Temperature, moisture and freeze–thaw controls on CO₂ production in soil incubations from northern peatlands

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Peat accumulation in high latitude wetlands represents a natural long-term carbon sink, resulting from the cumulative excess of growing season net ecosystem production over non-growing season (NGS) net mineralization in soils. With high latitudes experiencing warming at a faster pace than the global average, especially during the NGS, a major concern is that enhanced mineralization of soil organic carbon will steadily increase CO₂ emissions from northern peatlands. In this study, we conducted laboratory incubations with soils from boreal and temperate peatlands across Canada. Peat soils were pretreated for different soil moisture levels, and CO₂ production rates were measured at 12 sequential temperatures, covering a range from −10 to +35 °C including one freeze–thaw event. On average, the CO₂ production rates in the boreal peat samples increased more sharply with temperature than in the temperate peat samples. For same temperature, optimum soil moisture levels for CO₂ production were higher in the peat samples from more flooded sites. However, standard reaction kinetics (e.g., Q₁₀ temperature coefficient and Arrhenius equation) failed to account for the apparent lack of temperature dependence of CO₂ production rates measured below 0 °C, and a sudden increase after a freezing event. Thus, we caution against using the simple kinetic expressions to represent the CO₂ emissions from northern peatlands, especially regarding the long NGS period with multiple soil freeze and thaw events.

Surface air temperatures in northern high latitude regions are increasing two to six times faster than the global average¹, a trend also observed in Canada¹. Climate models project that this rate of warming will continue through the twenty-first century, with the greatest warming occurring during the fall, winter and spring period, which is generally non-growing season (hereafter ‘NGS’). Given that microbially mediated soil organic carbon (SOC) mineralization (or heterotrophic soil respiration) responds positively to increasing temperatures, warming during the NGS could substantially alter the carbon balance of high latitude soils. For example, arctic tundra ecosystems might already be shifting from annual carbon sinks to sources due to increasing NGS carbon emissions⁴–⁷. The extent to which such shifts apply to a wider spectrum of cold region soil carbon pools remains uncertain, however⁸,⁹.

In addition to the temperature-dependent slowing down of microbial metabolic activity, soil organic matter decomposition during the NGS can also be inhibited by soil freezing, which limits the supply of molecular oxygen (O₂) and the availability of liquid water, even if there are sufficient organic substrates remaining from the growing season. Especially where the soil is poorly drained, a portion of the growing-season plant CO₂ fixation is therefore not mineralized¹⁰. This leads to the accumulation of organic peat layers that can be preserved for up to several millennia¹¹. The resulting northern peatlands have played a substantial role in storing atmospheric CO₂.

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since the last deglaciation\textsuperscript{12,13}. Nevertheless, recent rapid climate warming, as well as direct human encroachment of peatlands, threaten to release this globally important carbon pool into the atmosphere. While Canada has the second largest area coverage of peatland\textsuperscript{14}, much of its southern peatlands have already been lost to agriculture and impacted by other anthropogenic disturbances\textsuperscript{15}. In Canada’s north (i.e., above 50°N), ongoing ecological and biogeochemical changes driven by climate warming are affecting the carbon storage potential of remaining natural peatlands.

Recent advances in data synthesis and modeling have highlighted the sensitivity of global soil respiration to climate change\textsuperscript{16}, with significant implications for NSG soil carbon emissions in cold regions\textsuperscript{8}. Nonetheless, the parameterization and calibration of process models used to simulate future trajectories of soil carbon–climate feedbacks remain poorly constrained when it comes to winter season carbon losses\textsuperscript{8}. For example, a new version of the Canadian Model for Peatlands (CaMP v2.0) has been developed for inclusion into the national carbon budget estimations\textsuperscript{17}. However, the model integrates soil carbon emissions over annual time steps and, hence, does not account for seasonal variations in peat decomposition kinetics, which in turn produces unaccountable uncertainties in model predictions under changing climate conditions\textsuperscript{14,18–20}.

One way to improve the representation of NGS soil carbon dynamics, is to incorporate variable reaction parameters that reflect the seasonal changes of key soil environmental variables, in particular soil temperature and moisture\textsuperscript{21}. For example, the temperature sensitivity coefficient ($Q_{10}$) has been widely used as a simple numerical parameter to correct rates of soil carbon mineralization and other microbially-controlled transformation processes\textsuperscript{22–25}. The $Q_{10}$ approach assumes an exponential increase in the reaction rate with increasing temperature, and a typical $Q_{10}$ value of 2 implies that the rate doubles with every 10 °C increase in the ambient temperature\textsuperscript{26}. Most land surface and dynamic vegetation models use the constant $Q_{10}$ values, but recent findings suggest that the $Q_{10}$ required to temperature-correct CO$_2$ emissions may be higher (i.e., $Q_{10} > 2$) in cold climate regions\textsuperscript{26–29} likely because of the interplay of multiple factors affecting soil physiochemical properties (e.g., those accompanying pore water freeze and thaw) during the NGS\textsuperscript{30,36}.

To date, a considerable amount of soil carbon $Q_{10}$ values have been reported from field and laboratory studies, thus enabling global syntheses\textsuperscript{22,24}. However, a relatively limited number of studies have focused on the warming response of organic-rich wetland and peatland soils despite their disproportionately large share of the global SOC stock\textsuperscript{10,32}, and considering the rapid NGS changes in high latitudes\textsuperscript{4,44}. In addition to the effect on the temperature regime of northern peatland soils, rapid climate warming is also driving changes in landscape hydrology and, thus, in the moisture status of the soils\textsuperscript{21,33}. Thus, latitudinal, and seasonal variations in soil temperature and moisture should be considered when estimating SOC mineralization rates. Without consideration of such variations, projections of future annual peatland CO$_2$ emissions will be fraught with uncertainty, especially at higher latitudes where the impacts of climate change are expected to be most pronounced in the winter and shoulder seasons (fall and spring) compared to summer\textsuperscript{4}.

Because it is challenging to have regular or continuous access to multiple remote peatland locations in winter-time to perform in situ flux measurements, laboratory experiments with field samples can help with the calibration of model parameters (for example, $Q_{10}$ values as discussed above). The lab environment allows one to perform measurements of CO$_2$ production rates under simulated NGS conditions by controlling key environmental factors, such as temperature, moisture content and freezing-induced changes in the accessibility of organic substrates\textsuperscript{30,34–37}.

In this study, we conducted laboratory incubations with peat samples of different depth intervals retrieved at seven Canadian peatland locations in two climate zones (boreal and temperate). We measured CO$_2$ release fluxes under variable moisture levels across a temperature sequence ranging between −10 and +35 °C. The resulting CO$_2$ production rates were fitted to equations describing the observed moisture and temperature trends. Our objectives were to (1) delineate systematic differences in the $Q_{10}$ of CO$_2$ production between sites, and relate them to the ecoclimatic, sample depth and moisture content, (2) identify the optimum moisture levels for CO$_2$ production, and (3) assess the impact of a soil freezing event on the CO$_2$ production during the event and following thaw. We also explored the use of the Macromolecular Rate Theory (MMRT; see Alster et al.\textsuperscript{38} and references therein) as a general framework to describe the measured CO$_2$ production rates.

Materials and methods

Field peat sampling. Peat samples were collected from three peatland sites in “Eastern Cool Temperate Forest” ('temperate' hereafter) and four sites in “Boreal Forest & Woodland” ('boreal' hereafter) biogeoclimatic vegetation zones ('ecoclimate' hereafter), based on the classification of Baldwin et al.\textsuperscript{39} (Vegetation Zones of Canada: a Biogeoclimatic Perspective—Level 1; see Table 1 and Fig. 1). We chose this recent land classification map over the other commonly used national terrestrial ecoczones classification\textsuperscript{14} because it closely parallels the Köppen-Geiger climate classes\textsuperscript{40} for the sampling sites (see Supplementary Figure S1). At each field location, peat samples were collected from depth intervals 0–10, 10–20 and 20–30 cm with a wide-toothed saw. The top 5 cm was removed to minimize large litter debris. All the samples were collected wet, immediately stored in a peat chamber (Mandel Scientific Anaerobic Chamber, AC11-074) and remained uncovered at room temperature (25 °C) under anoxic atmosphere for two weeks while being hand-mixed at regular daily intervals. Next, the total porosity, bulk density and moisture content of the peat samples were determined gravimetrically from the saturated mass, oven-dried mass (2–10 g wet soil at 80 °C for 24 h) and original volume of the sample, following the methods of Gardner\textsuperscript{41}.

Laboratory incubations and CO$_2$ production rates. After the two weeks of chamber-drying, the peat samples were homogenized and divided into five subsamples of approximately equal mass and volume. These
| Site                          | Latitude (°N) | Longitude (°W) | Peatland type | Ecoclimate* | MAT† (°C) | Description‡                                                                                     | Depth (cm) | BD (g cm⁻³) | OC§ (%) |
|------------------------------|---------------|----------------|---------------|-------------|-----------|-------------------------------------------------------------------------------------------------|------------|-------------|---------|
| Old Crow Flats, Yukon        | 68.1148       | 140.05085      | Sphagnum tundra | Boreal     | − 8.27    | Open vegetation cover with Sphagnum moss, Eriophorum tussocks, Ledum decumbens and Rubus chamaemorus. Samples were collected from moss cover | 0–10      | 0.055       | 46.4    |
| Blackstone Uplands, Yukon    | 64.91937      | 138.28308      | Tussock tundra | Boreal     | − 4.1     | Poorly drained low-grade terrain with red Sphagnum moss, Eriophorum tussocks, and some lichen (e.g., Flavocetraria nivalis). Ground is slightly hummocky and sampling was from the hummock top | 0–10      | 0.067       | 46.1    |
| Churchill, Manitoba          | 58.72247      | 93.8477        | Lichen tundra  | Boreal     | − 6.47    | Polygonal peat plateau. Plateau surface forms slightly raised ground dissected by ice wedge troughs. Lichens cover with some sedges, ericaceous shrubs (Vaccinium vitis-idea, Ledum decumbens) and Rubus chamaemorus | 0–10      | 0.111       | 48.6    |
| James Bay Bog, Ontario       | 52.73733      | 83.97382       | Sphagnum bog    | Boreal     | − 0.45    | Raised bog, hummocky with some lawn. Ground cover dominated by moss, mostly Sphagnum fuscum, but in places up to 50% Cladonia rangiferina, with 5 to 50% vascular plants mostly Rubus Chamaemorus and Chamaedaphne calyculata, and sparse tree cover, mostly Picea mariana | 0–10      | 0.087       | 46.1    |
| Turkey Lakes, Ontario        | 47.04809      | 84.40709       | Hardwood swamp  | Temperate   | 4.73      | Mixed stands of Fraxinus nigra, Thuja occidentalis, Acer rubrum, yellow birch and Larix laricina. Understoreys with the seedlings and saplings of the trees, various ferns, herbs (e.g., Carex pulicaris, Carex triangularis, Impatiens capensis) and a mix of feather and Sphagnum mosses | 0–10      | 0.143       | 44.5    |
| Cartier Treed, Ontario       | 46.3976       | 81.3123        | Treed poor fen  | Temperate   | 4.1       | Pronounced hummock-hollow microtopography with continuous Sphagnum moss understory, moderately dense shrub layer (Rhododendron groenlandicum, Chamaedaphne calyculata), sparse sedges in hollows, and 5–10 m tall, well-spaced/open canopy Picea mariana trees. Samples were collected from hummocks. Moss water pH was ~ 4.4 and water table 80 cm below the surface at time of sampling | 0–10      | 0.132       | 46.2    |
Table 1. Peatland sampling site information. Peat samples collected from each of the 10-cm depth intervals were analyzed for dry bulk density (BD, g cm⁻³) and organic carbon content (OC, %). *According to Vegetation Zones of Canada: a Biogeoclimatic Perspective—Level 1⁴⁹. †Mean annual temperature data from 1981–2010 Climate Normals, Government of Canada (see weather station names in Supplementary Table S2). ‡Turkey Lakes site description from Webster et al.⁴². §Samples are combusted at 600 °C in an elemental analyzer.

| Site                        | Site Latitude (°N) | Site Longitude (°W) | Peatland type | Ecoclimate* | MAT † (°C) | Description‡ | Depth (cm) | BD (g cm⁻³) | OC § (%) |
|-----------------------------|-------------------|---------------------|--------------|-------------|------------|---------------|------------|-------------|----------|
| Cartier Lawn, Ontario       | 46.3777           | 81.312              | Open poor fen| Temperate   | 4.1        | A thick floating mat (>5 m) in a terrestrial-izing pond adjacent to the treed fen site with little microtopography, continuous Sphagnum moss cover, and moderately dense (~35% cover) Eriophorum spissum, sparse low shrubs (Chamaedaphne calyculata, Vaccinium oxycoccos), and small (0.5–1 m) Larix laricina. Pore water pH was ~4.3 and water table 5 cm below the surface at time of sampling | 0–10       | 0.161       | 46.2      |
|                             |                   |                     |              |             |            |               | 10–20      | 0.052       | 46.7      |
|                             |                   |                     |              |             |            |               | 20–30      | 0.100       | 48.8      |

Figure 1. The locations of peatland sites across different vegetation (ecoclimate) zones of Canada (Baldwin et al. 2019) with the nearest weather stations (circles) for which Canadian Climate Normals 1981–2010 data are available (Environment and Climate Change Canada). The map was created using ArcGIS Pro (Esri). The site names correspond to the information given in Table 1.
subsamples were placed in five 500-mL glass mason jars, filling less than half the space. The initial dry masses of peat varied by site and depth, from ~120 to ~600 g (or ~24 to ~120 g per jar), with lower masses for the drier and lower density peat samples (e.g., those from the shallowest depth intervals). One in each of the five-jar series was incubated at the gravimetric moisture content of the peat remaining after the two weeks of chamber-drying. The moisture contents of the peat in the other four jars were adjusted by adding an artificial water solution prepared based on the chemical composition of the soil water collected at the corresponding site (Supplementary Table S1). The moisture contents were calculated as water-filled pore space (%WFPS), based on the measured gravimetric moisture, bulk density and porosity, plus the volume of water solution added to the jar. We aimed at %WFPS ranging from the gravimetric moisture content of the peat up to 100%. Thus, each jar and the contained peat corresponded to a given site, depth interval, and moisture condition (e.g., Blackstone, 0–10 cm, 100% WFPS).

The jars were incubated with their lids closed and, therefore, water loss and changes in soil moisture were assumed to be negligible. The jars were incubated in an environmental chamber (Percival I-41NL XC9) which was cycled through a series of 12 sequential temperature settings of 25, 35, 25, 15, 10, 1, −2, −5, −10, −2, 5 and 25 °C. Note that the incubation temperature range was not restricted for a given sample location/depth to the site-specific in-situ soil temperatures. Rather, we imposed the same temperature range and trajectory (including one freezing event) for all the samples to assess the temperature effect, separate from other site- and soil-specific variables. Specifically, the incubation started at room temperature 25 °C, that is, the same temperature under which the peat samples were dried in the anoxic chamber. Thus, the peat samples adjusted to the aerobic and variable moisture conditions without experiencing a sudden temperature change when transferred out of the anoxic chamber. At the high temperature end, 35 °C is usually near the optimum temperature for soil microbial activity. By extending the temperature to the optimum value, we expected a more robust curve-fitting analysis than for a smaller temperature range. Then, we purposely included repeated measurements at 25 °C to check for potential long-term drifts in the rates of CO2 production due, for example, to carbon substrate depletion or an irreversible shift in the microbial community structure, as commonly observed in long-term incubations. The temperature in the incubation chamber was lowered up to −10 °C, considering that soil temperature at 0.1 m depth in northern permafrost peatland have been shown to drop to as low as −10 °C during the winter.

At each setting, the temperature fluctuations were within ±0.5 °C. The CO2 production rates were determined after 48 h of incubation at a given temperature using an automated multiplexer CO2-flux system (LI-8100 and LI-8150, LI-COR Biosciences, Lincoln, NE, USA). The rates of headspace CO2 concentration increase with time (dCO2/dt, μmol m−2 s−1) were measured following the method of Davidson et al.[81], considering the variable headspace volume of the jars (Note: for more details of the measurement setup and procedure, see Rezanezhad et al.[45]). To compare CO2 production among the different peat subsamples, the rates were normalized to the dry weight of the peat in each jar and expressed in units of μmol g−1 h−1. Overall, our incubation experiment was not aimed at quantifying the in-situ CO2 emissions from northern peatlands, but to delineate the response of CO2 production rates to variable temperature and moisture, as well as to a freeze–thaw event.

Fitting of CO2 production rates. For each jar experiment (corresponding to a given site, depth interval, and moisture content), the CO2 production rates measured across a range of temperatures (RT) were fitted to the following exponential equation[45,46] with temperature (T, °C) as the independent variable:

\[ R_T = \alpha e^{\frac{Q_{10}}{RT}} \]  

(1)

where \( \alpha \) and \( Q_{10} \) are fitting coefficients determined by the method of least squares (‘fitnlm’ function in MATLAB R2020b), and \( R_T \) are the CO2 production rates normalized to the dry weight of the peat. Equation (1) is closely related to the Arrhenius equation, which states that the rate constant, \( k \), of a reaction can be described by:

\[ k = Ae^{-\frac{E_a}{RT}} \]  

(2)

where \( R \) is the universal gas constant, \( E_a \) is the (empirical) activation energy and \( A \) is called the frequency or pre-exponential factor. Assuming that, for a given jar experiment, the pool of organic matter being mineralized at the different temperatures remains the same, \( \alpha \) should be linearly related to \( A \). From Eqs. (1) and (2) it then follows that:

\[ Q_{10} = e^{\frac{10E_a}{RT}} \]  

(3)

which implies that using \( Q_{10} \) as a measure of the temperature sensitivity is equivalent to assuming the CO2 production rate follows the Arrhenius equation. Also, note that \( E_a \) may vary with temperature, although for relatively small temperature ranges, as considered in our experiments, \( E_a \) is usually treated as a constant[44]. Practically, the \( Q_{10} \) value represents the proportional rate increase in CO2 production for a 10 °C increase of temperature (i.e., \( Q_{10} = R_{T+10}/R_T \)) and thus provides a simple metric to express the temperature sensitivity of the soil CO2 production kinetics.

The Arrhenius equation predicts that a reaction rate (here as the CO2 production rate) always increases with increasing temperature. Microbially-mediated reaction processes, however, often exhibit a temperature optimum at which the rate reaches its maximum value. This feature is reproduced by the Macromolecular Rate Theory (MMRT) (e.g.,[47–50]) according to which the temperature dependence of enzyme-catalyzed reactions is given by:
where the temperature $T$ is now expressed in degrees K ($0 \, ^\circ\text{C} = 273.15 \, \text{K}$), $T_0$ (K) is a reference temperature point set at a few degrees below the estimated optimum temperature, $R$ is the universal gas constant, $k_B$ is Boltzmann’s constant, and $h$ is Planck’s constant.

Equation (4) has three fitting parameters: the changes in the standard enthalpy ($\Delta^\circ\Delta H_{T_0}$), entropy ($\Delta^\circ\Delta S_{T_0}$), and heat capacity ($\Delta^\circ\Delta C_p$) for the activation reaction (i.e., the formation of the enzyme–substrate transition state), with the standard Gibbs energy of activation then given by: $\Delta^\circ\Delta G = \Delta^\circ\Delta H - T \Delta^\circ\Delta S$. Note that for a heat capacity change of activation equal to zero, $\Delta^\circ\Delta C_p = 0$, MMRT predicts an exponential temperature dependence of the CO$_2$ production rate that is equivalent to that of the Arrhenius equation. Physically plausible deviations from this exponential dependence should yield values $\Delta^\circ\Delta C_p < 0$. For further details on the application of MMRT to microbial reaction processes, see Alster et al. and references therein. In this study, we attempted to find a general relationship between the CO$_2$ production rate ($k$) and temperature ($T$) using Eq. (2) and (4) with the fitting parameters determined by the method of least squares (‘fitnlm’ function in MATLAB R2020b).

Using the measurements at 25 $^\circ$C across the range of experimental moistures, the CO$_2$ production rates ($R_{25 \, ^\circ\text{C}}$) were fitted to a second-order polynomial equation (e.g., $R_{25 \, ^\circ\text{C}} = a x^2 + b x + c$)

$$R_{25 \, ^\circ\text{C}} = a x^2 + b x + c$$

where $x$ is the moisture content (% WFPS). Equation (5) was fitted to the measured rates using the ‘polyfit’ function in MATLAB R2020b. Note that $R_{25 \, ^\circ\text{C}}$ reaches its maximum value at $x = -b/2a$ (% WFPS), which therefore provides a measure of the optimum moisture condition for CO$_2$ production.

**Statistical analyses.** The statistical analyses were performed in R version 4.0.3 using the ‘rpart,’ ‘ggplot2,’ and ‘stats’ packages. A decision tree learning approach (function ‘rpart’) was used to identify the relative importance of ecoclimate, peat depth, and soil moisture content in predicting the $Q_{10}$ values. The $Q_{10}$ values were then summarized for each site and ecoclimate in box-and-whisker plots (function ‘ggplot’). One-way ANOVA (function ‘aov’) was performed to examine whether the $Q_{10}$ variance using the mean site air temperature was chosen as a simple local climate proxy. Additionally, the mean annual air temperature range, calculated as the difference between the July (warmest) and January (coldest) mean temperatures, was also considered as a possible explanatory variable.

**Results**

**Temperature dependence of CO$_2$ production rates.** For all sites, depth intervals and moisture contents, the CO$_2$ production rates increased exponentially with temperature (Fig. 2 and Supplementary Figures S2-6). The $Q_{10}$ distributions extracted from the exponential temperature fits to Eq. (1) are summarized in Fig. 3 for the individual sampling sites (panel a) and grouped according to ecoclimate zone (panel b). Mean $Q_{10}$ values differed significantly between the seven sites [$F(6, 88) = 14.18$, $p < 0.001$] (Supplementary Table S2), although for the boreal ecoclimate zone the intra-site variance exceeded the differences in mean $Q_{10}$ among the four sites [$F(3, 46) = 2.01$, $p = 0.126$]. While the differences between the Cartier Lawn (temperate) and James Bay Bog (boreal) rates were not statistically significant [$t(21) = 2.01$, $p = 0.057$], the mean James Bay Bog $Q_{10}$ value was distinctly higher than that of Cartier Lawn (Fig. 3a). Thus, when the data from the sites were grouped according to ecoclimate zone (Fig. 3b), the difference between the two ecoclimate zones was significant [$t(83) = 7.89$, $p < 0.001$], with the boreal peat samples having a higher average $Q_{10}$ value (mean = 2.45, SD = 0.41) than the temperate peat samples (mean = 1.91, SD = 0.25).

The mean $Q_{10}$ values of the sites were inversely correlated to air temperature (for temperature records covering the 1981–2010 period). This was true whether considering monthly or yearly averaged site temperatures (Supplementary Figure S7). Thus, colder sites yielded higher mean $Q_{10}$ values than warmer sites. Using simple linear regressions, the differences in the annual mean air temperature explained 74% of the observed variations in mean $Q_{10}$ values [$R^2 = 0.74$, $F(1, 5) = 14.12$, $p = 0.013$]. In addition, the site mean $Q_{10}$ values correlated positively with the annual temperature range, expressed as the difference between the coldest and the warmest month [$R^2 = 0.78$, $F(1, 5) = 18.01$, $p = 0.008$] (Fig. 3c). Note that, without the Old Crow Flats site, an even stronger linear trend emerged [$R^2 = 0.95$, $F(1, 4) = 82.13$, $p < 0.001$].

**Moisture dependence of CO$_2$ production rates.** The rates in the 25 $^\circ$C incubations systematically exhibited downward concave trends with respect to moisture content (% WFPS) (Fig. 2 and Supplementary Figures S2-6). The optimum moisture contents fell mostly in the range 60–90% as shown in Fig. 3d. For all the sites, except the Turkey Lakes swamp, the optimum WFPS increased between the near surface depth interval (0–10 cm) and the depth interval below (10–20 cm), although this trend did not continue when moving to the deepest depth interval (20–30 cm). The optimum WFPS in the top layer varied notably from site to site, but without a systematic differentiation between boreal and temperate sites (Fig. 3d). Overall, the variability in $Q_{10}$
Figure 2. Examples of CO₂ production rates measured for peat samples from one boreal (James Bay Bog, a) and one temperate site (Cartier Lawn, b) at different moisture contents (%WFPS; see Table 1). The rates are fitted with Eq. (1) for temperature and Eq. (5) for the moisture dependence. The effect of moisture variation at a fixed incubation temperature (25 °C) is shown in the first leftmost column; the optimum moisture levels for maximum CO₂ production are indicated on the panels. The effects of temperature variations on the CO₂ production at varying peat moisture contents are shown in the next five columns (in increasing order of %WFPS) with the fitted $Q_{10}$ values. The results for the other sites are presented in Supplementary Figures S2 to S6.
values was related in descending order to the variations in the following variables: ecoclimate zone > soil moisture > depth interval (Figure S8).

**CO₂ production rates at low temperatures (−10 °C ≤ T ≤ 10 °C).** Because the ecoclimate site location emerged as the variable with the highest explanatory power, the CO₂ production rate data from all sites in each of the ecoclimate zones were binned together. The two resulting ecoclimate datasets are shown in Fig. 4, arranged from left to right according to the sequence of temperature steps imposed in the incubations. For both ecoclimate zones, the CO₂ production rates at 25 °C measured during the first and third temperature steps (that is, before the freezing event), as well as the 25 °C rates measured during the last temperature step (that is, after

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**Figure 3.** Boxplots of Q₁₀ values calculated from incubations at different moisture contents and peat depths grouped by peatland site (a) and by ecoclimate zone (b). Boxes extend from the first to the third quartile with the inside horizontal line corresponding to the second quartile (median). Vertical lines extend to the 1.5 interquartile range of each box. The red circles (in panels a and b) indicate mean (i.e., average) Q₁₀ values. The latter are used in the linear regression analysis against the site air temperatures and the annual temperature range, i.e., the difference between coldest and warmest month (panel c; for monthly results, see Supplementary Figure S7). Dashed fitted line in plot (c) represents the linear trend without the Old Crow Flats site data. The bar graphs (d) compare the optimum moisture levels for CO₂ production by sampling depth and site.
the freezing event) were not significantly different from one another according to a one-way ANOVA analysis of temperate \( F(2,132) = 2.282, p = 0.106 \) and boreal \( F(2,147) = 0.485, p = 0.617 \) rates.

The inset of Fig. 4 shows the data for the low temperature range (≤ 10 °C) that encompasses the period of freezing temperatures (i.e., the sequence of − 2, − 5, − 10 and − 2 °C steps). As can be seen, measurable CO₂ production rates were detectable even during the sub-zero conditions. The mean rates were higher for the temperate than boreal locations, apart from the − 10 °C rates. Interestingly, the expected decreasing trend of the rates with decreasing temperature was not observed. In fact, for each ecoclimate zone, the mean rates remained relatively constant over the − 2 to − 10 °C range, with values comprised between 0.047 and 0.053 µmol g⁻¹ h⁻¹ for the temperate and between 0.024 and 0.040 µmol g⁻¹ h⁻¹ for the boreal peat samples. Furthermore, following thaw, the mean CO₂ production rates observed at 5 °C were similar (temperate samples) or even higher (boreal samples) than those measured at 10 °C before freezing.

**CO₂ production rates: model fits.** The CO₂ production rates grouped according to ecoclimate zones were fitted to both the Arrhenius equation and the MMRT models. Fits were obtained separately for rates measured before freezing (i.e., for temperatures ranging between 1 and 35 °C) and after the start of freezing (i.e., for temperatures ranging from − 10 to 25 °C). The fitted curves and the resulting model parameter values can be found on Fig. 5, together with the observed mean CO₂ production rates. Note that the model fitting was done using all the available rate data, not just the mean rates.

For data above 10 °C, the Arrhenius fitted curves closely matched the mean CO₂ production rates, both before and after the freezing event. For both ecoclimate groups, the Arrhenius fits yielded activation energies (\( E_a \)) after freezing that were 10–12% lower than before freezing. At temperatures ≤ 10 °C, the Arrhenius equation did not reproduce the temperature trends of the mean CO₂ rates. In particular, the Arrhenius equation did not account for the near-constant rates in the + 1 to − 10 °C range, or for the differences between rates measured before and after the onset of freezing.

The MMRT model yielded better fits to the average CO₂ production rates measured at the lower temperatures (≤ 10 °C), but only if the before and after freezing data were fitted separately. For both ecoclimates, the fits yielded lower optimum temperatures and more negative \( \Delta T \Delta C_p \) values for the rate data collected after the onset of freezing. As a result, the shape of the fitted MMRT curves for the data collected during and after the freezing event deviated significantly from the simple exponential, or Arrhenius-type, temperature dependence.

**Discussion**

**CO₂ production kinetics: ecoclimate zones.** On average, the boreal sites exhibit higher \( Q_{10} \) values for CO₂ production than their temperate counterparts (Fig. 3a-b) although, within any given site, \( Q_{10} \) values vary considerably with changes in moisture content and peat depth (Fig. 2). The strong dependence of \( Q_{10} \) on the
ecoclimate zone is in line with previous work that reports higher temperature sensitivities of the decomposition of soil organic matter under colder climate conditions. Some studies have attributed observed regional variability more to soil types and land use. In the present study, we attempted to limit the influence of these variables by only including soils from undisturbed peatlands.

The annual air temperature range (i.e., the difference between the coldest and the warmest month) explains most of the spread in the mean $Q_{10}$ values between the seven sites (Fig. 3c). Possibly, this reflects an adaptive response of the soil heterotrophic communities to the yearly temperature range. The temperature-dependent microbial production of different extracellular enzymes can, for instance, modulate the overall temperature sensitivity of soil respiration. Soil temperatures, however, may diverge significantly from the air temperature, especially in the colder, sub-arctic regions due to the insulating effects of snow and vegetation cover. Whether this may help explain the apparent departure of the mean $Q_{10}$ value for the Old Crow Flats site from the trend line in Fig. 3c remains to be seen.

Tropical peat soils have $Q_{10}$ values (~ 1.3–1.8) that generally fall below the mean $Q_{10}$ values observed here, thus suggesting a broad latitudinal trend that could help attenuate the response of global peatland CO2 emissions to ongoing and future climate warming. According to climate model projections, by the end of the century most peatlands may have transitioned into a warmer climate zone (Supplementary Figure S1). If climate adaptation reduces the temperature sensitivity of peatland CO2 production (e.g., by lowering current $Q_{10}$ values of boreal peat soils to temperate values), peat decomposition rates will still increase but the slope of the increase would be less steep than anticipated without thermal adaptation. Further research is required, however, to determine how long it would take for thermal adaptation to readjust latitudinal gradients of temperature sensitivities under future climate conditions and, thus, to more realistically account for the feedback to climate of peat soil carbon cycling.

A complete mechanistic understanding of the thermal adaptation of soil microbes, and hence CO2 production kinetics, remains elusive. Nonetheless, empirical model parameter values obtained under relevant environmental conditions can still significantly improve regionalized models of carbon emissions from peatlands. For example, the most recent version of the Canadian peatland model (CaMP v2.0) incorporates variable SOC decay rate coefficients for different peatland ecological types but does not explicitly consider differences due to ecoclimate. Our results show that the temperature sensitivity of CO2 production may vary significantly and, possibly, systematically, depending on ecoclimate location. This finding therefore implies that including ecoclimate-specific temperature sensitivities of peat mineralization in models such as CaMP could significantly reduce parameter uncertainties when assessing the contribution of peat decomposition to continental-scale CO2 budgets.

Figure 5. The grouped CO2 production rates for boreal (a) and temperate (b) ecoclimate (as in Fig. 4). The data points and fitted lines correspond to one of the two cases: before or after the start of sub-zero temperatures (i.e., 'before freezing': from the first temperature setting at 25 °C to 1 °C; and 'after freezing': from the −2 °C to the last 25 °C setting; see Fig. 4 x-axis for the incubation temperature steps). Data are fitted to both the Arrhenius and MMRT models (see Methods). The light symbols show all the rate data collected at a given temperature, the darker symbols are the mean rates.
CO₂ production kinetics: soil moisture. The effect of soil moisture was evaluated using the CO₂ production rates at 25 °C for which we have the most complete set of replicate measurements. The concave down shapes of the moisture response curves (Fig. 2) are consistent with other studies. However, the optimum WFPS values obtained here for the peat soils (Fig. 3d) tend to be higher than the ~60% moisture levels typically observed for more upland and mineral soils (e.g., 37,43,63–66).

Optimum moisture levels for soil respiration emerge from complex interactions among physiochemical (e.g., pore water distribution, oxygen availability, and solute diffusion47,68) and biological factors (e.g., microbial enzyme activity regulation51). The general shift to higher optimum moisture levels in peat soils is likely, at least in part, due to their high organic matter content and complex (multi-domain) pore structure compared to non-peat soils51,69,70. Still, the organic carbon concentrations of our peat samples fall in a fairly narrow range (Table 1), while the optimum WFPS levels at 25 °C vary between 59 and 96% (Fig. 2 and Supplementary Figures S2-6).

The variations in the optimum WFPS for CO₂ production appear to be related to in situ water table conditions, as inferred from the peatland type and microtopography at the sampling location (Table 1). For example, the peat soils from the Blackstone and Cartier Treed sites were collected on hummocks, where the water table is relatively deep below the surface and the top peat often drier than in sunken areas of the landscape, such as hollows51. Similarly, prior to our sampling, the James Bay Bog site had been experiencing a lowering of the water table due to nearby mining operations (e.g., Eriophorum) covers are characterized by shallow water tables and wet conditions52. By contrast, the soils in the permafrost tundra with fen type vegetation (e.g., Sphagnum) exhibit lower moisture contents than those in the peatlands53. There is likely a correlation between the water table depth and the measured optimum WFPS may reflect an adaptation of peatland microbial communities to the prevailing in situ moisture conditions34,55.

Rewetting of peatlands has been proposed as a management strategy to restore their carbon sink function (e.g., 67,68). The experimental results presented here support a systematic decline of the CO₂ production rates at full saturation (i.e., 100% WFPS). However, the complete evaluation of the effect of peatland rewetting on carbon emissions must consider the concomitant increases in CH₄ release due to the decreased availability of oxygen76. Our results further indicate that CO₂ production rates also decrease when the moisture level drops below the optimum WFPS. In the field, drier conditions due to a lowering of the water table would shift the zone of optimum moisture level downwards and, therefore, not necessarily translate in a reduction of the CO₂ emissions at the soil surface.

In addition, there is evidence that, given enough time, the microbial decomposer communities may adapt to drier conditions. Arnold et al.67, for example, showed that in long-term (>300 days) incubations the simulated drainage of wetland soils resulted in cumulatively more organic carbon being mineralized. Recent work has also been exploring the role of moisture on the temperature sensitivity of soil CO₂ production (e.g., 37,49). Together, the results of these studies call for more systematic and data-driven approaches to represent the regulating role of soil moisture in soil carbon process models67,68,77 that, in turn, would improve the reliability of predicted responses of peatland carbon budgets to changes in soil moisture regime due to climate and land use changes.

CO₂ production kinetics: impact of freezing. The production of CO₂ at sub-zero temperatures in both ecoclimate groups exhibits small but measurable rates through the –2 to –10 °C temperature range (Fig. 4). The rates measured here are similar to sub-zero rates in the same temperature range reported in other low-temperature incubation experiments (see, for example, the data compiled by Natali et al.37). Most importantly, the observed trends do not follow the exponential drop of the rates with cooling below 0 °C that is often assumed8. On average, the sub-zero CO₂ production rates are remarkably constant and of comparable magnitude for both ecoclimates (Fig. 5). The lack of a temperature dependence of the average sub-zero rates is in apparent contradiction with standard reaction rate models.

The higher-than-expected CO₂ production rates at temperatures below 0 °C could reflect an enhanced release of biodegradable organic substrates caused by the freezing of the soil that, in the case of peat soils, may be facilitated by the high moisture contents8,78. Such a release of labile organic substrates is supported by the higher rates measured at 5 °C after thawing, compared to the rates measured at 10 °C prior to freezing. Thus, a possible explanation for the near-constant rates in the –2 to –10 °C range is that the enhanced supply of labile organic compounds counterbalances the lowering of temperature below 0 °C. Because more labile compounds have lower activation energies of degradation, they also result in lower temperature sensitivities. This is consistent with the lower fitted $E_a$ and $Q_{10}$ values after the freezing event for both ecoclimate groups (the values are given on Fig. 5).

The divergence of the temperature trends of the CO₂ production rates before and after freezing is not captured by the Arrhenius rate equation, even after adjusting the $E_a$ values (Fig. 5). By contrast, the MMRT model is able to account for the enhanced rates observed after thawing in the 0 to 10 °C range. For both ecoclimate groups, the post-thaw MMRT fits yield more negative $\Delta C_p$ values and a lowering of the optimum temperatures ($T_{\text{opt}}$). These changes are in line with the “Enzyme Rigidity Hypothesis” (see Alster et al.38 for details), according to which cold-adapted enzymes are more sensitive, or “less rigid,” than enzymes adapted to warmer temperatures. Thus, in addition to releasing more labile organic substrates, the cooling to sub-zero temperatures could also have been accompanied by the selective activation of “cold-adapted enzymes” by the microbial communities.

When dynamic enzyme responses modulate the rates of organic matter degradation, the temperature dependence is expected to depart from the Arrhenius equation and exhibit an optimum biological temperature ($T^{*}_{\text{opt}}$)37. However, the predicted post-freezing $T^{*}_{\text{opt}}$ values, 23.4 and 29.6 °C, would need to be verified by increasing the experimental temperature range well above 25 °C (which was the final temperature setting in our experiments), and including more frequent measurement intervals38. Nonetheless, the MMRT provides a more flexible approach that could be useful in modeling the in-field observed seasonal hysteresis in the temperature trends of CO₂ emissions (e.g., 20,30,58), as well as CH₄ emissions82–84.
Implications for CO₂ emissions during the NGS. The failure of the Arrhenius and MMRT models to reproduce the observed temperature trends of the sub-zero CO₂ production rates is not entirely surprising. Both models only account for the temperature dependence of molecular reaction rates. In addition to increasing the supply of labile substrates, freezing may have other consequences for the cryogenic soil environment, including changes in the distribution of unfrozen water and the transport pathways of solutes and gases. Together, these changes modulate the measured rates at which CO₂ is released from the soil. Thus, it may be advisable not to extend the use of reaction rate models to simulate sub-zero CO₂ production rates in soils, or those portions of the soils, that undergo seasonal freezing. To properly represent frozen peat soils, future experimental work should consider measuring CO₂ production over longer time spans than in the present study, weeks to months rather than the 48 h per temperature step here, and extend the temperature range covered.

The experimental results imply that freezing impacts the rates of CO₂ production upon thawing; this is especially evident from the 5 °C post-thawing data (Figs. 4 and 5). Typically, the increase in CO₂ emission fluxes during spring thaw observed in the field are attributed to a combination of the pulsed release of trapped gas that accumulated under the snow and ice layer during winter and de novo microbial CO₂ production due to the sudden increase in temperature and oxygen supply upon snowmelt. Our results suggest that the enhanced access to relatively labile organic compounds mobilized during freezing may additionally contribute to the observed post-thaw CO₂ pulses.

It will be important to further determine whether the post-thaw CO₂ rate enhancement observed in our experiments is short-lived, and possibly insignificant in the overall NGS CO₂ emission budget, or whether it could help sustain CO₂ production throughout the shoulder season. The importance of enhanced carbon loss from peatlands induced by encroaching permafrost thaw is now well recognized. However, the role of shallow freeze–thaw cycles on NGS CO₂ emissions need to be better quantified and the responsible mechanisms understood. At higher latitudes, freeze–thaw cycles are expected become more frequent in the springtime, because trends in snow cover (and its temperature insulating effect) are decreasing.

Summary and conclusions

This study highlights the statistically significant variations in the temperature sensitivity of peat soil CO₂ production rates between the cold-temperate and boreal ecoclimate zones. Given the higher Q₁₀ values, in combination with spatial and seasonal patterns of global warming, boreal peatlands may increase future NGS CO₂ losses to a larger degree than temperate peatlands. The variable temperature sensitivities under different climate conditions need to be accounted for when assessing future global trajectories of peatland carbon pool stability.

Peatland soils tend to support high organic matter mineralization activity at higher moisture levels compared to well-drained mineral soils. Our findings suggest that the optimum moisture level for maximum CO₂ production is a function of peatland type (e.g., bog or fen) and microtopography (e.g., samples from hummock or hollow), which, in turn, are related to the relative position of the water table. The variations in optimum moisture content may thus reflect an adaptation of the resident microbial communities to the prevailing moisture conditions.

Freezing of the peat samples causes a shift in the temperature response curve for the CO₂ production rates. This shift is captured by the MMRT model, but not by the conventional Arrhenius equation. Furthermore, the reaction rate models are unable to account for the near-constant CO₂ production rates measured across the sub-zero temperature range (−2 to −10 °C). More research is needed to unravel the various processes and structural properties controlling CO₂ production in frozen peat soils and their response to the changes in temperature and hydrology under climate warming. Only with this improved mechanistic understanding supported by precise rate data will process models be able to grapple with the implications of climate and land use changes for NGS carbon budgets of northern peatlands.

Received: 6 July 2021; Accepted: 19 November 2021
Published online: 01 December 2021

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Acknowledgements
Funding was provided by the Natural Sciences and Engineering Research Council (Advancing Climate Change Science in Canada: ACCPJ 536050-18 and Discovery Grant: RGPIN-2015-03801), and the Canada First Excellence Research Fund under the Global Water Futures (GWF) program (including GWF grant Winter Soil Processes in Transition). The authors thank James Seward, Heather Townsend, Nicole Balliston, Tasha-Leigh Gauthier, and Mason Dominico for their help with peat sampling and Marianne Vandergriendt, Kelly Tran, Mercedes Huynh, and Shuhuan Li for their assistance with laboratory experiments and analyses.

Author contributions
F.R., L.F. and P.V.C. led the experiments and prepared the dataset. F.R., N.B., J.S.P., W.L.Q., P.R. and K.W. collected peat samples for the experiments and contributed to writing the site descriptions. E.B. conducted data analysis and led the preparation of the manuscript. F.R., L.F., S.S. and P.V.C. provided feedback on the initial analysis results. F.R., S.S., N.B., W.L.Q. and K.W. reviewed and provided feedback on the first version of the manuscript, and all authors contributed to improving the manuscript onward. F.R. and P.V.C. reviewed and approved the final version of the manuscript for submission.

Competing interests
The authors declare no competing interests.

Additional information
Supplementary Information The online version contains supplementary material available at https://doi.org/10.1038/s41598-021-02606-3.

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