more common in 2012 and was observed in all of the chambers, while in 2011 the variation occurred mostly in chambers 2, 5, and 6. For example, during the first 2 weeks of June 2011 a clear diurnal cycle coinciding with the variation in air temperature was observed (Figs. 10 and 11a–b). Higher CH₄ uptake was observed during the night and morning hours, while the sink decreased towards the midday and started to grow again towards the night (Fig. 11a). This particular 2-week period was associated with high daytime air temperatures, reaching almost +30°C, and a relatively low WTL (Figs. 10 and 11a–b). After that, at about mid-June, the weather type changed to cool and wet and the diurnal variation was diminished or absent (Figs. 10 and 11d–e). Such behavior, illustrated in Fig. 10, was typical for the rest of the growing season data: diurnal variation occurred more often with dry weather, while during and after the rain the variation ceased for a while. However, WTL itself did not have an impact on the diurnal cycle.

Pooling the data of these 2-week periods into hourly means implies that the soil temperature may exert a strong control on the CH₄ exchange (Fig. 11). However, as temperature often – though not always – tends to correlate with wind speed, particularly in the summer, it is necessary to consider the effect of both these variables. The correlations calculated from hourly data indicate that it was the wind speed and not the temperature which played the major role in causing the diurnal variation in CH₄ flux (Figs. S6–S27). However, this relationship was not comprehensive, as the correlation with wind speed was absent during some periods. In qualitative terms, we observed that the drier the soil was, the greater the impact of wind speed.

To explain this correlation, we investigated the relationship between the parameter $c_{\text{exp}}$ (Eq. 2) and ambient WS in 2-week to 1-month periods to diminish the possible impact of seasonality (Figs. S28–S40). $c_{\text{exp}}$ represents the curvature of concentration time series during each chamber closure. For example, the $c_{\text{exp}}$ determined for chamber 2 became less negative when WS exceeded 2 m s⁻¹. This means that the curvature was weaker for cases of high WS.

3.4.3 Correlation between the flux and meteorological variables

To understand the general driving factors behind all the variation, either diurnal or seasonal, Pearson correlation coefficients between the hourly CH₄ flux and key environmental variables were calculated for different seasons, averaging the correlations determined separately for each chamber (Table 4).
Table 4. Correlation coefficients ($r$) between hourly CH$_4$ flux and environmental variables averaged for six chambers during different seasons. All six chambers with $p < 0.01$ are bolded and underlined; five chambers with $p < 0.01$ are bolded; four chambers with $p < 0.01$ are underlined; three chambers with $p < 0.01$ are italicized. AirT is air temperature, ST is soil surface temperature, PAR is photosynthetically active radiation, WS is wind speed measured above the canopy, $T_x$ is soil temperature at a depth of $x$ cm, and WTL is water table level. For clarity, negative $r$ values typically denote situations where CH$_4$ uptake increases when the value of explaining variable increases. The season definitions are the same as in Table 3.

\[
\begin{array}{ccccccccccc}
\text{Season} & \text{AirT} & \text{ST} & \text{PAR} & \text{WS} & \text{T2} & \text{T5} & \text{T10} & \text{T20} & \text{T30} & \text{WTL}\ast \\
\hline
\text{Spring 2011} & -0.03 & -0.04 & -0.18 & -0.08 & -0.41 & -0.52 & -0.61 & -0.67 & -0.70 & 0.52 \\
\text{Summer 2011} & -0.05 & 0.05 & 0.17 & 0.30 & 0.08 & -0.26 & -0.37 & -0.54 & -0.61 & 0.60 \\
\text{Autumn 2011} & 0.19 & -0.16 & 0.16 & 0.40 & -0.19 & -0.19 & 0.2 & 0.21 & 0.22 & 0.39 \\
\text{Winter 2011–2012} & 0.34 & -0.23 & -0.17 & -0.05 & -0.70 & -0.75 & -0.59 & -0.83 & -0.83 & -0.83 \\
\text{Spring 2012} & 0.48 & -0.57 & 0.00 & 0.15 & -0.69 & -0.74 & nd & 0.82 & 0.82 & 0.47 \\
\text{Summer 2012} & -0.16 & -0.17 & 0.14 & 0.21 & -0.30 & -0.42 & nd & 0.60 & 0.65 & 0.59 \\
\text{Autumn 2012} & -0.47 & -0.53 & 0.11 & 0.28 & -0.58 & 0.61 & -0.45 & 0.64 & 0.64 & 0.60 \\
\text{Winter 2012–2013} & 0.29 & 0.37 & 0.07 & 0.02 & -0.37 & -0.73 & -0.79 & -0.79 & -0.79 & -0.35 \\
\end{array}
\]

* Negative WTL denotes water level below the soil surface, i.e., positive correlation results from increasing uptake with decreasing WTL. nd means not determined.

There was a highly significant negative correlation between CH$_4$ flux and soil temperature (i.e., higher uptake at higher temperature). The deeper soil temperatures (at 20 and 30 cm depths) showed the best correlation in all seasons. Also, T5 and T2 often correlated with CH$_4$ flux. The highest correlations were observed in springs, winters, and in autumn 2012, and all these correlations were significant in all the chambers with the exception of summer 2011. Also, the correlations of ambient and soil surface temperature were systematically lower than those of the deeper soil temperatures. The correlation with surface soil temperatures was absent or very low in summer 2011, when only the deeper soil temperatures showed a significant correlation.

WTL and CH$_4$ flux correlated positively in both summers (i.e., the deeper the WTL, the higher the uptake), while a negative correlation (i.e., the deeper the WTL, the higher the emission) was found for the winters. The correlation between the flux and PAR was always low ($|r| < 0.2$) or absent.

A significant positive correlation between WS and the flux was found in both summers and autumns and in spring 2012, meaning that the estimated sink decreased when WS increased. This correlation was especially clear in the time periods when there was diurnal variation in the flux, but it was nonexistent when no diurnal cycle was observed, which is consistent with the result reported above. However, even though the correlation was high in autumn 2011, no statistically significant diurnal variation in the flux was observed at that time (data not shown). Of all the inspected quantities, WS was the best explanatory factor of the diurnal cycle of CH$_4$ flux; this was followed by temperature quantities.
4 Discussion

4.1 Chamber closure time and flux calculation method

In this study, we found that using the linear regression method in flux calculations resulted in 20–50 % lower flux estimates for most of the time, in comparison to fluxes calculated with an exponential fit. However, in winter and early spring, i.e., at the time of low absolute CH4 fluxes (< 2.5 μg CH4 m–2 h–1), linear regression gave more reliable results with clearly lower hour-to-hour noise in the fluxes. The uncertainty associated with exponential regression with low fluxes was caused by the decreased signal-to-noise ratio in the concentration data, leading to more or less arbitrary values of the concentration change estimated at t = 0. This explains why the difference between the fluxes estimated by the linear and exponential models was highest during winter and spring.

The use of exponential regression can be considered especially justified during summer and autumn when the concentration development inside the chamber is strongly nonlinear. During these periods, linear regression gave significantly lower flux estimates with an average difference of 12.8 ± 1.8 %. In winter and spring, however, the average seasonal difference was as high as 44 %. Concerning the annual balance, it should be noted that this difference has a greater influence during summer and autumn when the fluxes are up to 2 orders of magnitude larger. The mean difference during the whole measurement period (27.5 ± 0.3 %) is in agreement with many previous studies; for example, Anthony et al. (1995) and Pedersen et al. (2010) reported a 35 and 34 % decrease in flux values, respectively, when using linear regression instead of exponential regression. Kutzbach et al. (2007) reported that underestimations for multiple sites varied mostly within 20–60 %, while Phlaitie et al. (2013) found an average underestimation of 30 % under laboratory conditions at different flux levels. However, the above-mentioned studies focused on either CH4 and CO2 emissions or CO2 uptake and did not include measurements related to a small soil sink of CH4. Since the exponential fit results in larger uptake estimates, which we consider to better represent the correct flux at the time of the chamber closure, we agree with the previous studies that recommend the use of exponential regression in flux calculation. However, due to the considerable noise generated for small fluxes, we recommend that, in future studies employing temporally high-resolution data, the fluxes should be calculated initially with both methods to determine the threshold for “low” fluxes. As this threshold value is dependent on the measurement system and method, it cannot be generalized.

We found that increasing the fitting period up to 15 min from the beginning of the concentration time series systematically decreased the flux estimated by the linear regression model, but only after this period exceeded 140 s. In contrast, the flux estimated by the exponential regression decreased until the fitting window covered 190 s after which the estimated flux was stabilized. To disentangle whether these anomalous patterns in the beginning of both curves were caused by an initial disturbance to the measurement or they represented a real phenomenon affecting the flux, we removed 2 min of data from the beginning of concentration time series. This removed the plateau with large variation from the linear fit results but did not change the shape shown for the exponential fit in Fig. 4b (results not shown). This suggests that the concentration data during the first minute of measurement are perturbed by chamber closure and that...
the linear fit is more sensitive to this than the exponential fit, when using a short fitting period. However, the exponential fit seems to overestimate the flux if the fit is limited to a short time window, but this overestimation is not related to a possible disturbance in the beginning of the data.

Because the concentration development is nonlinear, it is not only the length of the fitting period but also the start of this period, which is important for the flux estimate. The more data are removed from the beginning, the smaller the estimated CH$_4$ uptake becomes. Ueyama et al. (2015) noticed that the flux estimates by linear and exponential regressions both decreased with increasing closure time. Their slopes of linear regression (14 ± 2 % over 5 min) decreased faster than those in our study (16.2 ± 0.2 % over 13 min), suggesting that their concentration data were more nonlinear. The difference could be partly explained by the fact that Ueyama et al. (2015) used smaller chambers and measured higher fluxes, which both result in a faster decrease of the vertical concentration gradient between the soil and the chamber headspace.

### 4.2 CH$_4$ exchange dynamics in a peatland forest

#### 4.2.1 Annual balances

The measurement site (excluding the ditches) was a small annual CH$_4$ sink (varying from 172 ± 31 to 267 ± 55 mg CH$_4$ m$^{-2}$ yr$^{-1}$, ± standard error of the mean) over the 2-year measurement period. While we do not have measurements prior to drainage, it is possible to roughly estimate the pre-drainage fluxes from measurements conducted at similar sites. As Lettosuo was originally a herb-rich, tall sedge birch-pine fen, it can be considered similar to the site reported by Nylänen et al. (1998). That site has high CH$_4$ emissions of 25 g CH$_4$ m$^{-2}$ yr$^{-1}$, so it is obvious that the drainage of Lettosuo has turned the peat soil from a CH$_4$ source to a small CH$_4$ sink. However, it should be noted that our calculations of annual and daily CH$_4$ exchange do not include the emissions from the ditches, which have been found to be highly variable: from 0 to 600 mg CH$_4$ m$^{-2}$ day$^{-1}$ (e.g., Minkkinen et al., 1997; Minkkinen and Laine, 2006; Luan and Wu, 2015). At Lettosuo, where the ditches cover 2–3 % of the area, we estimated, based on 41 manual chamber flux measurements from six points made in the latter half of 2011, that CH$_4$ emissions (per m$^2$ ditch) from the ditches averaged 22 ±38, ±SD g CH$_4$ m$^{-2}$ yr$^{-1}$ (unpublished data). Simple upscaling suggests that, when ditches are accounted for, Lettosuo is a small annual source of CH$_4$ to the atmosphere, although the high uncertainties associated with this calculation should be noted.

Previous studies of drained peatland forests in Finland have reported uptake rates varying between 10 and 970 mg CH$_4$ m$^{-2}$ yr$^{-1}$ (Alm et al., 1999; Minkkinen et al., 2007; Ojanen et al., 2010; Lohila et al., 2011). The average annual uptake in boreal upland forests typically varies from about 100 to 500 mg CH$_4$ m$^{-2}$ yr$^{-1}$ (Smith et al., 2000; Dutaur and Verchot, 2007; Lohila et al., 2016), but also annual net emissions from upland forests have been reported (e.g., Sundqvist et al., 2015; Lohila et al., 2016). Thus, the net annual CH$_4$ exchange measured at Lettosuo (excluding the ditches) was well within the typical range of the average CH$_4$ sinks reported for boreal upland forests.

#### 4.2.2 Emission peaks

A few larger CH$_4$ emission peaks were observed during and after heavy rainfall events in summer, but not all chambers responded to the same rainfall events. The largest number and magnitude of these short bursts of CH$_4$ were recorded with the Sphagnum-dominated chamber (6), which was expected as Sphagnum mosses favor wet spots. These rainfall events turned the soil from a sink to a source (up to 200 µg CH$_4$ m$^{-2}$ h$^{-1}$ in chamber 6) for a short period ranging from a few hours to few days. This was possibly due to increased water saturation and decreased air-filled pore space leading to reduction in oxygen diffusion, which could promote methanogenic activity and suppress methane oxidation.

Similarly to our results, Nylänen et al. (1995) observed that a drained peatland soil can switch to a CH$_4$ source (up to 200 µg CH$_4$ m$^{-2}$ h$^{-1}$) during water saturation event. However, the emission peaks at Lettosuo were relatively small when compared to upland mineral soil forest sites, where emissions of up to 3.7 mg CH$_4$ m$^{-2}$ h$^{-1}$ have been observed in wet conditions (e.g., Savage et al., 1997; Lohila et al., 2016). Similarly, the maximum hourly uptake in summer at Lettosuo was rather similar to the fluxes reported for the above-mentioned upland forests (from −50 to −120 µg CH$_4$ m$^{-2}$ h$^{-1}$ at Lettosuo vs. −40 to −80 µg CH$_4$ m$^{-2}$ h$^{-1}$ at upland forests).

#### 4.2.3 Spatial variation

There were relatively large differences in the annual net CH$_4$ exchange rates between the chambers (from −104 mg CH$_4$ m$^{-2}$ yr$^{-1}$ in chamber 3 to −505 mg CH$_4$ m$^{-2}$ yr$^{-1}$ in chamber 5), even though all the chambers were located within a maximum distance of about 15 m from each other. The difference in the soil surface temperature between the chambers was usually less than 2 K, which indicates that the soil temperature was not the main factor determining the observed spatial variation in fluxes. Since we do not have WTL data for each of the chambers separately, we cannot quantitatively evaluate its role in the spatial variation of fluxes. It is unlikely that the variation in the WTL solely could explain the difference: even though chambers 4 and 5 were located at the same distance from a ditch, and probably had the deepest WTL, chamber 5 showed the highest uptake, while chamber 4 was one of the smallest sinks. Hence, although WTL is likely to explain part of the spatial variation in the CH$_4$ fluxes, there are potentially
many other factors, such as the vegetation composition and small-scale soil properties. The smallest sink was observed in chambers 1, 3, and 4, which were characterized with the lowest (3) and highest (1, 4) vascular green area (VGA\text{max}) values. Thus it seems that it was not the amount of ground vegetation which affected the sink, but a more relevant factor could have been the coverage of mosses vs. that of vascular plants within the collar, especially that of the forest mosses *Pleurozium schreberi* and *Dicranum polysetum*, which were particularly abundant in the highly oxidizing chambers 2 and 5. Due to the small number of chambers, however, the relationship between the forest floor vegetation and the CH\text{4} exchange may be coincidental and can only be speculated.

### 4.2.4 Diurnal variation

All the chambers recorded diurnal variation in CH\text{4} flux at some time during the study period with most of the variation observed during late spring and early summer. Typically, CH\text{4} uptake was at its highest during the night and decreased towards the afternoon. The diurnal variation was more common and occurred more often in all chambers in 2012, while in 2011 it occurred mostly in chamber 2. This variation usually ceased or was at least greatly diminished during and after rainfall events, but usually it appeared again after a couple of days. WTL as such, however, did not have an impact on the diurnal cycle, which suggests that the conditions in the soil surface were much more important for this phenomenon.

Although the diurnal variation seemingly followed the patterns in the air and soil surface temperatures, it was best explained by the WS measured above the canopy (below canopy WS is not available). To further study this relationship, we tested the correlation between the parameter \( c_{\text{exp}} \) (Eq. 2) and WS. \( c_{\text{exp}} \) represents the curvature in the exponential fit, being negative whenever the concentration increase during a chamber closure shows a slowing shape. As we only selected negative, i.e., uptake, fluxes here, it follows that a more negative \( c_{\text{exp}} \) indicates a higher curvature in the concentration evolution. Should leaking be responsible for the smaller CH\text{4} uptake during daytime, as the observed relationship between WS and CH\text{4} exchange implied, it would be logical to find higher curvatures with higher WS. Such a relationship was recently found by Pirk et al. (2016) for CH\text{4} emission chamber flux data from pristine peatlands. However, we did not observe such a relationship in our CH\text{4} uptake data. For example, in chamber 2, in which the diurnal cycle was most explicit, an increasing \( c_{\text{exp}} \) was determined for most of the periods studied. There were only a few chambers and periods when \( c_{\text{exp}} \) decreased with increasing WS. Thus we must conclude that the diurnal variation in our data is related to the technical operation of the chamber rather than environmental conditions. Nevertheless, as the temperature and WS correlated strongly, it is possible that some of the observed pattern was due to some microbial or environmental factor.

We hypothesize that, rather than chamber leaking, the main underlying factor for the clear negative correlation between WS and CH\text{4} uptake is related to changes in the soil storage and thus the changes imposed by chamber closure to the concentration gradient within the top soil and the adjacent air layer. Prior to the closure, this gradient is controlled by atmospheric mixing and hence strongly affected by the ambient WS. During a calm night with a cool soil surface, turbulent mixing is strongly suppressed and molecular diffusion gains importance, while windy and sunny conditions result in much smaller vertical gradients due to vigorous turbulence that is also able to perturb the top-soil pore space. After the chamber is closed, the concentration gradient adjusts to the constant mixing generated by a fan. Thus, the change in concentration gradient depends on the mixing conditions that prevail above the target surface just before the chamber is introduced and how these relate to the mixing rate of the chamber headspace air. In the nocturnal case outlined above, mixing is enhanced after the chamber closure, resulting in a higher CH\text{4} uptake in the chamber.

The absence of the diurnal cycle in winter, and during and after the rain in summer, can be explained by the increased soil moisture content, which decreases the air-filled pore space in soil, thus hampering the wind-induced mixing effect at the soil–atmosphere interface and by slowing down the diffusion rate (Pirk et al., 2016).

It should be noted that the situation is different when CH\text{4} exchange is measured above a forest canopy with the EC method. In that case, the measurement does not significantly disturb atmospheric mixing and increased mechanical turbulence potentially enhances vertical gas exchange. Such positive correlation between the downward CH\text{4} flux and WS, with higher sinks during the daytime, has been reported by Wang et al. (2013). This is consistent with the results of our fan-speed test, described in Koskinen et al. (2014) who measured CO\text{2} respiration by the same chamber system. The CH\text{4} flux data from the same test showed a higher CH\text{4} uptake with higher fan speed (data not shown).

A wind-induced diurnal cycle suggests that the current chamber setup potentially leads to an over- or underestimation of the actual uptake rate during lower or higher fan-induced mixing, respectively, as compared to ambient mixing by wind. The chamber construction could be improved by making the fan speed vary as a function of the ambient wind speed, so as to mimic the variations in atmospheric mixing. However, we can expect that the systematic bias resulting from the wind response is minimized when employing automated sampling that facilitates continuous measurements. Our results imply that sporadic sampling with manual chambers, which is typically limited to the daytime, would have resulted in lower uptake estimates for this site than the extensive data collected with our automatic system.
4.3 Driving factors for the CH$_4$ efflux

The seasonal CH$_4$ fluxes correlated best with soil temperatures at the depths of 20 cm and 30 cm, but significant correlations occurred also with soil temperatures at other depths for most of the study period. The correlations with air and soil surface temperatures were lower. The correlations were always negative, indicating that higher temperatures promoted the soil CH$_4$ uptake. This observation could be attributed to increased consumption of CH$_4$ by methanotrophs in higher temperatures that enhance methanotrophic activity (e.g., Mohanty et al., 2007). However, it is likely that in addition to – or even instead of – the increased methanotrophic activity, there are other reasons behind this relationship. The covariation of temperature with other variables, such as ground water level and phenology, all typically peaking in July–August, may lead to spurious correlation between temperature and CH$_4$ flux. Indeed, the flux was also correlated with WTL, the correlation being significantly positive (higher uptake with lower WTL) in spring, summer, and autumn, but negative in winter. At our site, the soil layers most favorable for methane production and oxidation are located at clearly different depths in the soil, the first being found below the water table and the latter much closer to the soil surface (A. Putkinen, unpublished data). Both of these have distinct temperature and moisture responses, which are practically impossible to disentangle by examining the net CH$_4$ flux observed at the surface.

In addition to the correlations found in the hourly data, we found evidence that lowering WTL increases the daily CH$_4$ uptake in the latter part of summer, when WTL $<-40$ cm. In the beginning of the summer, the daily fluxes were better explained by the soil temperature, while after mid-July the WTL overshadowed the temperature as a control of the daily fluxes.

In pristine peatlands, temperature has been shown to correlate positively with the CH$_4$ emission rate (e.g., Mikkeli et al., 1995; Bellisario et al., 1999; Mastepanov et al., 2013). In drained peatland forests, significant correlations between CH$_4$ flux and temperature have been found in CH$_4$-emitting ecosystems (Nykänen et al., 1998; Minkkinen and Laine, 2006), although the direction of the correlation has been found to differ between fens and bogs. In contrast, no significant correlations with temperature were found in peatland forests that mainly showed CH$_4$ uptake (Ojanen et al., 2010; Wang et al., 2013). Both CH$_4$ emission and uptake have been found to correlate with WTL (e.g., Bellisario et al., 1999; Ojanen et al., 2010).

In addition to temperature, WTL, and WS, CH$_4$ flux has been found to correlate with tree stand volume (Ojanen et al., 2010; Minkkinen et al., 2007), which is, in contrast, an indirect measure of the WTL. Also, PAR has been observed to correlate positively with CH$_4$ emissions in a *Sphagnum*-dominated mixed mire (Mikkeli et al., 1995). In this study, however, the correlations with PAR were low or absent.

5 Conclusions

In this paper, we have presented a 2-year data set of CH$_4$ exchange measured at the forest floor of a boreal forestry-drained fen. These results show that automated chamber measurements with an accurate online gas analyzer make it possible to observe small CH$_4$ fluxes even during the winter with snow cover. Although the fluxes at our site were relatively low throughout the year, we succeeded in catching the annual cycle in the CH$_4$ uptake. Our results indicate that the forest floor of this peatland site acted as a small annual CH$_4$ sink (mean balance $-219$ mg CH$_4$ m$^{-2}$ yr$^{-1}$), although completing the balance with the emissions from ditches indicates that the site is likely to be a small CH$_4$ source.

In spite of the low flux detection limit of the measurement system, our results indicate that it is necessary to pay attention to the flux calculation methods, and instead of choosing between linear and exponential fits we decided on a combination of both. Even though the fit based on linear regression was observed to typically result in a smaller flux than an exponential fit, its use was justified for low fluxes by showing that it produced more robust estimates when the concentration change during chamber closure was small and thus more affected by measurement noise. In addition, we demonstrated that both the length of the fitting period and the starting time of this window had a significant effect on the flux estimates and thus cannot be selected arbitrarily.

The CH$_4$ uptake, measured with closed chambers, was observed to correlate with wind speed, which caused a corresponding diurnal cycle. However, this was partly attributed to aerodynamic effects due to chamber closure, which are dependent on atmospheric mixing prior to the closure. Thus, the chamber construction could be potentially improved by adjusting the chamber fan speed according to the ambient wind speed. As this variation is partly related to changes in the soil CH$_4$ storage, the error introduced in the annual balance estimated from short-term fluxes can be diminished by continuous measurements fully covering the diurnal cycle. Continuous long-term measurements also facilitate the analysis of the environmental factors that control CH$_4$ exchange. However, in order to understand the biological processes involved in CH$_4$ production and oxidation, i.e., the processes behind the net CH$_4$ flux observed, additional measurements are necessary, focusing on the production and oxidation potentials and the within-soil gas gradients.

Since the considerations of the measurement system performance are site and system specific, we recommend that any future study should address the procedures involved in flux calculation, including the fitting method and the length and delay of the fitting period, based on the analysis presented above. In particular, we recommend using the flux limit method applied in this study, i.e., using linear regression for low fluxes and exponential regression for fluxes above a threshold to be determined.
Data availability. The calculated fluxes and measured headspace temperatures are included in the Supplement. The rest of the meteorological data can be requested from the corresponding author.

The Supplement related to this article is available online at doi:10.5194/bg-14-1947-2017-supplement.

Competing interests. The authors declare that they have no conflict of interest.

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