



Exchange of CO₂ in Arctic tundra: impacts of meteorological variations and biological disturbance

Efrén López-Blanco^{1,2}, Magnus Lund¹, Mathew Williams², Mikkel P. Tamstorf¹, Andreas Westergaard-Nielsen³, Jean-François Exbrayat^{2,4}, Birger U. Hansen³, and Torben R. Christensen^{1,5}

¹Department of Biosciences, Arctic Research Center, Aarhus University, Frederiksborgvej 399, 4000 Roskilde, Denmark

²School of GeoSciences, University of Edinburgh, Edinburgh, EH93FF, UK

³Center for Permafrost (CENPERM), Department of Geosciences and Natural Resource Management, University of Copenhagen, Øster Voldgade 10, 1350 Copenhagen, Denmark

⁴National Centre for Earth Observation, University of Edinburgh, Edinburgh, EH93FF, UK

⁵Department of Physical Geography and Ecosystem Science, Lund University, Sölvegatan 12, 223 62 Lund, Sweden

Correspondence to: Efrén López-Blanco (elb@bios.au.dk)

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Abstract. An improvement in our process-based understanding of carbon (C) exchange in the Arctic and its climate sensitivity is critically needed for understanding the response of tundra ecosystems to a changing climate. In this context, we analysed the net ecosystem exchange (NEE) of CO₂ in West Greenland tundra (64° N) across eight snow-free periods in 8 consecutive years, and characterized the key processes of net ecosystem exchange and its two main modulating components: gross primary production (GPP) and ecosystem respiration (R_{eco}). Overall, the ecosystem acted as a consistent sink of CO₂, accumulating -30 g C m^{-2} on average (range of -17 to -41 g C m^{-2}) during the years 2008–2015, except 2011 (source of 41 g C m^{-2}), which was associated with a major pest outbreak. The results do not reveal a marked meteorological effect on the net CO₂ uptake despite the high interannual variability in the timing of snowmelt and the start and duration of the growing season. The ranges in annual GPP (-182 to -316 g C m^{-2}) and R_{eco} (144 to 279 g C m^{-2}) were > 5 fold larger than the range in NEE. Gross fluxes were also more variable (coefficients of variation are 3.6 and 4.1 % respectively) than for NEE (0.7 %). GPP and R_{eco} were sensitive to insolation and temperature, and there was a tendency towards larger GPP and R_{eco} during warmer and wetter years. The relative lack of sensitivity of NEE to meteorology was a result of the correlated response of GPP and R_{eco} . During the snow-free season of the anomalous year of 2011, a biological disturbance related to a larvae outbreak reduced

GPP more strongly than R_{eco} . With continued warming temperatures and longer growing seasons, tundra systems will increase rates of C cycling. However, shifts in sink strength will likely be triggered by factors such as biological disturbances, events that will challenge our forecasting of C states.

1 Introduction

Quantifying the climate sensitivity of carbon (C) stocks of the terrestrial biosphere is a major challenge for Earth system science (Williams et al., 2005). In the Arctic, organic soil C storage has the potential for very large C releases following thaw (Koven et al., 2011) that could create a positive feedback on climate change and accelerate the rate of global warming. Recent reviews have estimated the Arctic terrestrial C pool to be 1400–1850 Pg C, more than twice the size of the atmospheric C pool (Hugelius et al., 2014; McGuire et al., 2009; Tarnocai et al., 2009) and approximately 50 % of the global soil organic C pool (AMAP, 2011; McGuire et al., 2009). Further, Arctic ecosystems have experienced an intensified warming tendency, reaching almost twice the global average (ACIA, 2005; AMAP, 2011; Callaghan et al., 2012c; Serreze and Barry, 2011). The projected Arctic warming is also expected to be more pronounced in coming years (AMAP, 2011; Callaghan et al., 2012a; Christensen et al., 2007; Grøndahl et al., 2008; Meltofte et al., 2008) and tem-

perature, precipitation and growing season length will likely increase in the Arctic (ACIA, 2005; Christensen et al., 2007, 2004; IPCC, 2007). Given this situation, an improvement in our process-based understanding of CO₂ exchanges in the Arctic and their climate sensitivity is critical (McGuire et al., 2009).

Measuring the interannual C exchange variability in the Arctic tundra is challenging due to extreme conditions and the patchy nature of the landscape linked to microtopography. Different eco-types are linked to different C exchange rates (Bubier et al., 2003). Synthesis studies have found a significant spatial variability in NEE (Lafleur et al., 2012; Mbufong et al., 2014) between different tundra sites (Lindroth et al., 2007; Lund et al., 2010) and also large temporal variability within sites (Aurela et al., 2004, 2007; Christensen et al., 2012; Grøndahl et al., 2008; Lafleur et al., 2012). Minor variations in the key process of photosynthesis (gross primary production, GPP) and ecosystem respiration (R_{eco}) may promote important changes in the sign and magnitude of the C balance (Arndal et al., 2009; Elberling et al., 2008; IPCC, 2007; Lund et al., 2010; Tagesson et al., 2012; Williams et al., 2000). With continued warming temperature and longer growing seasons, tundra systems will likely have enhanced GPP and R_{eco} rates, but long-term data with which to investigate and quantify these responses are rare. Further, the effects on net CO₂ sequestration are not known, and may be altered by long-term processes such as vegetation shifts and short-term disturbances like insect pest outbreaks, complicating the prognostic forecast of upcoming C states (Callaghan et al., 2012b; McGuire et al., 2012). Consequently, there is a need to understand how the C cycle behaves over timescales from days to years and the links to environmental drivers. There is a lack of reference sites in the Arctic from which full measurement-based data are available, documenting carbon fluxes at the terrestrial catchment scales. Here we investigate the functional responses of C exchange to environmental characteristics across eight snow-free periods in 8 consecutive years in West Greenland.

In recent decades, eddy covariance has become a fundamental method for carbon flux measurements at the landscape scale (Lasslop et al., 2012; Lund et al., 2012; Reichstein et al., 2005). Eddy covariance measurements of land-atmosphere fluxes or net ecosystem exchange (NEE), of CO₂ can be gap-filled and subsequently separated into the modulating components of GPP and R_{eco} using flux partitioning algorithms (Reichstein et al., 2005). These techniques are critical for providing a better understanding of the C uptake vs. C release behaviour (Lund et al., 2010), but they also allow for an examination of the environmental effects on ecological processes (Hanis et al., 2015). However, large gaps in the measured fluxes may introduce significant uncertainties in the C budget estimations. Moreover, GPP and R_{eco} estimates can be calculated in different ways. Some algorithms fit an instantaneous temperature–respiration curve to night-time data to calculate R_{eco} and estimate GPP (Lasslop

et al., 2012; Reichstein et al., 2005); others calculate R_{eco} from a light-response curve (Gilmanov et al., 2003; Lindroth et al., 2007; Lund et al., 2012; Mbufong et al., 2014; Runkle et al., 2013). Unfortunately, different interpretations of the flux gap filling and partitioning lead to different estimates of NEE, GPP and R_{eco} as well as undefined uncertainties.

The main objectives of this paper are (1) to explore the uncertainties in NEE gap filling and partitioning obtained from different approaches, (2) to determine how C uptake and C storage respond to the meteorological variability, and (3) to identify how the environmental forcing affects not only the interannual variability, but also the hourly, daily, weekly and monthly variability of NEE, GPP and R_{eco} . The intention of this paper is to elaborate on the information gathered in an existing catchment area under an extensive cross-disciplinary ecological monitoring programme in low Arctic West Greenland, established under the auspices of the Greenland Ecosystem Monitoring (GEM) (<http://www.g-e-m.dk>). Using a long-term (8-year) data set to explore uncertainties in NEE gap-filling and partitioning methods and to characterize the interannual variability of C exchange in relation to driving factors can provide our understanding of land-atmosphere CO₂ exchange in Arctic regions with a novel input. Our overarching hypothesis was that both GPP and R_{eco} would respond positively to warmer and longer growing seasons. However NEE response to warming would be more complex and variable (positive or negative) depending on subtle balances between plant and microbial climate sensitivity.

2 Materials and methods

2.1 Site description

Field measurements were conducted in the low Arctic Kobbefjord drainage basin in south-western Greenland (64°07' N; 51°21' W) (Fig. 1a). The study area is located ~20 km SE of Nuuk, the Greenlandic capital. Kobbefjord has been subject to extensive environmental research activities (the Nuuk Ecological Research Operations) since 2007 (<http://www.nuuk-basic.dk>). The lowland site is located 500 m from the south-eastern shore of the bottom of Kangerluarsunnguaq Fjord (Kobbefjord), and 500 m from the western shore of the 0.7 km² lake called “Badesø” (Fig. 1b). Three glaciated mountains, all above 1000 m a.s.l., surround the site. The landscape consists of a fen area surrounded by heath, copse and bedrock. The current fen vegetation is dominated by *Scirpus cespitosus*, whereas the surroundings are dominated by heath species such as *Empetrum nigrum*, *Vaccinium uliginosum*, *Salix glauca* and copse species such as *S. glauca* and *Eriophorum angustifolium* (Bay et al., 2008). Kobbefjord belongs to the “arctic shrub tundra” (bioclimate zone E) according to The Circumpolar Arctic Vegetation Map (CAVM Team, 2003; Walker et al., 2005). This map is

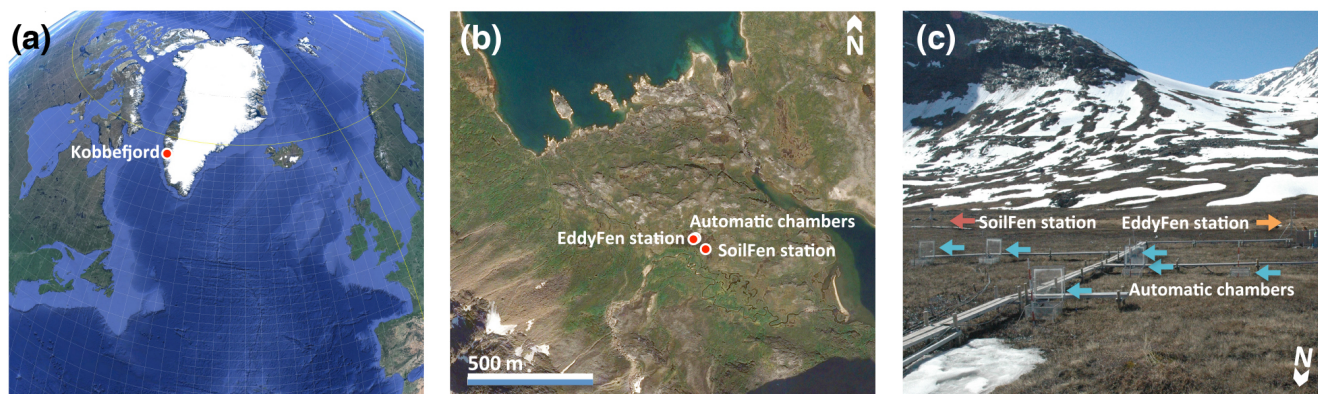


Figure 1. (a) Location of Kobbefjord in Greenland, 64°07' N; 51°21' W (source: Google Earth Pro). (b) Location of EddyFen station, automatic chambers and SoilFen station in Kobbefjord (source: Google Earth Pro, 16 July 2013). (c) Eddy covariance (orange arrow) from EddyFen station, six automatic chambers (light blue arrows) and SoilFen station (pale red arrow) (photo by Efrén López Blanco, 27 June 2015).

based on the summer warmth index (SWI), which is the sum of the monthly mean temperature above 0 °C from May to September and the southernmost bioclimatic zone E has limits of 20–35. In 2010 and 2012, the weather conditions led the area to experience temperatures from warmer climatic zones (SWI ca. 36 and 35 respectively). For the 1961–1990 period, the mean annual air temperature was -1.4 °C and the annual precipitation was 750 mm (Cappelen, 2013). The sunlight hours between May and September range from 14 to 21 h. Outcalt's frost number (Nelson and Outcalt, 1987) indicates that discontinuous permafrost should be present, although no permafrost has been found. Nonetheless, thin lenses of ice may remain until late summer.

2.2 Measurements

We have used eddy covariance (EC) data on NEE, measured during the snow-free period from 2008 to 2015. Measurements typically started around the end of the snowmelt (ca. May–June) and extended until the freeze-in period (between September and October). Once the snow melts, the growing season (i.e. the part of the year when the weather conditions allow plant growth) has been reported as the most relevant period defining both spatial (Lund et al., 2010; Mbufong et al., 2014) and temporal (Aurela et al., 2004; Groendahl et al., 2007; Lund et al., 2012) CO₂ variability. The EC measurements were conducted at the EddyFen station (Fig. 1b and c), located in a wet lowland, 40 m a.s.l. The EC tower is equipped with a closed-path infrared CO₂ and H₂O gas analyser LI-7000 (LI-COR Inc, USA) and a 3-D sonic anemometer Gill R3-50 (Gill Instruments Ltd, UK). The anemometer was installed at a height of 2.2 m, while the air intake was attached 2.0 m above terrain on the steel stand. Adjacent to the EddyFen station, an independent system (Fig. 1b and c) measures round-the-clock net CO₂ fluxes using an automatic chamber (AC) method based on Goulden and Crill (1997).

The transparent chambers, each covering a known surface area of 60 cm by 60 cm, with a height of 30 cm, can be opened and closed by the computer in succession for 10 min every hour. When the chamber closes, a CO₂ analyser (SBA-4, PP Systems, UK) monitors both the CO₂ concentration by a close loop of tubing (further information about the set up can be found in Mastepanov et al. (2013)). Nearly 20 m from the EddyFen station, the automated SoilFen (Fig. 1b and c) station provides environmental variables such as air and surface temperature (Vaisala HMP45C), soil temperature at different depths (Campbell scientific 10ST) and relative humidity (Vaisala HMP45C). Two kilometres from these stations, an automatic weather station provides complementary ancillary data such as short- and long-wave radiation (with a CNR1 instrument), photosynthetic active radiation (with a Kipp & Zonen PAR Lite instrument), precipitation (using an Ott Pluvio instrument) and snow depth (with a Campbell Scientific SR 50). The water table depth data were monitored using a piezometer located next to each of the six autochambers. Finally, a robust daily estimate of the timing of snowmelt was analysed at a pixel level from a time-lapse camera (HP e427) located at 500 m a.s.l. (Westergaard-Nielsen et al., 2013).

2.3 Data handling

2.3.1 Data collection and pre-processing

Data collection from the EddyFen station was performed using Edisol software (Moncrieff et al., 1997). Raw data files were processed using EdiRe software (version 1.5.0.32, R. Clement, University of Edinburgh) calculating the CO₂ fluxes on a half-hourly basis. The flux processing integrated despiking (Højstrup, 1993), 2-D rotation, time lag removal by covariance optimization, block averaging, frequency response correction (Moore, 1986) and Webb–Pearman–

Leuning correction (Webb et al., 1980). For more information, see Westergaard-Nielsen et al. (2013). Ancillary data (air temperature, soil temperature, incoming short-wave radiation, relative humidity, PAR and precipitation) were temporally resampled using R (R Development Core Team, 2015). Time-series-related packages such as *zoo* (Zeileis and Grothendieck, 2005), *xts* (Ryan and Ulrich, 2014) and *lubridate* (Grolemund and Wickham, 2011) were used to get the ancillary data aligned with the flux data on a half-hourly basis.

2.3.2 Generating robust and complete flux time series

Before the CO₂ flux time series were analysed, we applied three different processing techniques (u^* filtering, gap filling and partitioning) to (1) filter the NEE data for quality, (2) fill the NEE gaps and (3) separate NEE into GPP and R_{eco} . The identification of periods with insufficient turbulence conditions (indicated by low friction velocity u^*) is important for avoiding biases and uncertainties in EC fluxes. To control the data quality, the u^* thresholds were bootstrapped by identifying conditions with inadequate wind turbulence according to the method described in (Papale et al., 2006). We subsetted the data to similar environmental conditions, aside from friction velocity: 8 years and 7 temperature classes. Within each year/temperature subset the u^* threshold (5, 50 and 95 % of bootstrap) was estimated at 1000 samples per year. We used the subsequent gap filling and partitioning based on these different subsets to propagate the uncertainty of u^* threshold estimation across NEE, GPP and R_{eco} .

Our gap-filling method was similar to Falge et al. (2001), using the marginal distribution sampling (MDS) algorithm, re-adapted from Reichstein et al. (2005) in REdDyProc (Reichstein et al., 2016). MDS takes into account similar meteorological data available with different window sizes (Moffat et al., 2007). Parallel to this approach, we also gap-filled the original EC NEE data with an independent AC NEE data set (2010–2013). AC data were collected simultaneously with EC data, and so we can use them as a cross check. The EC NEE was predicted from AC NEE based on linear regression models. The subsequent product was gap-filled using the MDS algorithm (REddyProc).

We separated NEE into its two main components (GPP and R_{eco}) using two approaches: (1) the REdDyProc partitioning tool (Reichstein and Moffat, 2014) and (2) a light-response curve (LRC) approach (Lindroth et al., 2007; Lund et al., 2012). A brief description of each flux partitioning method is provided in the Supplement (Eq. S1). After the flux partitioning comparison, we used REdDyProc-based GPP and R_{eco} estimates on further analyses.

2.3.3 Flux uncertainties

In order to estimate the NEE gap-filling uncertainty, we assessed three different sources of uncertainty. First, we addressed the 95 % confidence interval of the EC prediction based on AC data. Second, we inferred the random uncertainty of filled half-hourly values from the spread of variables with otherwise very similar environmental conditions. REdDyProc uses the gap filling to also estimate an observation uncertainty for the measured NEE, by temporarily introducing artificial gaps (T. Wutzler and M. Migliavacca (BGC-Jena), personal communication). Finally, we assessed the effect of uncertainty in the estimate of the u^* threshold. In the u^* -NEE relationship we want to exclude the probably false low fluxes (absolute NEE values) at low u^* . When choosing a lower u^* threshold, the associated lower flux will contribute to the gap filling and the annual sums. Therefore, there is a tendency of a lower absolute NEE associated with lower u^* . The difference between the 5 and 95 % of bootstrap provides a means of the uncertainties based on the u^* filters. We summed and propagated all these sources of uncertainties over time. The GPP and R_{eco} uncertainties include the bias from the one-to-one flux comparison obtained from each model. The micrometeorological sign convection used in this study present uptake fluxes (GPP) as negative, while the released fluxes (R_{eco}) are shown as positive.

2.4 Identifying environmental forcing

Snow- and phenology-related variables such as the end of the snowmelt period and the start, end and length of the growing season are important components that shape the Arctic CO₂ dynamics. In this study we defined the end of the snowmelt period as the day of year when more than 80 % of the surface of the fen was considered snow free; the threshold was chosen in agreement with suggestions previously reported in Hinkler et al. (2002) and Westergaard-Nielsen et al. (2015). For the start, end and length of the growing season (GS_{start} , GS_{end} , GS_{length}); the GS_{start} and the GS_{end} were defined as the first and last days on which the consecutive 3-day NEE average was negative (i.e. CO₂ uptake) and positive (i.e. CO₂ release) respectively (Aurela et al., 2004), while GS_{length} is the number of days between GS_{start} and GS_{end} .

A random forest machine-learning algorithm (Breiman, 2001; Pedregosa et al., 2011) was utilized in a data-mining exercise to identify how the environmental controls affect the variability of NEE, GPP and R_{eco} . Random forest calculates the relative importance of explanatory variables over the response variables. Here, we use photosynthetic active radiation (PAR), air temperature (T_{air}), precipitation (Prec) and vapour pressure deficit (VPD) to explain the response of C fluxes (NEE, GPP and R_{eco}) to climate variability. Each decision tree in the forest is trained on different random subset of the same training data set. The random forest is a classifier that groups explanatory variables and, in each fi-

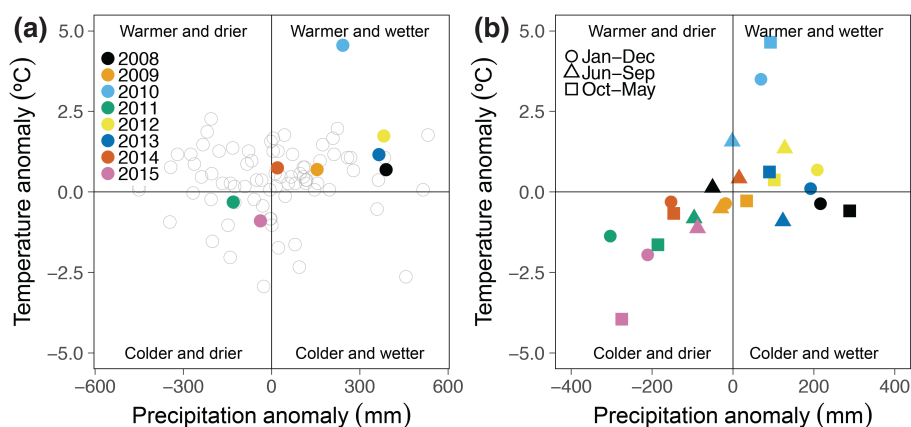


Figure 2. (a) Annual temperature (°C) and precipitation (mm) anomalies of the analysed years (2008–2015) compared to the 1866–2007 time series shown as empty circles (Cappelen, 2016), and (b) within the 2008–2015 period including annual (January to December), warm season (July to September) and cold season (October to May) averages.

nal cluster, a multiple linear regression is built to reproduce fluxes as function of driving factors. This approach has been used to extrapolate maps of biomass (Baccini et al., 2012; Exbrayat and Williams, 2015). This version of random forest sums the relative importance of each variable from 0 up to 100 %, which correspond to the fraction of decision in which a variable is involved to cluster the data. We applied random forests to assess the relative importance of PAR, T_{air} , Prec and VPD at different temporal scales (hourly, daily, weekly and monthly), aggregating them at the timescale indicated and lumping all the years together. (Table S1; Supplement). Moreover, we evaluated the diurnal, seasonal and annual pattern for each explanatory variable (data binned per hour; this is one random forest per hour of the day, day of the year and year respectively). To make sure that these results were not an artefact of the partitioning method that is based on a relationship between hourly R_{eco} and T_{air} , we performed the same analyses using daytime and night-time only hourly NEE as respective proxies for GPP and R_{eco} . Based on these results (Table S2) we concluded that the approach was robust for the Kobbefjord site.

3 Results

3.1 Interannual and seasonal variation of environmental and phenological variables

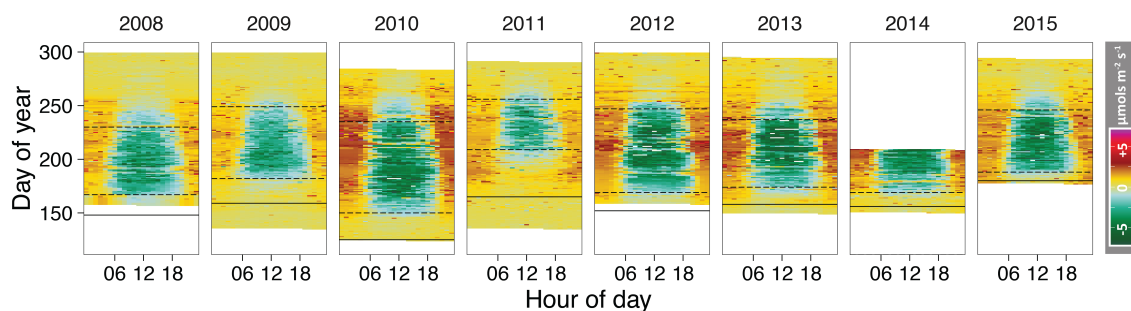
The annual mean temperature documented from Nuuk (-0.5°C) and Kobbefjord (-0.4°C) in the 2008–2015 period were generally warmer compared to the long time series between 1866 and 2007 (Cappelen, 2016; Fig. S1; Supplement), with an annual temperature average of -1.5°C . The 2008–2015 period temperature also exhibited larger variability (coefficients of variation (CV) = 283.3 %) compared to the 1866–2007 period (CV = 79.3 %). The 2008–2015 mean annual temperature measured in Kobbefjord fluctuated be-

tween -1.7°C in 2011 and 3.4°C in 2010. Moreover, the mean annual precipitation documented from the nearby station of Nuuk (885 mm) and the one measured across the 8 year-study in Kobbefjord (862 mm) were both significantly higher than the 1931–2007 mean (689 mm), although less variable (CV = 30.8 and 24.5 %). Overall, 2008, 2009, 2010, 2012, 2013 and 2014 have shown warmer and wetter anomalies while 2011 and 2015 presented colder and drier anomalies compared to the long-term mean (Fig. 2a). Among the 8 study years (Fig. 2b), the temperature and precipitation anomalies in the warm season (June to September) ranged from about -1°C (2011, 2013 and 2015) to $+1.5^{\circ}\text{C}$ (2010) and -96 mm (2011) to about $+125\text{ mm}$ (2012 and 2013). The cold season (October to May) anomalies have shown greater variability compared to the warm season, and 2010, 2012 and 2013 experienced warmer and wetter winters, while 2011 and 2015 were colder and drier.

The end of the snowmelt period and the growing season start and length presented high interannual variability (CVs were 9.5, 9.0 and 19.0 %). Kobbefjord became snow free on DOY 154 (3 June for non-leap years, SD = 15) on average. On average, the site switched from being a source of CO₂ to a sink (GS_{start}) on DOY 175 (24 June, SD = 20), and remained so (GS_{end}) until DOY 241 (29 July, SD = 8.4) (Table 1). The GS_{start} and the $\text{GS}_{\text{length}}$ did not follow a consistent pattern among the analysed years, the growing season timing have fluctuated substantially. The high interannual variability of the GS_{start} correlated with variations in temperature, end of snowmelt period and VPD ($p < 0.05$). The highest variability was observed during 2009–2012. The 2010's $\text{GS}_{\text{length}}$ was nearly twice as long as in 2011. Indeed, GS_{start} in 2011 differs only by 26 days from the GS_{end} in 2010.

Table 1. Summary of the phenology-related variables for the period 2008–2015.

| | 2008 | 2009 | 2010 | 2011 | 2012 | 2013 | 2014 | 2015 |
|-----------------------------------|------|------|------|------|------|------|------|------|
| Maximum snow depth (m) | 0.6 | 1.0 | 0.3 | 1.4 | 1.0 | 0.6 | 1.1 | 1.2 |
| End of snowmelt period (DOY) | 148 | 159 | 125 | 165 | 152 | 158 | 156 | 176 |
| Beginning of growing season (DOY) | 167 | 182 | 150 | 209 | 169 | 174 | 169 | 188 |
| End of growing season (DOY) | 230 | 249 | 235 | 256 | 247 | 237 | – | 246 |
| Length of growing season (DOY) | 63 | 67 | 85 | 47 | 78 | 63 | – | 58 |

**Figure 3.** Time series of gap-filled NEE (2008–2015) based on autochamber data (2010–2013) and the MDS algorithm (from REddyProc). Green represents C uptake while the orange–dark-red denotes C release. The solid lines represent the end of the snowmelt period, while the area within the dashed lines represents the period between the start and the end of the growing season.

3.2 Data processing and quality

The NEE gap filling and subsequent partitioning obtained from different approaches exposed inconsistencies in performance and specific uncertainties in the seasonal C budget calculation. During the eight study snow-free periods, data gaps made up 46.5 % of the record from the EddyFen station due to unfavourable micro-meteorological conditions, instrument failures, maintenance and calibration (Jensen and Christensen, 2014) but also due to the rejection of low-quality flux measurements or too low u^* . In 2014 a major instrument failure forced the station to stop measurements in the middle of the season. In 2010 and 2012 there were two more interruptions in the measurements (data gaps of > 20 days), although the problems could be solved before the end of the season. Such prolonged gaps led to unreliable gap-filled NEE estimates. The REddyProc MDS algorithm tended to fill these large gaps with high peaks of respiration at noontime, coercing C uptake underestimation. For this reason, an independent AC NEE data set (2010–2013) was tested to gap-fill EC data (Figs. 3 and S2). The R^2 obtained from the EC-AC correlations was always > 0.70 (2010: $R^2 = 0.80$, $p < 0.001$; 2011: $R^2 = 0.72$, $p < 0.001$; 2012: $R^2 = 0.80$, $p < 0.001$; 2013: $R^2 = 0.84$, $p < 0.001$). By using AC data, the proportion of missing data was reduced to 28 %, and we found that the random uncertainty from the combination of AC and MDS algorithm decreased by 5 % on average. By using the u^* filtering and the AC data together with EC, there was an increase in ~ 6 % in terms of C sink strength. Moreover, the propagated uncertainty in

NEE never exceeded $\pm 1.8 \text{ g C m}^{-2}$, mainly because the error related to u^* filtering was low. Further, we hypothesized that different flux partitioning approaches would lead to different estimates of GPP and R_{eco} . However, the results suggest a relatively good agreement (Fig. 4). There was a higher degree of agreement with regard to GPP ($R^2 = 0.83$) compared with R_{eco} ($R^2 = 0.30$). LRC tended to estimate 12 and 15 % larger GPP and R_{eco} respectively compared to REddyProc.

3.3 Interannual and seasonal variation of CO₂ ecosystem fluxes

Overall, land–atmosphere CO₂ exchange measured for the snow-free periods of 2008–2015, omitting 2011, acted as a sink of CO₂, taking up -30 g C m^{-2} on average (range -17 to -41 g C m^{-2}) (Fig. 5; Table 2). The cumulative NEE showed a characteristic pattern during the measurement period (Fig. 5), with an initial loss of carbon in early spring right after snowmelt (also observed in Fig. 3), followed by an intense C uptake as assimilation exceeded respiratory losses, triggered by increases in temperature, PAR and vegetation growth. This transition point matched the growing season start, when NEE switched from positive values (a net C source) to negative values (a net C sink). Eventually, the ecosystem turned again into a net C source, defining the growing season end. Even with high interannual variability in terms of the end of snowmelt time and growing season start/length (Table 1), the results do not show a marked meteorological effect on the NEE. The ranges in annual GPP (-182 to -316 g C m^{-2}) and R_{eco} (14–

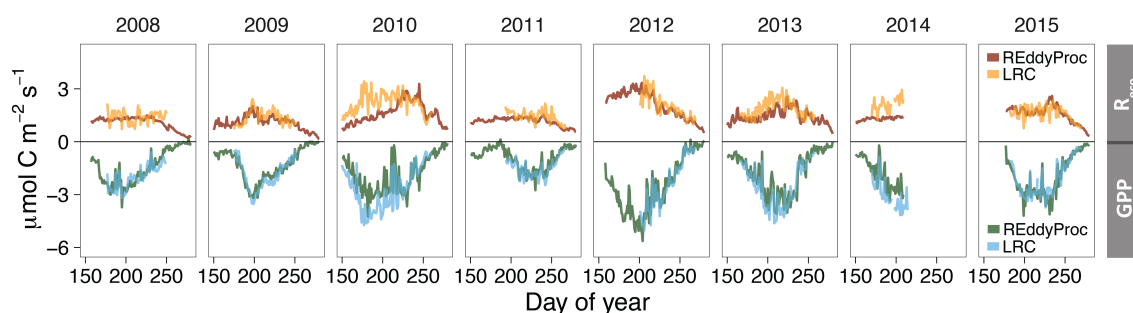


Figure 4. Time series of daily mean GPP (negative fluxes) and R_{eco} (positive fluxes) from 2008 to 2015 calculated by REddyProc (dark green and dark red) and LRC (orange and light blue).

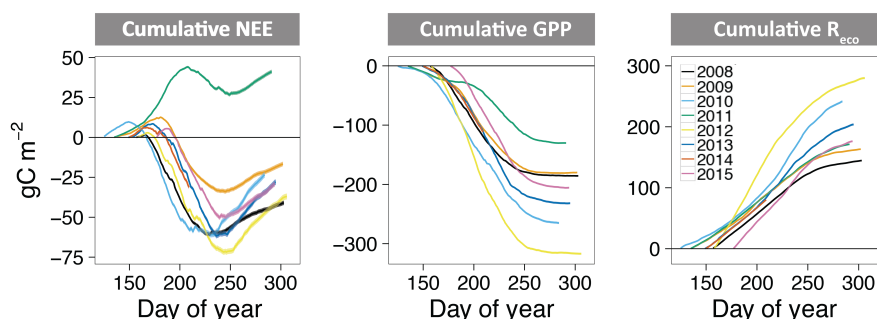


Figure 5. Cumulative NEE, GPP and R_{eco} from 2008 through 2015 including the u^* filtering and random errors.

279 gC m⁻²) (Table 2) were > 5 fold larger and more variable (CVs are 3.6 and 4.1 % respectively) than for NEE (0.7 %). There was a tendency towards larger GPP and R_{eco} during warmer and wetter years (Fig. S3), but there were no warmer and drier years during the study period. The strongest growing season CO₂ uptake occurred in 2012 (NEE = -74.2 gC m⁻²; GS_{length} = 78 days), followed by 2010 (NEE = -70.0 gC m⁻²; GS_{length} = 85 days) (Tables 1 and 2). A lengthening of the growing season did not increase the net carbon uptake in this study. In other words, an earlier end of the snowmelt resulting in a longer growing season length did not lead to a stronger carbon sink.

The anomalous year, 2011, constituted a relatively strong source of CO₂ (41 gC m⁻²) and was associated with a major pest outbreak, which reduced GPP more strongly than R_{eco} . Data on the larvae of the moth *Eurois occulta*, collected from pitfall traps in the surrounding *Salix*- and *Empetrum*-dominated plots, showed a strong peak at the beginning of the 2011 growing season (Lund et al., 2017), coinciding with high NEE and very low GPP (Fig. 4). In 2011 up to 2078 larvae were observed, while in other years only 14 (2008), 82 (2009), 186 (2010), 0 (2012) and 8 (2013) were observed. It is likely that the reduced primary production in the wetland area was a partial response to the *Eurois occulta* outbreak.

The daily aggregated NEE–GPP relationships displayed consistent linear correlation (2008–2015: $R^2 = 0.77$, $p < 0.001$) across the assessed years (Fig. 6a). The linear cor-

relations were weaker in 2010 and 2011. A hysteresis was detected in 2010 (i.e. long growing season with higher R_{eco} in autumn than in spring), while strong C releases were observed in 2011 across June and July. The relation between GPP and R_{eco} , which can be understood as the degree of coupling between inputs and outputs of C and therefore the degree of C sink strength, showed non-linear patterns (Fig. 6b). The curved behaviour is likely because GPP increased more than R_{eco} during early growing season, except for in 2011. Moreover, R_{eco} lagged behind GPP due to (1) the vegetation green-up in the first part of the growing season and (2) the higher respiration rates due to increased biomass in the second part. The years with clearer hysteresis coincide with the years with positive temperature anomalies (i.e. 2010, 2012 and 2013) of the 2008–2015 series. It is worth mentioning the different directions (clockwise or anticlockwise) in the hysteresis observed in these years between June, July and August. The data suggest that the clockwise 2012 hysteresis was due to greater gross C cycling (GPP and R_{eco}) in June and July favoured by warmer conditions, while in 2010 (anticlockwise hysteresis), the higher gross C fluxes were measured in August with warmer and wetter conditions (Fig. S4).

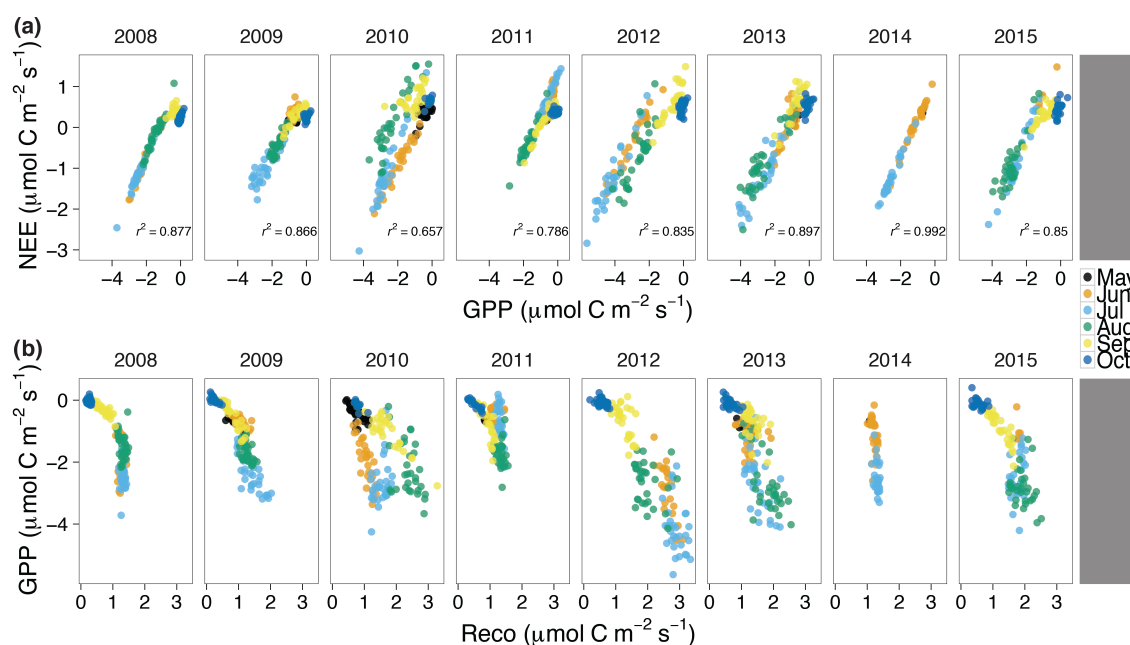
3.4 Environmental forcing

The varied importance of meteorological variables (such as PAR, T_{air} , VPD and precipitation) obtained from random forest at different temporal scales (hourly, daily, weekly and

Table 2. Summary of the measuring periods and the growing season CO₂ fluxes for the period 2008–2015.

| | 2008 | 2009 | 2010 | 2011 | 2012 | 2013 | 2014 | 2015 |
|--|----------------|----------------|----------------|----------------|----------------|----------------|--------------------|----------------|
| First measurement (DOY) | 157 | 135 | 124 | 135 | 158 | 149 | 150 | 177 |
| Last measurement (DOY) | 303 | 304 | 282 | 287 | 305 | 295 | 209* | 294 |
| Missing data (%) | 57.6 | 42.3 | 28.6 | 35.4 | 32.3 | 29.8 | 44.9* | 40.0 |
| NEE in measuring period (g C m ⁻²) | -41.3 ±1.4 | -16.9 ±1.4 | -24.4 ±1.9 | 40.7 ±1.3 | -37.0 ±1.8 | -28.1 ±1.7 | -28.7* ±1.1 | -31.5 ±1.6 |
| NEE in growing season (g C m ⁻²) | -62.3 | -45.9 | -70.0 | -16.2 | -74.2 | -69.7 | -35.3 ^a | -55.8 |
| Maximum daily uptake (DOY) | 195 | 205 | 182 | 230 | 204 | 220 | 192 ^a | 199 |
| Maximum uptake (μmol m ⁻² s ⁻¹) | -2.4 | -1.7 | -3.0 | -1.4 | -2.8 | -2.5 | -1.9 ^a | -2.3 |
| Estimated GPP (g C m ⁻²) | -185.5 ±1.4 | -181.8 ±1.4 | -266.1 ±1.9 | -130.6 ±1.3 | -316.2 ±1.9 | -230.7 ±1.7 | -106.8* ±1.1 | -206.1 ±1.6 |
| Estimated R _{eco} (g C m ⁻²) | 144.2 ±1.3 | 164.9 ±1.3 | 241.6 ±1.8 | 171.3 ±1.2 | 279.2 ±1.8 | 202.6 ±1.7 | 78.1* ±1.1 | 174.6 ±1.5 |

Where applicable: ± sum of the autochamber, random and *u** filtering uncertainties. * incomplete growing season data set.

**Figure 6.** Interannual variability between (a) NEE–GPP and (b) GPP–*R*_{eco} relationships. The data were daily aggregated and coloured per month.

monthly) showed differences in behaviour depending on the time aggregation utilized (Fig. 7). PAR dominated NEE and GPP while *T*_{air} correlated the most with *R*_{eco} in hourly averages, whereas *T*_{air} became increasingly important at longer temporal aggregations for all the fluxes (Fig. 7). VPD and precipitation were not as important as the other variables while the use of water table depth in the analysis was discarded due to its very low impact on CO₂ fluxes. In general, NEE and GPP showed similar distributions of importance, reinforcing the linear relationships found between NEE and GPP (Fig. 6). The standard deviation of the variables' importance (across 1000 decision trees) tended to increase at coarser time aggregations.

Changes of environmental forcing (PAR, *T*_{air} and VPD) across diurnal, seasonal and annual timescales reveal patterns of functional responses to C fluxes. The diurnal cycle analyses on hourly data showed the changes in importance between day- and night-time (Fig. 8). NEE and GPP had two predominant variables (*T*_{air} and PAR) determining the variability at daytime. PAR was important at dawn (06:00 WGST) and dusk (20:00 WGST), while *T*_{air} was more important at other times. This performance indicates a threshold response to PAR, and a more continuous response to temperature. On the other hand, *R*_{eco} was mainly driven by *T*_{air} at both night-time and daytime. VPD and PAR had a negligible impact on *R*_{eco}. The seasonal pattern importance showed

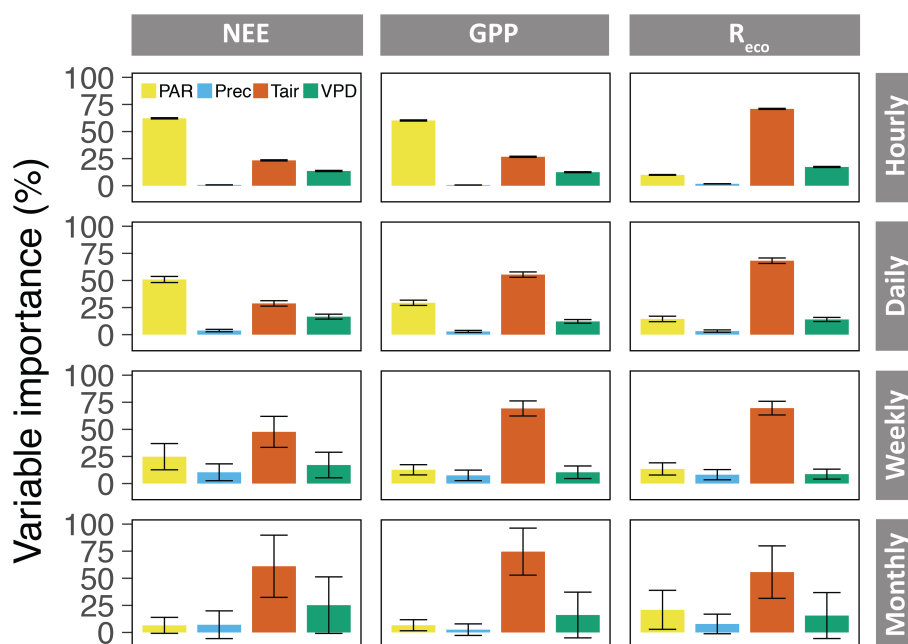


Figure 7. Importance of environmental variables PAR (yellow), T_{air} (orange), Prec (pink) and VPD (green), explaining variability in NEE, GPP and R_{eco} (partitioned by REddyproc) at different temporal aggregations (hourly, daily, weekly and monthly) when all the years were lumped together. Thick bars and error bars represent the mean \pm standard deviation of the importance across 1000 decision trees.

PAR dominating NEE and GPP from early June to early October (Fig. 8), while T_{air} and VPD became more important before and after the snow-free conditions. In terms of CO₂ emission (R_{eco}) the pattern is less clear and noisier, although T_{air} appeared to be the most important variable. Finally, the annual pattern exposes a performance in line with previous results; i.e. PAR dominated NEE and GPP while R_{eco} was more sensitive to variations of T_{air} . Interestingly, the random forest analysis revealed a decrease in PAR's importance in 2011, the same year in which the sharp decrease in C sink strength was exposed.

4 Discussion

4.1 Data processing and quality

The NEE gap filling and subsequent partitioning into GPP and R_{eco} are needed to understand the CO₂ flux responses to the environmental forcing. However, these procedures expose unavoidable uncertainties in the seasonal C budget calculation (Table 2) and partial inconsistencies between approaches (Fig. 4). In this study, we used an MDS gap-filling technique, an enhancement to the standard look-up table. Both methods have shown a good overall performance compared to other procedures such as non-linear techniques or semi-parametric models but slightly inferior to artificial neural network (Moffat et al., 2007). However, the MDS gap filling alone introduced NEE estimates out of range across the

two extensive gaps in 2010 and 2012 (Fig. S2). Quantifying the uncertainty introduced by measurement gaps is complex (Falge et al., 2001; Moffat et al., 2007; Papale et al., 2006). One possibility would be a sensitivity analysis of time series with artificially introduced gaps (Dragomir et al., 2012; Pirk et al., 2017). However, the choice of gap length and position is difficult and would render uncertainty to the uncertainty assessment itself. Instead, we used the EC prediction based on independent autochamber (AC) measurements between 2010 and 2013. The agreement between EC and AC was always $R^2 > 0.72$ and $p < 0.001$, and the 95 % confidence intervals of the predictions were reported together with the resulting uncertainties (Table 2). Although the AC data itself incorporated a new source of uncertainty in the calculations, we consider this method to be less weak than an unreliable gap-filling estimate. We used the AC as platform with which to decrease the gap length and the total random uncertainty (Aurela et al., 2002) before the MDS algorithm was applied. AC was used together with MDS and was never used as an independent gap-filling procedure.

The NEE partitioning obtained from REddyProc and LRC suggests a relatively good agreement in model performance. The one-to-one comparison between different approaches found a better agreement between different approaches for GPP compared to R_{eco} . In this analysis, REddyProc produced smoother R_{eco} estimates compared to the noisier GPP estimates, whereas the LRC results were the other way around. This is mainly because measurement noise goes into GPP for the REddyProc method, and into R_{eco} for the LRC method. RE-

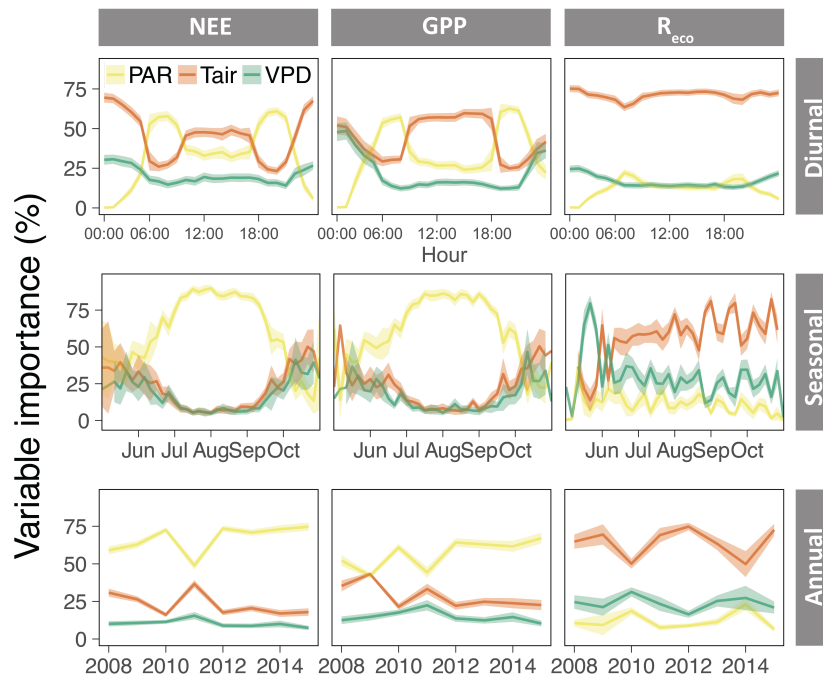


Figure 8. Diurnal, seasonal and annual importance of environmental variables PAR (yellow), T_{air} (orange), and VPD (green), explaining variability in NEE, GPP and R_{eco} . Thick lines and shading represent the mean \pm standard deviation of the importance across 1000 decision trees.

dyProc retrieves positive GPP values, whereas the LRC method results in negative R_{eco} values. Both scenarios are not fully convincing, although it is not straightforward as to how they should be treated. Removing all positive GPP/negative R_{eco} values would risk removing only one side of the extremes. Besides night-time-based (REddyProc) and daytime-based (LRC) partitioning approaches, several implementations have been proposed to improve the algorithm's performance. Lasslop et al. (2010) has modified the hyperbolic LRC to account for the temperature sensitivity of respiration and the VPD limitation of photosynthesis. Further, Runkle et al. (2013) proposed a time-sensitive multi-bulk flux-partitioning model, where the NEE time series was analysed in 1-week increments as the combination of a temperature-dependent R_{eco} flux and a PAR-dependent flux (GPP). However, it remains uncertain as to under which circumstances each partitioning approach is more appropriate, especially in the boundaries between low- and high-Arctic due to the lack of dark night during polar days (when light is not a limiting factor for plant growth). Since there are few methods with an unclear precision, an evaluation study on the effect of using different partitioning approaches along latitudinal gradients would be very beneficial to assessing the suitability for each method.

4.2 Interannual and seasonal variation of CO₂ ecosystem fluxes

The balance between the two major gross fluxes in terrestrial ecosystems, photosynthetic inputs (GPP) and respiration outputs (R_{eco}), displayed larger temporal variability than did NEE. These results suggest that both GPP and R_{eco} were strongly coupled and sensitive to meteorological conditions such as insolation and temperature (Figs. 7 and 8). Interestingly, the tendency to warmer and wetter conditions led to greater rates of C cycling associated with larger GPP and R_{eco} (Fig. S3). This result does not entirely coincide with Peichl et al. (2014), even though they performed a similar analysis for a Swedish boreal fen. This finding points towards the complexity in the response of wetland ecosystems towards changing environmental conditions. The response is dependent on many things, such as hydrological settings, and these differ between sites. In this study, larger rates of C uptake (GPP) were linked to larger rates of C release (R_{eco}), with the exception of the anomalous year 2011. The relative insensitivity of NEE to meteorological conditions during the snow-free period could be the result of the correlated response of ranked cumulative GPP and R_{eco} (Fig. 5) (Richardson et al., 2007; Wohlfahrt et al., 2008). This site likely receives more precipitation than many other tundra ecosystems and has no permafrost; thus the NEE response to climate could be less variable. However, as Kobbefjord is located in a coastal area, it is not surprising that it receives high pre-

precipitation, and other ecosystems such as coastal blanket bogs often receive even more precipitation without a clear impact of drought effect on the NEE sensitivity (Lund et al., 2015). Furthermore, permafrost adds another layer of complexity to the C dynamics (Christensen et al., 2004; Koven et al., 2011; Schuur et al., 2015). Although some studies showed similarities of CO₂ fluxes in various northern wetland ecosystems with and without permafrost (Lund et al., 2015), permafrost has a strong influence on the hydrology of peatlands (Åkerman and Johansson, 2008), and therefore their topography and distribution of vegetation (Johansson et al., 2013). Especially in the context of climate warming permafrost thaw can cause large changes to the ecosystems. Further, this study agrees with Parmentier et al. (2011) and Lund et al. (2012), who suggested that a longer growing season does not necessarily increase the net carbon uptake. Here a more negative NEE indicated a stronger C sink (i.e.) in 2012 compared to 2010. Parmentier et al. (2011) hypothesized that this behaviour is due to site-specific differences, such as meteorology and soil structure, and that changes in the carbon cycle with longer growing seasons will not be uniform around the Arctic. Thus, the effects of climate change on the tundra C balance of are not straightforward to infer.

NEE measured at Kobbefjord from 2008 to 2015 indicates a consistent sink of CO₂ (within a range of -17 to -41 g C m⁻²) with exception of the year 2011 ($+41$ g C m⁻²) (Table 2). The year 2011, associated with a major pest outbreak, reduced GPP more strongly than R_{eco} (Fig. 5) and Kobbefjord turned into a strong C source within an episodic single growing season. The return to substantial cumulative CO₂ sink rates following the extreme year of 2011 shows the ability of the ecosystem to recover from the disturbance (Lund et al., 2017). Indeed, the ecosystem not only shifted back from being a C source to a C sink, but it also changed rapidly from one year to the next. Thus we found evidence in Kobbefjord of ecosystem resilience to the meteorological variability, similar to other cases described in other northern sites (Peichl et al., 2014; Zona et al., 2014). Only a few reference sites have reported similar decreases in net C uptake, but in no case as large as the one observed here. Zona et al. (2014) described an effect of delayed responses to an unusual warm summer in Alaska. Their results suggested that vascular plants, which have enhanced their physiological activity during the warmer summer, might have difficulties readapting to cooler, but not atypical, conditions, which have provoked a significant decrease in GPP and R_{eco} the following year. In their study, the ecosystem returned to be a fairly strong C sink after 2 years, suggesting strong ecosystem resilience. Moreover, Hanis et al., 2015 have reported comparable C sink–C source variations in a Canadian fen within the growing season due to changes in the water table depth. Drier and warmer than normal conditions have triggered an increase in C source strength. Finally, during an extensive outbreak of autumn and winter moths in a sub-arctic birch forest in Sweden, Heliasz et al. (2011) observed

a similar decrease in net sink of C (most likely due to weaker GPP) across the growing season. However, the C source strength (NEE = 40.7 g C m⁻²) found in 2011 at Kobbefjord was higher compared to these other cases. To our knowledge, such abrupt disturbance concerning C sink strength in Arctic tundra has not been previously reported, excluding severely burned landscapes (Rocha and Shaver, 2011).

A combination of different factors could have led to the sharp change in C balance observed between 2010 and 2011, both physical and biological. The year 2010 had the warmest mean annual temperature (3.4 °C compared to the -0.4 °C mean annual temperature for 2008–2015) and the warmest mean wintertime temperature (-2.7 °C compared to the -6.79 °C mean for 2008–2015) (Fig. 2a). These climatic conditions generated the thinnest (maximum daily snow depth of 0.3 m compared to an average of 0.9 m) (Table 1) and shortest-lasting snowpack. Consequently, 2010 had the longest growing season (85 days) and very high growing season C uptake (-70 g C m⁻²). Increases in temperature can lead to high respiration rates during early winter (Commane et al., 2017; Zona et al., 2016) but also during the following summer (Helfter et al., 2015; Lund et al., 2012), which is related to soil temperature and snow dynamics. Further, in Kobbefjord the year 2011 had one of the lowest mean annual temperatures and mean wintertime temperatures (-1.7 and -6.1 °C respectively), which created the thickest (maximum daily snow depth of 1.4 m) and the longest-lasting snowpack, leading to the shortest growing season for the study period (only 47 days). According to Lund et al. (2012), soils will be insulated from low temperatures when below thick snowpack, which acts as a lid and prevents R_{eco} from being released to the atmosphere until the snowmelt period. Finally, an outbreak of larvae of the noctuid moth *Eurois occulta* occurred in 2011, overlapping the observed abrupt decrease in C sink strength. Although we cannot provide a quantification of change attributed to meteorological variations and biological disturbances, there is evidence showing that the moth outbreak could partially have decreased the C sink strength in Kobbefjord. In an undisturbed scenario, the meteorological conditions in 2015, colder and drier than the mean 2008–2015 period (Fig. 2) but similar to 2011, would have stimulated similar behaviours in terms of C fluxes. However, the cumulative fluxes in 2015 (Fig. 5) followed analogous patterns compared to other years. This evidence agrees with the literature (Callaghan et al., 2012b; Lund et al., 2017) on the fact that tundra systems can fluctuate in sink strength influenced by factors such as episodic disturbances or species shifts, events which are very difficult to predict.

4.3 Environmental forcing

Our data indicate that the importance of the main environmental controls (radiation and temperature) for C fluxes did vary across diurnal, seasonal and annual cycles, but also between time aggregations. The hourly variability of NEE and

GPP (Figs. 7 and 8) was mostly dependent on PAR because of the threshold nature on radiation control on GPP. Overall, the results indicate that environmental factors that can change rapidly such as PAR will have a high influence on short timescales (Stoy et al., 2014). The increased importance of PAR at 08:00 and 20:00 h WGST coincides with the sharp gradient in light at dawn and dusk (Fig. 8). The control of PAR on GPP is not a new finding itself, but the random forest approach helps to quantify its importance. There is no GPP at night, and therefore there will be a strong increase/decrease in GPP at dawn/dusk. The seasonal pattern also showed that radiation is the single main driver for NEE and GPP between early June and early October, supported by the longer daytime. Further, PAR appeared to be a limiting factor for annual NEE in 2011, increasing further the complexity around this anomalous year. These results agree with the literature (Groendahl et al., 2007; Stoy et al., 2014), suggesting that the uptake of CO₂ is partially controlled by radiation for the photosynthetic physiology at the leaf scale. Arctic plants are usually well adapted to environments with low light levels, reporting near-maximum rates ranging from 10 to 25 °C (Oechel and Billings, 1992; Shaver and Kummerow, 1992).

Photosynthesis is restricted by low temperature, so enzymatically driven processes such as carbon fixation are more sensitive to low temperature than the light-driven biophysical reactions (Chapin et al., 2011). In this paper the daily, weekly and monthly aggregated variability of C fluxes was primarily linked to T_{air} . Moreover, the random forest analyses revealed a strong diurnal pattern with a marked contribution of T_{air} to variations in NEE and GPP (both at night-time and between 08:00 and 18:00 h WGST). These results agree with Lindroth et al. (2007), who recognized T_{air} as the key driver of NEE seasonal trends in northern peatlands. However, in this analysis both NEE and GPP had similar responses to common environmental forcing, contrary to the results in Reichstein et al. (2007). In order to circumvent the potential circularity conflicts based on the use of partitioning products, we filtered daytime NEE (true GPP) and night-time NEE (true R_{eco}), obtaining very similar results (Table S2). Further, our data also suggest that R_{eco} is often dominated by air temperature. The patterns observed here are in agreement with findings on plant respiration dynamics (Heskel et al., 2016; Lloyd and Taylor, 1994; Tjoelker et al., 2001).

In this study, environmental drivers related to water availability such as VPD and precipitation were not found to be as influential as other assessed variables. We did not find significant relationships between CO₂ fluxes and the water table depth. Thus, there was no apparent water limitation on carbon dynamics during the 8-year period. However, the complex interactions based on changes in temperature and soil moisture particularly over full annual cycles and for sites with permafrost, should be further explored. Our results contrast with Strachan et al. (2015), who described water table depth as an important driver regulating the CO₂ balance, and

others, who found that CO₂ emissions increase during dry years due to increased decomposition rates and a reduction in GPP (Aurela et al., 2007; Lund et al., 2007; Oechel et al., 1993; Peichl et al., 2014), whereas other sites act as sinks during relatively wet years (Lafleur et al., 1997). The fen in Kobbefjord is probably quite resistant to droughts since it is fed with water from the surroundings.

5 Conclusions

We have analysed eight snow-free periods in 8 consecutive years in a West Greenland tundra (64° N) focusing on the net ecosystem exchange (NEE) of CO₂ and its photosynthetic inputs (GPP) and respiration outputs (R_{eco}). Here, the NEE gap filling exposed inherent uncertainties in the seasonal C budget calculation, but there were also inconsistencies between the flux partitioning approaches used. We find that Kobbefjord acted as a consistent sink of CO₂ during the years 2008–2015, except 2011, which was associated with a major pest outbreak. The results do not show a marked meteorological effect on the net C uptake. However, the relative insensitivity of NEE during the snow-free period was driven by the correlated, balancing responses of GPP and R_{eco} , both more variable than NEE and sensitive to temperature and insolation. In this paper we show a tendency towards larger GPP and R_{eco} during wetter and warmer years. The anomalous year 2011, affected by a biological disturbance, constituted a relatively strong source of CO₂ and reduced GPP more strongly than R_{eco} . A novel analysis assessing the changes of environmental forcing across diurnal, seasonal and annual timescales unmasked patterns of functional responses to C fluxes.

Despite the fact that we analysed an 8-year data set, the results do not provide a complete picture due to the lack of year-round data (Grøndahl et al., 2008). The snow season should be taken into account for a comprehensive understanding of complete C budget (Aurela et al., 2002; Commane et al., 2017; Zona et al., 2016) and the delayed effect of wintertime-based variables such as snow depth or snow cover on the C fluxes. Because some studies have suggested that GPP and R_{eco} increase with observed changes in climate and NEE trends remain unclear (Lund et al., 2012), it is challenging to produce strong evidence while the data remains scarce and fragmented. Hence, there is a need for increased efforts in monitoring Arctic ecosystem changes over the full annual cycle (Euskirchen et al., 2012; Grøndahl et al., 2008). Future work is also required with C flux modelling in order to explore process-based insights of C exchange balance in the Arctic tundra and the interactions of photosynthesis and R_{eco} with changes in C stocks.

Data availability. Measurement data from the Greenland Ecosystem Monitoring (GEM) programme are freely available from the GEM database (<http://data.g-e-m.dk>). Post-processed data

and scripts are available from the authors upon request (elb@bios.au.dk).

The Supplement related to this article is available online at <https://doi.org/10.5194/bg-14-4467-2017-supplement>.

Author contributions. EL-B, ML, MW, TRC and MPT designed the experiment. Data preparation and analysis were primarily performed by EL-B with contribution from ML (eddy covariance data processing, data quality control and LRC partitioning), MW and TRC (experimental set up), BUH (data gathering from Nuuk Ecological Research Operations, GeoBasis), AW-N (daily estimate of the timing of snowmelt) and J-FE (random forest approach). EL-B prepared the manuscript with active contributions from all co-authors.

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

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References

- ACIA: Arctic Climate Impact Assessment, Cambridge University Press, New York, 1042 pp., 2005.
- Åkerman, H. J. and Johansson, M.: Thawing permafrost and thicker active layers in sub-arctic Sweden, *Permafrost Periglac.*, 19, 279–292, <https://doi.org/10.1002/ppp.626>, 2008.
- AMAP: Snow, Water, Ice and Permafrost in the Arctic (SWIPA), Oslo, Norway, 2011.
- Arndal, M. F., Illeris, L., Michelsen, A., Albert, K., Tamstorf, M., and Hansen, B. U.: Seasonal Variation in Gross Ecosystem Production, Plant Biomass, and Carbon and Nitrogen Pools in Five High Arctic Vegetation Types, *Arct. Antarct. Alp. Res.*, 41, 164–173, <https://doi.org/10.1657/1938-4246-41.2.164>, 2009.
- Aurela, M., Laurila, T., and Tuovinen, J.-P.: Annual CO₂ balance of a subarctic fen in northern Europe: importance of the wintertime efflux, *J. Geophys. Res.-Atmos.*, 107, ACH 17-11–ACH 17-12, <https://doi.org/10.1029/2002JD002055>, 2002.
- Aurela, M., Laurila, T., and Tuovinen, J.-P.: The timing of snow melt controls the annual CO₂ balance in a subarctic fen, *Geophys. Res. Lett.*, 31, L16119, <https://doi.org/10.1029/2004GL020315>, 2004.
- Aurela, M., Riutta, T., Laurila, T., Tuovinen, J.-P., Vesala, T., Tuittila, E.-S., Rinne, J., Haapanala, S., and Laine, J.: CO₂ exchange of a sedge fen in southern Finland – the impact of a drought period, *Tellus B*, 59, 826–837, <https://doi.org/10.1111/j.1600-0889.2007.00309.x>, 2007.
- Baccini, A., Goetz, S. J., Walker, W. S., Laporte, N. T., Sun, M., Sulla-Menashe, D., Hackler, J., Beck, P. S. A., Dubayah, R., Friedl, M. A., Samanta, S., and Houghton, R. A.: Estimated carbon dioxide emissions from tropical deforestation improved by carbon-density maps, *Nat. Clim. Change*, 2, 182–185, 2012.
- Bay, C., Aastrup, P., and Nymand, J.: The NERO line. A vegetation transect in Kobbefjord, West Greenland, National Environmental Research Institute, Aarhus University, Aarhus, 2008.
- Breiman, L.: Random forests, *Mach. Learn.*, 45, 5–32, <https://doi.org/10.1023/a:1010933404324>, 2001.
- Bubier, L. J., Bhatia, G., Moore, R. T., Roulet, T. N., and Lafleur, M. P.: Spatial and temporal variability in growing-season net ecosystem carbon dioxide exchange at a large peatland in Ontario, Canada, *Ecosystems*, 6, 353–367, <https://doi.org/10.1007/s10021-003-0125-0>, 2003.
- Callaghan, T. V., Johansson, M., Brown, R. D., Groisman, P. Y., Labba, N., Radionov, V., Barry, R. G., Bulygina, O. N., Essery, R. L. H., Frolov, D. M., Golubev, V. N., Grenfell, T. C., Petrushina, M. N., Razuvaev, V. N., Robinson, D. A., Romanov, P., Shindell, D., Shmakin, A. B., Sokratov, S. A., Warren, S., and Yang, D.: The changing face of arctic snow cover: a synthesis of observed and projected changes, *AMBIO*, 40, 17–31, <https://doi.org/10.1007/s13280-011-0212-y>, 2012a.
- Callaghan, T. V., Johansson, M., Key, J., Prowse, T., Ananicheva, M., and Klepikov, A.: Feedbacks and interactions: from the arctic cryosphere to the climate system, *AMBIO*, 40, 75–86, <https://doi.org/10.1007/s13280-011-0215-8>, 2012b.
- Callaghan, T. V., Johansson, M., Prowse, T. D., Olsen, M. S., and Reiersen, L.-O.: Arctic cryosphere: changes and Impacts, *AMBIO*, 40, 3–5, <https://doi.org/10.1007/s13280-011-0210-0>, 2012c.
- Cappelen, J.: Greenland – DMI Historical Climate Data Collection 1873–2012, Danish Meteorological Institute, Copenhagen, 2013.
- Cappelen, J.: Greenland – DMI Historical Climate Data Collection 1784–2015, Danish Meteorological Institute, Copenhagen, 2016.
- CAVM Team: Circumpolar Arctic Vegetation Map, scale 1 : 7500000, in: Conservation of Arctic Flora and Fauna (CAFF) Map No. 1, US Fish and Wildlife Service, Anchorage, Alaska, 2003.
- Chapin, F. S., Matson, P. A., and Vitousek, P. M.: Principles of Terrestrial Ecosystem Ecology, Springer, New York, 2011.
- Christensen, J. H., Hewitson, B., Busuioac, A., Chen, A., Gao, X., Held, I., Jones, R., Kolli, R. K., Kwon, W.-T., Laprise, R., Magaña Rueda, V., Mearns, L., Menéndez, C. G., Räisänen, J., Rinke, A., Sarr, A., and Whetton, P.: Regional Climate Projections, in: Climate Change 2007: Climate Change 2007: The Physical Science Basis: Working Group I Contribution to the Fourth Assessment Report of the IPCC, edited by: Solomon, S., Qin, D., Manning, M., Chen, Z., Marquis, M., Averyt, K. B.,

- Tignor, M., and Miller, H. L., Cambridge University Press, New York, 2007.
- Christensen, T. R., Johansson, T., Åkerman, H. J., Mastepanov, M., Malmer, N., Friborg, T., Crill, P., and Svensson, B.: Thawing sub-arctic permafrost: effects on vegetation and methane emissions, *Geophys. Res. Lett.*, 31, L04501, <https://doi.org/10.1029/2003GL018680>, 2004.
- Christensen, T. R., Jackowicz-Korczyński, M., Aurela, M., Crill, P., Heliasz, M., Mastepanov, M., and Friborg, T.: Monitoring the multi-year carbon balance of a subarctic tundra mire with micrometeorological techniques, *AMBIO*, 41, 207–217, <https://doi.org/10.1007/s13280-012-0302-5>, 2012.
- Commane, R., Lindaas, J., Benmergui, J., Luus, K. A., Chang, R. Y.-W., Daube, B. C., Euskirchen, E. S., Henderson, J. M., Karion, A., Miller, J. B., Miller, S. M., Parazoo, N. C., Randerson, J. T., Sweeney, C., Tans, P., Thoning, K., Veraverbeke, S., Miller, C. E., and Wofsy, S. C.: Carbon dioxide sources from Alaska driven by increasing early winter respiration from Arctic tundra, *P. Natl. Acad. Sci. USA*, 114, 5361–5366, <https://doi.org/10.1073/pnas.1618567114>, 2017.
- Dragomir, C. M., Klaassen, W., Voiculescu, M., Georgescu, L. P., and van der Laan, S.: Estimating annual CO₂ flux for Lutjewad Station using three different gap-filling techniques, *Sci. World J.*, 2012, 10, <https://doi.org/10.1100/2012/842893>, 2012.
- Elberling, B., Nordstrøm, C., Grøndahl, L., Sjøgaard, H., Friborg, T., Christensen, T. R., Ström, L., Marchand, F., and Nijs, I.: High-arctic soil CO₂ and CH₄ production controlled by temperature, water, freezing and snow, in: *Advances in Ecological Research*, Academic Press, 441–472, 2008.
- Euskirchen, E. S., Bret-Harte, M. S., Scott, G. J., Edgar, C., and Shaver, G. R.: Seasonal patterns of carbon dioxide and water fluxes in three representative tundra ecosystems in northern Alaska, *Ecosphere*, 3, 1–19, <https://doi.org/10.1890/ES11-00202.1>, 2012.
- Exbrayat, J.-F. and Williams, M.: Quantifying the net contribution of the historical Amazonian deforestation to climate change, *Geophys. Res. Lett.*, 42, 2968–2976, <https://doi.org/10.1002/2015GL063497>, 2015.
- Falge, E., Baldocchi, D., Olson, R., Anthoni, P., Aubinet, M., Bernhofer, C., Burba, G., Ceulemans, R., Clement, R., Dolman, H., Granier, A., Gross, P., Grünwald, T., Hollinger, D., Jensen, N.-O., Katul, G., Keronen, P., Kowalski, A., Lai, C. T., Law, B. E., Meyers, T., Moncrieff, J., Moors, E., Munger, J. W., Pilegaard, K., Rannik, Ü., Rebmann, C., Suyker, A., Tenhunen, J., Tu, K., Verma, S., Vesala, T., Wilson, K., and Wofsy, S.: Gap filling strategies for defensible annual sums of net ecosystem exchange, *Agr. Forest Meteorol.*, 107, 43–69, [https://doi.org/10.1016/S0168-1923\(00\)00225-2](https://doi.org/10.1016/S0168-1923(00)00225-2), 2001.
- Gilmanov, T. G., Verma, S. B., Sims, P. L., Meyers, T. P., Bradford, J. A., Burba, G. G., and Suyker, A. E.: Gross primary production and light response parameters of four Southern Plains ecosystems estimated using long-term CO₂-flux tower measurements, *Global Biogeochem. Cy.*, 17, <https://doi.org/10.1029/2002GB002023>, 2003.
- Goulden, M. L. and Crill, P. M.: Automated measurements of CO₂ exchange at the moss surface of a black spruce forest, *Tree Physiol.*, 17, 537–542, <https://doi.org/10.1093/treephys/17.8-9.537>, 1997.
- Groendahl, L., Friborg, T., and Soegaard, H.: Temperature and snow-melt controls on interannual variability in carbon exchange in the high Arctic, *Theor. Appl. Climatol.*, 88, 111–125, <https://doi.org/10.1007/s00704-005-0228-y>, 2007.
- Grolemund, G. and Wickham, H.: Dates and Times Made Easy with lubridate, *J. Stat. Softw.*, 40, 1–25, 2011.
- Grøndahl, L., Friborg, T., Christensen, T. R., Ekberg, A., Elberling, B., Illeris, L., Nordstrøm, C., Rennermalm, Å., Sjøgaard, C., and Sjøgaard, H.: Spatial and inter-annual variability of trace gas fluxes in a heterogeneous high-arctic landscape, in: *Advances in Ecological Research*, edited by: Meltofte, H., Christensen, T. R., Elberling, B., Forchhammer, M. C., and Rasch, M., Academic Press, 473–498, 2008.
- Hanis, K. L., Amiro, B. D., Tenuta, M., Papakyriakou, T., and Swystun, K. A.: Carbon exchange over four growing seasons for a subarctic sedge fen in northern Manitoba, Canada, *Arctic Sci.*, 1, 27–44, <https://doi.org/10.1139/as-2015-0003>, 2015.
- Helfter, C., Campbell, C., Dinsmore, K. J., Drewer, J., Coyle, M., Anderson, M., Skiba, U., Nemitz, E., Billett, M. F., and Sutton, M. A.: Drivers of long-term variability in CO₂ net ecosystem exchange in a temperate peatland, *Biogeosciences*, 12, 1799–1811, <https://doi.org/10.5194/bg-12-1799-2015>, 2015.
- Heliasz, M., Johansson, T., Lindroth, A., Mölder, M., Mastepanov, M., Friborg, T., Callaghan, T. V., and Christensen, T. R.: Quantification of C uptake in subarctic birch forest after setback by an extreme insect outbreak, *Geophys. Res. Lett.*, 38, L01704, <https://doi.org/10.1029/2010GL044733>, 2011.
- Heskel, M. A., O'Sullivan, O. S., Reich, P. B., Tjoelker, M. G., Weerasinghe, L. K., Penillard, A., Egerton, J. J. G., Creek, D., Bloomfield, K. J., Xiang, J., Sinca, F., Stangl, Z. R., Martinez-de la Torre, A., Griffin, K. L., Huntingford, C., Hurry, V., Meir, P., Turnbull, M. H., and Atkin, O. K.: Convergence in the temperature response of leaf respiration across biomes and plant functional types, *P. Natl. Acad. Sci. USA*, 113, 3832–3837, <https://doi.org/10.1073/pnas.1520282113>, 2016.
- Hinkler, J., Pedersen, S. B., Rasch, M., and Hansen, B. U.: Automatic snow cover monitoring at high temporal and spatial resolution, using images taken by a standard digital camera, *Int. J. Remote Sens.*, 23, 4669–4682, <https://doi.org/10.1080/01431160110113881>, 2002.
- Højstrup, J.: A statistical data screening procedure, *Meas. Sci. Technol.*, 4, 153–157, 1993.
- Hugelius, G., Strauss, J., Zubrzycki, S., Harden, J. W., Schuur, E. A. G., Ping, C.-L., Schirmer, L., Grosse, G., Michaelson, G. J., Koven, C. D., O'Donnell, J. A., Elberling, B., Mishra, U., Camill, P., Yu, Z., Palmtag, J., and Kuhry, P.: Estimated stocks of circumpolar permafrost carbon with quantified uncertainty ranges and identified data gaps, *Biogeosciences*, 11, 6573–6593, <https://doi.org/10.5194/bg-11-6573-2014>, 2014.
- IPCC: Climate Change 2007: The Physical Science Basis: Working Group I Contribution to the Fourth Assessment Report of the IPCC, Cambridge University Press, New York, 2007.
- Jensen, L. M. and Christensen, T. R.: Nuuk Ecological Research Operations, 7th Annual Report 2013, Aarhus University, DCE-Danish Centre for Environment and Energy, Roskilde, Denmark, 94, 2014.
- Johansson, M., Callaghan, T. V., Bosiö, J., Åkerman, H. J., Jackowicz-Korczyński, M., and Christensen, T. R.: Rapid responses of permafrost and vegetation to experimentally increased

- snow cover in sub-arctic Sweden, *Environ. Res. Lett.*, 8, 035025, <https://doi.org/10.1088/1748-9326/8/3/035025>, 2013.
- Koven, C. D., Ringeval, B., Friedlingstein, P., Ciais, P., Cadule, P., Khvorostyanov, D., Krinner, G., and Tarnocai, C.: Permafrost carbon-climate feedbacks accelerate global warming, *P. Natl. Acad. Sci. USA*, 108, 14769–14774, <https://doi.org/10.1073/pnas.1103910108>, 2011.
- Lafleur, P. M., McCaughey, J. H., Joiner, D. W., Bartlett, P. A., and Jelinski, D. E.: Seasonal trends in energy, water, and carbon dioxide fluxes at a northern boreal wetland, *J. Geophys. Res.-Atmos.*, 102, 29009–29020, <https://doi.org/10.1029/96JD03326>, 1997.
- Lafleur, P. M., Humphreys, E. R., St. Louis, V. L., Myklebust, M. C., Papakyriakou, T., Poissant, L., Barker, J. D., Pilote, M., and Swystun, K. A.: Variation in Peak Growing Season Net Ecosystem Production Across the Canadian Arctic, *Environ. Sci. Technol.*, 46, 7971–7977, <https://doi.org/10.1021/es300500m>, 2012.
- Lasslop, G., Reichstein, M., Papale, D., Richardson, A. D., Arneeth, A., Barr, A., Stoy, P., and Wohlfahrt, G.: Separation of net ecosystem exchange into assimilation and respiration using a light response curve approach: critical issues and global evaluation, *Glob. Change Biol.*, 16, 187–208, <https://doi.org/10.1111/j.1365-2486.2009.02041.x>, 2010.
- Lasslop, G., Migliavacca, M., Bohrer, G., Reichstein, M., Bahn, M., Ibrom, A., Jacobs, C., Kolari, P., Papale, D., Vesala, T., Wohlfahrt, G., and Cescatti, A.: On the choice of the driving temperature for eddy-covariance carbon dioxide flux partitioning, *Biogeosciences*, 9, 5243–5259, <https://doi.org/10.5194/bg-9-5243-2012>, 2012.
- Lindroth, A., Lund, M., Nilsson, M., Aurela, M., Christensen, T. R., Laurila, T., Rinne, J., Riutta, T., Sagerfors, J., Ström, L., Tuovinen, J.-P., and Vesala, T.: Environmental controls on the CO₂ exchange in north European mires, *Tellus B*, 59, 812–825, <https://doi.org/10.1111/j.1600-0889.2007.00310.x>, 2007.
- Lloyd, J. and Taylor, J. A.: On the temperature dependence of soil respiration, *Funct. Ecol.*, 8, 315–323, <https://doi.org/10.2307/2389824>, 1994.
- Lund, M., Lindroth, A., Christensen, T. R., and Ström, L.: Annual CO₂ balance of a temperate bog, *Tellus B*, 59, 804–811, <https://doi.org/10.1111/j.1600-0889.2007.00303.x>, 2007.
- Lund, M., Lafleur, P. M., Roulet, N. T., Lindroth, A., Christensen, T. R., Aurela, M., Chojnicki, B. H., Flanagan, L. B., Humphreys, E. R., Laurila, T., Oechel, W. C., Olejnik, J., Rinne, J., Schubert, P. E. R., and Nilsson, M. B.: Variability in exchange of CO₂ across 12 northern peatland and tundra sites, *Glob. Change Biol.*, 16, 2436–2448, <https://doi.org/10.1111/j.1365-2486.2009.02104.x>, 2010.
- Lund, M., Falk, J. M., Friborg, T., Mbufong, H. N., Sigsgaard, C., Soegaard, H., and Tamstorf, M. P.: Trends in CO₂ exchange in a high Arctic tundra heath, 2000–2010, *J. Geophys. Res.-Biogeo.*, 117, G02001, <https://doi.org/10.1029/2011JG001901>, 2012.
- Lund, M., Bjerke, J. W., Drake, B. G., Engelsen, O., Hansen, G. H., Parmentier, F. J. W., Powell, T. L., Silvennoinen, H., Sottocornola, M., Tømmervik, H., Weldon, S., and Rasse, D. P.: Low impact of dry conditions on the CO₂ exchange of a Northern-Norwegian blanket bog, *Environ. Res. Lett.*, 10, 025004, <https://doi.org/10.1088/1748-9326/10/2/025004>, 2015.
- Lund, M., Raundrup, K., Westergaard-Nielsen, A., López-Blanco, E., Nymand, J., and Aastrup, P.: Larval outbreaks in West Greenland: instant and subsequent effects on tundra ecosystem productivity and CO₂ exchange, *AMBIO*, 46, 26–38, <https://doi.org/10.1007/s13280-016-0863-9>, 2017.
- Mastepanov, M., Sigsgaard, C., Tagesson, T., Ström, L., Tamstorf, M. P., Lund, M., and Christensen, T. R.: Revisiting factors controlling methane emissions from high-Arctic tundra, *Biogeosciences*, 10, 5139–5158, <https://doi.org/10.5194/bg-10-5139-2013>, 2013.
- Mbufong, H. N., Lund, M., Aurela, M., Christensen, T. R., Eugster, W., Friborg, T., Hansen, B. U., Humphreys, E. R., Jackowicz-Korczynski, M., Kutzbach, L., Lafleur, P. M., Oechel, W. C., Parmentier, F. J. W., Rasse, D. P., Rocha, A. V., Sachs, T., van der Molen, M. K., and Tamstorf, M. P.: Assessing the spatial variability in peak season CO₂ exchange characteristics across the Arctic tundra using a light response curve parameterization, *Biogeosciences*, 11, 4897–4912, <https://doi.org/10.5194/bg-11-4897-2014>, 2014.
- McGuire, A. D., Anderson, L. G., Christensen, T. R., Dallimore, S., Guo, L., Hayes, D. J., Heimann, M., Lorenson, T. D., Macdonald, R. W., and Roulet, N. T.: Sensitivity of the carbon cycle in the Arctic to climate change, *Ecol. Monogr.*, 79, 523–555, <https://doi.org/10.1890/08-2025.1>, 2009.
- McGuire, A. D., Christensen, T. R., Hayes, D., Heroult, A., Euskirchen, E., Kimball, J. S., Koven, C., Lafleur, P., Miller, P. A., Oechel, W., Peylin, P., Williams, M., and Yi, Y.: An assessment of the carbon balance of Arctic tundra: comparisons among observations, process models, and atmospheric inversions, *Biogeosciences*, 9, 3185–3204, <https://doi.org/10.5194/bg-9-3185-2012>, 2012.
- Meltofte, H., Christensen, T. R., Elberling, B., Forchhammer, M. C., and Rasch, M. (Eds.): Introduction, in: *Advances in Ecological Research*, Academic Press, 1–12, 2008.
- Moffat, A. M., Papale, D., Reichstein, M., Hollinger, D. Y., Richardson, A. D., Barr, A. G., Beckstein, C., Braswell, B. H., Churkina, G., Desai, A. R., Falge, E., Gove, J. H., Heimann, M., Hui, D., Jarvis, A. J., Kattge, J., Noormets, A., and Stauch, V. J.: Comprehensive comparison of gap-filling techniques for eddy covariance net carbon fluxes, *Agr. Forest Meteorol.*, 147, 209–232, <https://doi.org/10.1016/j.agrformet.2007.08.011>, 2007.
- Moncrieff, J. B., Massheder, J. M., de Bruin, H., Elbers, J., Friborg, T., Heusinkveld, B., Kabat, P., Scott, S., Soegaard, H., and Verhoef, A.: A system to measure surface fluxes of momentum, sensible heat, water vapour and carbon dioxide, *J. Hydrol.*, 188–189, 589–611, [https://doi.org/10.1016/S0022-1694\(96\)03194-0](https://doi.org/10.1016/S0022-1694(96)03194-0), 1997.
- Moore, C. J.: Frequency response corrections for eddy correlation systems, *Bound.-Layer Meteorol.*, 37, 17–35, <https://doi.org/10.1007/BF00122754>, 1986.
- Nelson, F. E. and Outcalt, S. I.: A computational method for prediction and regionalization of permafrost, *Arctic Alpine Res.*, 19, 279–288, <https://doi.org/10.2307/1551363>, 1987.
- Oechel, W. C. and Billings, W.: Effects of global change on the carbon balance of arctic plants and ecosystems, in: *Arctic Ecosystems in a Changing Climate: an Ecophysiological Perspective*, Academic Press, San Diego, 139–168, 1992.
- Oechel, W. C., Hastings, S. J., Vourlitis, G., Jenkins, M., Riechers, G., and Grulke, N.: Recent change of Arctic tundra ecosystems from a net carbon dioxide sink to a source, *Nature*, 361, 520–523, 1993.

- Papale, D., Reichstein, M., Aubinet, M., Canfora, E., Bernhofer, C., Kutsch, W., Longdoz, B., Rambal, S., Valentini, R., Vesala, T., and Yakir, D.: Towards a standardized processing of Net Ecosystem Exchange measured with eddy covariance technique: algorithms and uncertainty estimation, *Biogeosciences*, 3, 571–583, <https://doi.org/10.5194/bg-3-571-2006>, 2006.
- Parmentier, F. J. W., van der Molen, M. K., van Huissteden, J., Karsanaev, S. A., Kononov, A. V., Suzdalov, D. A., Maximov, T. C., and Dolman, A. J.: Longer growing seasons do not increase net carbon uptake in the northeastern Siberian tundra, *J. Geophys. Res.-Biogeo.*, 116, G04013, <https://doi.org/10.1029/2011JG001653>, 2011.
- Pedregosa, F., Varoquaux, G., Gramfort, A., Michel, V., Thirion, B., Grisel, O., Blondel, M., Prettenhofer, P., Weiss, R., Dubourg, V., Vanderplas, J., Passos, A., Cournapeau, D., Brucher, M., Perrot, M., and Duchesnay, É.: Scikit-learn: machine learning in python, *J. Mach. Learn. Res.*, 12, 2825–2830, 2011.
- Peichl, M., Öquist, M., Löfvenius, M. O., Ilstedt, U., Sagerfors, J., Grelle, A., Lindroth, A., and Nilsson, M.: A 12-year record reveals pre-growing season temperature and water table level threshold effects on the net carbon dioxide exchange in a boreal fen, *Environ. Res. Lett.*, 9, 055006, <https://doi.org/10.1088/1748-9326/9/5/055006>, 2014.
- Pirk, N., Sievers, J., Mertes, J., Parmentier, F.-J. W., Mastepanov, M., and Christensen, T. R.: Spatial variability of CO₂ uptake in polygonal tundra: assessing low-frequency disturbances in eddy covariance flux estimates, *Biogeosciences*, 14, 3157–3169, <https://doi.org/10.5194/bg-14-3157-2017>, 2017.
- R Development Core Team: R: A Language and Environment for Statistical Computing, R Foundation for Statistical Computing, Vienna, Austria, 2015.
- Reichstein, M., Moffat, A. M., Wutzler, T., and Sickel, K.: REdyProc: Data processing and plotting utilities of (half-)hourly eddy-covariance measurements, R package version 0.8-2/r14, 2016.
- Reichstein, M., Falge, E., Baldocchi, D., Papale, D., Aubinet, M., Berbigier, P., Bernhofer, C., Buchmann, N., Gilmanov, T., Granier, A., Grünwald, T., Havránková, K., Ilvesniemi, H., Janous, D., Knohl, A., Laurila, T., Lohila, A., Loustau, D., Matteucci, G., Meyers, T., Miglietta, F., Ourcival, J.-M., Pumpanen, J., Rambal, S., Rotenberg, E., Sanz, M., Tenhunen, J., Seufert, G., Vaccari, F., Vesala, T., Yakir, D., and Valentini, R.: On the separation of net ecosystem exchange into assimilation and ecosystem respiration: review and improved algorithm, *Glob. Change Biol.*, 11, 1424–1439, <https://doi.org/10.1111/j.1365-2486.2005.001002.x>, 2005.
- Reichstein, M., Papale, D., Valentini, R., Aubinet, M., Bernhofer, C., Knohl, A., Laurila, T., Lindroth, A., Moors, E., Pilegaard, K., and Seufert, G.: Determinants of terrestrial ecosystem carbon balance inferred from European eddy covariance flux sites, *Geophys. Res. Lett.*, 34, L01402, <https://doi.org/10.1029/2006GL027880>, 2007.
- Richardson, A. D., Hollinger, D. Y., Aber, J. D., Ollinger, S. V., and Braswell, B. H.: Environmental variation is directly responsible for short- but not long-term variation in forest-atmosphere carbon exchange, *Glob. Change Biol.*, 13, 788–803, <https://doi.org/10.1111/j.1365-2486.2007.01330.x>, 2007.
- Rocha, A. V. and Shaver, G. R.: Burn severity influences post-fire CO₂ exchange in arctic tundra, *Ecol. Appl.*, 21, 477–489, <https://doi.org/10.1890/10-0255.1>, 2011.
- Runkle, B. R. K., Sachs, T., Wille, C., Pfeiffer, E.-M., and Kutzbach, L.: Bulk partitioning the growing season net ecosystem exchange of CO₂ in Siberian tundra reveals the seasonality of its carbon sequestration strength, *Biogeosciences*, 10, 1337–1349, <https://doi.org/10.5194/bg-10-1337-2013>, 2013.
- Ryan, J. A. and Ulrich, J. M.: xts: eXtensible Time Series. R package version 0.9-7, 2014.
- Schuur, E. A. G., McGuire, A. D., Schadel, C., Grosse, G., Harden, J. W., Hayes, D. J., Hugelius, G., Koven, C. D., Kuhry, P., Lawrence, D. M., Natali, S. M., Olefeldt, D., Romanovsky, V. E., Schaefer, K., Turetsky, M. R., Treat, C. C., and Vonk, J. E.: Climate change and the permafrost carbon feedback, *Nature*, 520, 171–179, <https://doi.org/10.1038/nature14338>, 2015.
- Serreze, M. C. and Barry, R. G.: Processes and impacts of Arctic amplification: a research synthesis, *Global Planet. Change*, 77, 85–96, <https://doi.org/10.1016/j.gloplacha.2011.03.004>, 2011.
- Shaver, G. and Kummerow, J.: Phenology, resource allocation, and growth of arctic vascular plants, in: *Arctic Ecosystems in a Changing Climate: An Ecophysiological Perspective*, edited by: Svoboda, J., Academic Press, New York, 1992.
- Stoy, P. C., Trowbridge, A. M., and Bauerle, W. L.: Controls on seasonal patterns of maximum ecosystem carbon uptake and canopy-scale photosynthetic light response: contributions from both temperature and photoperiod, *Photosynth. Res.*, 119, 49–64, <https://doi.org/10.1007/s11120-013-9799-0>, 2014.
- Strachan, I. B., Pelletier, L., and Bonneville, M.-C.: Inter-annual variability in water table depth controls net ecosystem carbon dioxide exchange in a boreal bog, *Biogeochemistry*, 127, 99–111, <https://doi.org/10.1007/s10533-015-0170-8>, 2015.
- Tagesson, T., Mölder, M., Mastepanov, M., Sigsgaard, C., Tamstorf, M. P., Lund, M., Falk, J. M., Lindroth, A., Christensen, T. R., and Ström, L.: Land-atmosphere exchange of methane from soil thawing to soil freezing in a high-Arctic wet tundra ecosystem, *Glob. Change Biol.*, 18, 1928–1940, <https://doi.org/10.1111/j.1365-2486.2012.02647.x>, 2012.
- Tarnocai, C., Canadell, J. G., Schuur, E. A. G., Kuhry, P., Mazhitova, G., and Zimov, S.: Soil organic carbon pools in the northern circumpolar permafrost region, *Global Biogeochem. Cy.*, 23, GB2023, <https://doi.org/10.1029/2008GB003327>, 2009.
- Tjoelker, M. G., Oleksyn, J., and Reich, P. B.: Modelling respiration of vegetation: evidence for a general temperature-dependent Q₁₀, *Glob. Change Biol.*, 7, 223–230, <https://doi.org/10.1046/j.1365-2486.2001.00397.x>, 2001.
- Walker, D. A., Raynolds, M. K., Daniëls, F. J. A., Einarsson, E., Elvebakk, A., Gould, W. A., Katenin, A. E., Kholod, S. S., Markon, C. J., Melnikov, E. S., Moskalenko, N. G., Talbot, S. S., Yurtsev, B. A., and the other members of the C.T.: The Circumpolar Arctic vegetation map, *J. Veg. Sci.*, 16, 267–282, <https://doi.org/10.1111/j.1654-1103.2005.tb02365.x>, 2005.
- Webb, E. K., Pearman, G. I., and Leuning, R.: Correction of flux measurements for density effects due to heat and water vapour transfer, *Q. J. Roy. Meteor. Soc.*, 106, 85–100, <https://doi.org/10.1002/qj.49710644707>, 1980.
- Westergaard-Nielsen, A., Lund, M., Hansen, B. U., and Tamstorf, M.: Camera derived vegetation greenness in-

- dex as proxy for gross primary production in a low Arctic wetland area, *ISPRS J. Photogramm.*, 86, 89–99, <https://doi.org/10.1016/j.isprsjprs.2013.09.006>, 2013.
- Westergaard-Nielsen, A., Lund, M., Pedersen, S. H., Schmidt, N. M., Klosterman, S., Abermann, J., and Hansen, B. U.: Transitions in high-Arctic vegetation growth patterns and ecosystem productivity from 2000–2013 tracked with cameras, *AMBIO*, 46, Supplement 1, 39–52, <https://doi.org/10.1007/s13280-016-0864-8>, 2015.
- Williams, M., Eugster, W., Rastetter, E. B., McFadden, J. P., and Chapin III, F. S.: The controls on net ecosystem productivity along an Arctic transect: a model comparison with flux measurements, *Glob. Change Biol.*, 6, 116–126, <https://doi.org/10.1046/j.1365-2486.2000.06016.x>, 2000.
- Williams, M., Schwarz, P. A., Law, B. E., Irvine, J., and Kurpius, M. R.: An improved analysis of forest carbon dynamics using data assimilation, *Glob. Change Biol.*, 11, 89–105, <https://doi.org/10.1111/j.1365-2486.2004.00891.x>, 2005.
- Wohlfahrt, G., Hammerle, A., Haslwanter, A., Bahn, M., Tappeiner, U., and Cernusca, A.: Seasonal and inter-annual variability of the net ecosystem CO₂ exchange of a temperate mountain grassland: effects of weather and management, *J. Geophys. Res.-Atmos.*, 113, D08110, <https://doi.org/10.1029/2007JD009286>, 2008.
- Zeileis, A. and Grothendieck, G.: zoo: S3 infrastructure for regular and irregular time series, *J. Stat. Softw.*, 14, 1–27, 2005.
- Zona, D., Lipson, D. A., Richards, J. H., Phoenix, G. K., Liljedahl, A. K., Ueyama, M., Sturtevant, C. S., and Oechel, W. C.: Delayed responses of an Arctic ecosystem to an extreme summer: impacts on net ecosystem exchange and vegetation functioning, *Biogeosciences*, 11, 5877–5888, <https://doi.org/10.5194/bg-11-5877-2014>, 2014.
- Zona, D., Gioli, B., Commane, R., Lindaas, J., Wofsy, S. C., Miller, C. E., Dinardo, S. J., Dengel, S., Sweeney, C., Karion, A., Chang, R. Y.-W., Henderson, J. M., Murphy, P. C., Goodrich, J. P., Moreaux, V., Liljedahl, A., Watts, J. D., Kimball, J. S., Lipson, D. A., and Oechel, W. C.: Cold season emissions dominate the Arctic tundra methane budget, *P. Natl. Acad. Sci. USA*, 113, 40–45, <https://doi.org/10.1073/pnas.1516017113>, 2016.