Mismatch of N release from the permafrost and vegetative uptake opens pathways of increasing nitrous oxide emissions in the high Arctic

Fabrice Lacroix1,2,3 | Sönke Zaehle1 | Silvia Caldararu1 | Jörg Schaller4 | Peter Stimmler4 | David Holl5 | Lars Kutzbach5 | Mathias Göckede1

1Biogeochemical Signals (BSI), Max Planck Institute for Biogeochemistry, Jena, Germany
2Climate and Environmental Physics, University of Bern, Bern, Switzerland
3Oeschger Centre for Climate Change Research, University of Bern, Bern, Switzerland
4Leibniz Centre for Agricultural Landscape Research (ZALF), Müncheberg, Germany
5Institute of Soil Science, Center for Earth System Research and Sustainability (CEN), University Hamburg, Hamburg, Germany

Correspondence
Fabrice Lacroix, Biogeochemical Signals (BSI), Max Planck Institute for Biogeochemistry, Jena, Germany. Email: flacro@bgc-jena.mpg.de

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Abstract
Biogeochemical cycling in permafrost-affected ecosystems remains associated with large uncertainties, which could impact the Earth’s greenhouse gas budget and future climate policies. In particular, increased nutrient availability following permafrost thaw could perturb the greenhouse gas exchange in these systems, an effect largely unexplored until now. Here, we enhance the terrestrial ecosystem model QUINCY (QUantifying Interactions between terrestrial Nutrient CYcles and the climate system), which simulates fully coupled carbon (C), nitrogen (N) and phosphorus (P) cycles in vegetation and soil, with processes relevant in high latitudes (e.g., soil freezing and snow dynamics). In combination with site-level and satellite-based observations, we use the model to investigate impacts of increased nutrient availability from permafrost thawing in comparison to other climate-induced effects and CO₂ fertilization over 1960 to 2018 across the high Arctic. Our simulations show that enhanced availability of nutrients following permafrost thaw account for less than 15% of the total Gross primary productivity increase over the time period, despite simulated N limitation over the high Arctic scale. As an explanation for this weak fertilization effect, observational and model data indicate a mismatch between the timing of peak vegetative growth (week 26–27 of the year, corresponding to the beginning of July) and peak thaw depth (week 32–35, mid-to-late August), resulting in incomplete plant use of nutrients near the permafrost table. The resulting increasing N availability approaching the permafrost table enhances N loss pathways, which leads to rising nitrous oxide (N₂O) emissions in our model. Site-level emission trends of 2 mg N m⁻² year⁻¹ on average over the historical time period could therefore predict an emerging increasing source of N₂O emissions following future permafrost thaw in the high Arctic.

KEYWORDS
carbon, climate, high Arctic, nitrogen, nitrous oxide, permafrost, vegetation
1 | INTRODUCTION

High-latitude permafrost-affected soils cover over one-sixth of the exposed northern hemisphere land surface and store vast amounts of carbon and nitrogen (Hugelius et al., 2013, 2020; Obu, 2021; Palmtag et al., in review). However, large uncertainties remain associated to the past and future perturbation of greenhouse gas fluxes in these regions, which could have an impact on achieving global climate targets (Biskaborn et al., 2019; Bruhwiler et al., 2021; Canadell et al., 2022; Natali et al., 2021; Schädel et al., 2018; Schuur et al., 2015). On the one hand, increased carbon fixation linked to enhanced vegetation growth in these regions has already been reported over the past few decades, with further drastic changes expected in the future (Canadell et al., 2022; Pearson et al., 2013; Winkler et al., 2019). On the other hand, the warming and thawing of frozen soils expose large organic carbon and nitrogen pools, which were conserved for thousands of years, to aerobic or anaerobic degradation, contributing to increased greenhouse gas emissions to the atmosphere (Burke et al., 2017; de Vrese & Brovkin, 2021; McGuire et al., 2018; Schädel et al., 2016; Schuur et al., 2015; Voigt et al., 2020). The feedback could be especially critical in the high Arctic, a region with disproportionate carbon and nitrogen stocks vulnerable to permafrost thaw (Hugelius et al., 2020; Palmtag et al., in review).

Since temperatures in Arctic regions have already increased substantially over the historical time period (Ballinger et al., 2020; Screen & Simmonds, 2010) and are projected to further increase in the future (Cai et al., 2021; Previdi et al., 2020), estimating the degree of permafrost thawing and its feedbacks on vegetation and soil carbon stocks is of high importance in the context of quantifying the greenhouse gas budget of high latitudes (Schuur et al., 2015). In addition to implications of climate warming for the exchange of greenhouse gases CO₂ and CH₄ in high latitudes, recent measurement campaigns have reported evident sources of nitrous oxide (N₂O) to the atmosphere from the high Arctic (Maruschak et al., 2021; Voigt et al., 2020; Wilkerson et al., 2019). Previously, N₂O emissions from the high latitudes were largely neglected in the global greenhouse gas budgets, owing to low nitrogen availability in the upper soil (Voigt et al., 2020). Strong uncertainty still exists with respect to the high-latitude N₂O flux for the present day (Maruschak et al., 2021; Voigt et al., 2020; Wilkerson et al., 2019), as well as in terms of its the future evolution of (Voigt et al., 2017).

The implications that changing climatic conditions and permafrost thaw have for vegetation and microbial dynamics in the high latitudes are still poorly constrained. In particular, the perturbations that the thaw of permafrost may impose on nutrient cycles in these regions are largely unexplored (Beermann et al., 2017; Chadburn et al., 2017). Since Arctic tundra ecosystems are usually strongly nutrient-limited (Finger et al., 2016; Haag, 1974; Natali et al., 2012; Pedersen et al., 2020; Street & Calderaru, 2022), increased nutrient mobilization from permafrost layers could potentially enhance vegetation growth through plant uptake and/or microbial activity. Such fertilization effects have been demonstrated in a variety of Arctic site-level investigations (Blume-Werry et al., 2019; Pedersen et al., 2020; Street et al., 2018; Yang et al., 2021). Large stocks of highly labile nitrogen are reported in the high-Arctic permafrost (Fouché et al., 2020; Hugelius et al., 2020; Salmon et al., 2016; Palmtag et al., in review), which would likely be rapidly degraded to biologically-available compounds upon soil thaw.

The fate of nitrogen (N) mobilized following permafrost thaw remains largely uncertain, despite being a strong limiting element for permafrost ecosystems (Haag, 1974; Natali et al., 2012; Pedersen et al., 2020). Both plants and microbes compete for the sparse N in the shallow active soil layer (Monteux et al., 2020). It is debated to which degree increased N release in the deeper soil, for instance owing to thaw of permafrost, would be taken up by the vegetation, thus retaining newly released N within the biosphere (Beermann et al., 2017; Blume-Werry et al., 2019; Finger et al., 2016; Keuper et al., 2017; Koven et al., 2015; Natali et al., 2012; Norby et al., 2019; Pedersen et al., 2020), used by microbes and potentially outgassed as N₂ and N₂O (Elberling et al., 2010; Voigt et al., 2017, 2020), or exported laterally (Rasmussen et al., 2022; Treat et al., 2016). Phosphorus (P) could also provide an important constraint for plants and microbial activity in high-latitude ecosystems (Salmon et al., 2018; Schaller et al., 2019; Street et al., 2018; Yang et al., 2021; Zhang et al., 2014), thus its increasing availability could impact the degradation of organic matter in the deep soil.

Impacts of climate change on high Arctic terrestrial systems have been assessed in the past using terrestrial land surface models (e.g., de Vrese & Brovkin, 2021; de Vrese et al., 2021; Koven et al., 2015). However, important uncertainties in representing biophysical processes in these regions certainly still exist (Chadburn et al., 2017; Schädel et al., 2018; Yokohata et al., 2020). These approaches have, for one, largely omitted nutrient dynamics in the high latitudes and their impacts on vegetation growth and microbial activity. They have also used either coarse vertical distributions, or even only two layers, to represent vertical soil profiles at the global scale (Chadburn et al., 2017). Both these factors strongly limit the assessment of effects caused by thawing of the permafrost on vegetation growth and microbial activity, and thus of greenhouse gas exchange.

To advance the understanding of the fate of nutrients originating from permafrost thaw in the high Arctic, we enhanced the terrestrial biosphere model QUINCY (QUantifying Interactions between terrestrial Nutrient CYcles and the climate system) with additional modules addressing specific bio-physical features important in the high Arctic. Using this enhanced model version, in addition to site-level observational data, we analysed transient simulations over 1960 to 2018 to quantify the impacts that deepening of the active layer and increased nutrient availability have on vegetation dynamics and greenhouse gas exchange, with a particular focus on the fate of N upon thaw of permafrost and pathways leading to N₂O emissions.

2 | METHODS

We used observational data collected at the site level and performed site-level simulations over the Arctic Tundra for 1901–2018 with the
global terrestrial model QUINCY. We thereby aimed to improve the understanding of impacts caused by nutrients (N and P) mobilized following the thaw of permafrost, both for vegetation growth and greenhouse gas emissions. We implemented additional features in both plant and soil modules of QUINCY to better represent physical soil processes taking place in this cold region (i.e., soil freezing, snow and inundation) but also including specific soil conditions for permafrost-affected soils, as well as alterations to vegetation dynamics as a response to changes in the soil state (i.e., dynamic root depths). Using this extended model version, we performed sets of simulations to isolate effects arising from increased nutrient availability following permafrost thaw, other changes in climate, as well as CO₂ fertilization, while also using site-level observational data to verify our analysis when possible.

2.1 | The QUINCY model

The QUINCY terrestrial biosphere model simulates water and energy exchanges, as well as coupled C, N and P cycles in soil and vegetation (Thum et al., 2019). The model has already been applied and validated within a number of studies investigating nutrient cycles in vegetation and soils (Caldararu et al., 2020, 2022; Yu et al., 2020). It provides improvements compared to other terrestrial biosphere models used in global assessments (e.g., Canadell et al., 2022) through its state-of-the-art representation of vegetation dynamics (Thum et al., 2019; Text S1) and vertically explicit representation of N and P cycling (Caldararu et al., 2020; Thum et al., 2019; Figure 1; Text S2). The model also includes the production of N-containing gases N₂O and N₂ through nitrification and denitrification (Figure 1; Text S2). QUINCY discretises both soil physical parameters and biogeochemical pools in the vertical plane, which is highly relevant for permafrost-affected soils. Omitting or poorly representing nutrient cycles and vertical gradients in permafrost-affected soils have been identified as important reasons for the poor performance of global terrestrial biosphere models in the high latitudes (Chadburn et al., 2017).

The model accounts for five pools of soil organic matter, that is, structural, polymeric, and woody litter, fast and slow degrading soil organic matter, which have dynamic C to N stoichiometries (Table S1; Thum et al., 2019). Organic N is mineralized to ammonium (NH₄⁺), which in turn can be further oxidized to nitrate (NO₃⁻) under aerobic conditions (nitrification), both of which are dependent on temperature and soil moisture conditions (Table S1). Mineralization is constrained to nil at temperatures below zero. During nitrification, N₂O is produced at a fixed ratio of N nitrified. QUINCY computes uptake of NH₄⁺ and NO₃⁻ within the root zone dependently on available C for growth and tissue stoichiometry, and proportional to the current soil N availability and fine root distribution in the model (Table S1). Stored reserve N and dynamic stoichiometry within plants further allow the plants to adapt to fluctuations and low nutrient levels (Caldararu et al., 2020). Under anaerobic soil conditions, NO₃⁻ can undergo denitrification, producing N₂O and N₂ at a fixed ratio, dependently on temperature (Table S1). Aerobic and anaerobic volume fractions of the soil are calculated as in Zaehle et al. (2011).

**FIGURE 1** Conceptual scheme of N fluxes and pools in the QUINCY model extended for high-latitude processes. Note that our both active layer and the permanently frozen part of the soil comprise multiple model layers, and that the difference between rooting depth and permafrost table depth is exaggerated here. Atmospheric deposition inputs were also not considered in this study.
2.2 | Upgraded physical processes for high latitudes: Soil freezing, snow dynamics, organic matter effects, and inundation

To provide an improved representation of permafrost-affected ecosystems, we adapted the physical modules in QUINCY to account for effects of soil freezing and thawing, as well as of snow, on soil layer heat and water exchanges.

We extended the heat budget calculation with latent heat fluxes induced by soil freezing and thawing for every soil layer (Text S3). In similarity to other model implementations (Chadburn et al., 2015; Ekici et al., 2014), we assumed that a supercooled liquid water fraction remained present in the soil at freezing temperatures, which can also be observed in soil moisture measurements (Text S5). Additionally, water content is kept above the computed permanent wilting point in the surface soil layer for plant organs to survive the winter. The implementation of a multi-layer snow module (Text S4) served to represent the effects of snow coverage on the soil heat and water balance, preventing extreme cooling of the soil in the winter, while also delaying liquid water inputs to the soil until the melting season. Here, we used the same five-layer model structure as described in Ekici et al. (2014), with differences explained in Text S4.

We also accounted for the lower heat conductivity and higher specific heat capacity observed in soils with high organic contents, as often found in permafrost-affected soils. Similar to Burke et al. (2017), we used lower heat conductivity constants for the upper 30 cm of unfrozen soil, which is also where the highest soil organic carbon (SOC) contents are also simulated in the model and usually found in observations (Table S3).

We also implemented an inundation scheme to reproduce the retention of water over permanently frozen layers, which leads to saturated conditions in the spring/early summer as is the case in observations (Text S5; Göckede et al., 2019).

2.3 | Upgraded soil biogeochemistry for high latitudes: Initialisation and dynamics

Biogeochemical features in permafrost soils are unique in that organic material below the permafrost table is largely preserved, meaning that the permafrost stores stable pools of carbon and nutrients. Achieving an initial vertical state of the soil adequate for the representation of permafrost gradients is challenging in terrestrial models (Chadburn et al., 2017), and many global terrestrial land surface models only resolve surface and subsurface soil as two layers. Here, we used 18 vertical layers up to a maximum depth of 9.5 meters, leading to layer thicknesses between 10 and 20 cm at the simulated permafrost table.

We initialised total SOC contents and organic soil C:N ratios at our two case study sites Cherskiy and Samoylov using measurements undertaken at these sites (Table S3, Zubrzycki et al., 2013). We used the land cover type estimations of Palmtag et al. (2022, in review) for the other high-Arctic sites. This approach quantifies specific Arctic land cover type carbon and N contents based on 651 soil pedons from 16 study areas. For all sites, we used the averaged C:N ratio over the entire vertical plane. We assumed 5% of soil organic matter to be fast degrading, a conservative estimate with respect to estimates of up to 10% (Hugelius et al., 2020; Strauss et al., 2015). To initialize organic P contents, we used the default organic N:P mole ratio of 55 used in Thum et al. (2019) for the global scale due to the lack of observational constraints.

The SOC contents were initialised to maintain a higher content and reactivity of organic carbon in deeper layers (Figure S1). This is not the case in the standard model, which prescribes exponentially decreasing SOC with depth, thus yielding close to zero contents at permafrost depths. The SOC was prescribed to follow fitted exponential distributions derived for Cherskiy \( (k = -3.88 \text{ cm}^{-2}) \) and Samoylov \( (k = -0.009 \text{ cm}^{-2}) \) in the first 50 cm, and constant past this depth as an approximated permafrost table. The SOC and organic N concentrations were then normalized using observations for the available estimates for 1–300 cm and 1–100 cm made for Cherskiy and Samoylov. Due to lacking data, we used an average of the two vertical distributions to determine the vertical distribution of the other sites, which was then normalized according to SOC contents taken from Palmtag et al. (in review). To preserve reactive SOC at permafrost depths, our simulations were initialized at negative temperatures to prevent its degradation during the model spin-up. We performed a spin-up simulation for the soil active layer to equilibrate to an approximately non-perturbed state with respect to human-caused warming (Figure S1). As a result, total SOC and N slightly deviate from observations that they were initialized for, especially in the top soil (Figure S1).

Due to the changes made to the soil hydrology as a result of soil freezing and inundation (Text S5), vertical leaching of biogeochemical compounds past the permafrost table is essentially inhibited in the model. Within the active layer, however, vertical transport of biogeochemical compounds can happen as a result of drainage, diffusion or due to bio-turbation.

2.4 | Upgraded vegetation dynamics for high latitudes: Timing of growth and frozen-soil-dependent root depth

In our model version, the start of the growing season was delayed by snow cover. The growing degree day count only began when snow coverage was less than the minimum coverage threshold, defined here as 2 cm snow thickness (Table S2).

Additionally, we developed a module that considers plant roots to be adaptive to the seasonal freezing and thawing of the soil. In this scheme, fine root growth is only possible in unfrozen soil layers at every model time step. Instead of using a fixed root distribution, we prescribed the root distribution used in the standard QUINCY model in the upper 30 cm of the soil, which dictates an exponential decrease of root mass with a decay constant of 5.5 m\(^{-1}\), determined...
for grasses by Jackson et al. (1996). This reflects that most roots in permafrost systems are found in shallow soil depths and do decrease with depth there (Blume-Werry et al., 2019; Finger et al., 2016; Iversen et al., 2015). Below 30cm, we did not use a weighed profile, meaning that in these soil layers, root growth would happen at the same rate in all layers as long as they are not frozen. A major consequence of the dynamic root scheme is that nutrients stored within permanently frozen layers are unavailable for uptake by the plants. At the same time, nutrient pools at the bottom of the active layer only gradually become available for root uptake due to slowly increasing thaw depths during the summer.

2.5 Simulation inputs, forcing, and strategy

We performed model simulations from 1901-2018 for 15 tundra sites of continuous permafrost over the high-Arctic region, represented by the C3 grass plant functional type in the model (Figure S2, soil parameters for every site are given in Supplement site_info_arctic_grasslands.csv). The model was driven with meteorological data from the University of East Anglia Climatic Research Unit Japanese Reanalysis (CRU-JRA; Harris, 2019), which is a reanalysis of incoming solar radiation, surface temperature, humidity, precipitation, surface pressure, and surface winds with the goal of providing atmospheric forcing for land surface models. The dataset is available at a spatial resolution of 0.5°x0.5° and a 6-hourly temporal resolution. The atmospheric forcing data for our chosen sites were extracted from the CRU-JRA dataset according to their geographic coordinates, and the data were subsequently transformed to half-hourly timestep intervals using a weather generator (Zaehle & Friend, 2010). Furthermore, we used a vertical soil profile resolution of 18 layers, with increasing width down to a depth of 9.5 m. In our simulations, this set-up proved to be adequate for shallow permafrost table depths of 0.3-1.5 m that we simulated in the high Arctic.

We differentiated between effects of (i) atmospheric CO₂ fertilization, (ii) changes arising from perturbation of the climate, and (iii) increased availability of C, N and P from the permafrost on carbon cycle processes by conducting several sets of simulations for each high-Arctic site (Table 1). These simulations were performed after model spin-up periods of 300 years, which were driven with atmospheric conditions for 1901–1930, and were subsequently checked for important variable drifts. We conducted the first set of simulations (climate + withoutpermafrostCNP, Table 1), considering changes in climate, but initialising carbon and nutrients contents to exponentially decrease with depth, as in the standard model. By doing this, C, N and P contents at depth are close to zero, thus excluding any potential fertilization effect linked to a deepening active layer.

The second set of simulations was again driven by changing climate (climate), but this time also considering the release of carbon and nutrient pools from previously permanently frozen layers, that is, release from the permafrost. The third set of simulations additionally considered the impact of increasing atmospheric CO₂ levels on vegetation dynamics and carbon cycle processes (climate + CO₂). To isolate the effects of increasing nutrients/carbon availability following permafrost thaw, other climate effects and CO₂ fertilization, differences between scenario simulation trends were computed.

Furthermore, we performed a simulation without the model implementations for high latitudes (withoutHighLat). To further elucidate effects caused by nutrient limitation, we conducted an additional simulation with the same atmospheric forcing conditions as climate + CO₂, but fixed soil N and P concentrations at fivefold their prescribed levels (Nutrients5x), values chosen during test simulations with the goal of largely alleviating nutrient-limiting conditions.

2.6 Case study sites and observation datasets

We examined detailed results, including the evolution of biogeochemical fluxes, for two Siberian tundra sites with underlying continuous permafrost: Cherskiy (CHE) and Samoylov (SAM). The sites were chosen due to availability of datasets for soil conditions (soil temperature, snow depths, soil moisture, SOCC, and soil N), as well as eddy-covariance-based Gross primary productivity (GPP) estimates, to validate our results.

The Cherskiy site is located in Northeastern Siberia, in proximity to the Kolyma River. Mean annual temperature is around −11°C (Göckede et al., 2017). Underneath an organic peat layer of around 20cm, silty loam can be found up to the permafrost table depth of

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<td>Simulation/driver</td>
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around 60 cm. Vegetation near the site is dominated by tussock-forming species (Kittler et al., 2016). Soil temperature, snow depths, soil moisture and eddy-covariance fluxes for GPP were obtained as described in Kittler et al. (2016) and Göckede et al. (2017). Vertical profiles of SOC and total N were measured at the site and are provided in Table S3.

The Samoylov site is a river terrasse situated on an island of the Lena River delta. While the organic layer near the site can be up to 30 cm in thickness, it is typically less than 10 cm on the dry polygon rims due to differing hydrological conditions (Boike et al., 2018, 2019). The active layer is found at around 50 cm depth in mid-August, and the permafrost depth extends to over 400 cm below the surface. Vegetation coverage is dominated by sedges and mosses, dwarf shrubs and forbs (Kutzbach et al., 2004, 2007). We used soil temperature and snow depth measurements at the site reported in Boike et al. (2019) for the validation of our model and estimations of thaw depths. We constructed eddy-covariance GPP estimates for 2014–2017 based on CO₂ exchange product of Holl et al. (2019), applying a refined version of the bulk model approach of Runkle et al. (2013; Text S6). SOC and total soil N content measurements were taken from Zubrzycki et al. (2013).

In addition to the site-level data, we also compared modelled GPP to the MOD17A2H MODIS/Terra Gross Primary Productivity product (Running et al., 2015), which was extracted for the geographic coordinates of the individual sites at the original 500 m resolution.

### 2.7 Analysis strategy

We analysed trends in fluxes and temperatures for the time frame 1960–2018 via linear regression, computing statistical significance through Student’s t-tests, and multiplying the yearly trend with the entire analysis timespan. This timeframe was chosen due to the substantial increase in observed temperature past the 1960s. We also calculated weekly averages for the time period 1950–1970, which we assumed to be largely unperturbed, and the perturbed present-day timeframe 1998–2018, to compare the seasonal dynamics of the two time slices.

### 3 RESULTS

#### 3.1 Soil temperature and active layer depths

Simulated seasonal soil temperatures in the model are in good agreement with observed temperatures over a multi-year average when including the model extensions for high-latitude processes (Figure 2). The model reproduces seasonal temperature fluctuations in the top soil (6–8 cm, Figure 2a,b) and in the deeper soil (60–64 cm, Figure 2e,f), where soil temperature remains near the freezing point for an extended time period in the summer. Our simulations still yield positive summertime biases of 3–6°C for 6–8 cm and 28–32 cm at both Cherskiy and Samoylov sites, while simulated winter soil temperatures are too high at the Samoylov site.

Simulated mean maximum thaw depths of 0.7 and 0.6 m adequately reproduce averaged observed maximum thaw depths of 0.5–0.7 m and 0.4–0.6 m at Cherskiy and Samoylov, respectively (Boike et al., 2019; Göckede et al., 2017). Snow seasonality is reasonably represented over the year, although snow depth is overestimated at Samoylov and underestimated at Cherskiy with respect to observational data (Figure S4). The model struggles at estimating magnitudes of water content fluctuations and field saturation capacity; however, it reproduces low soil moisture in the winter period and majorly saturated conditions in the deeper soil over the entire summer (Figure S3).

#### 3.2 Mismatch of GPP and soil thaw peaks

Simulated annual GPP at Cherskiy and Samoylov amount to 353 and 107 g C m⁻² year⁻¹ in the climate + CO₂ simulation over a multiple year mean (Figure 3a,b), respectively. For the same sites, our eddy-covariance derived products provide annual mean GPP estimates of 321 and 134 g C m⁻² year⁻¹, and the MODIS-based product quantifies annual mean GPP of 266 and 125 g C m⁻² year⁻¹ (Figure 3a,b; Figure S5). Over the pan-Arctic scale, we compute an average GPP of 307 g C m⁻² year⁻¹ and find good agreement of modelled GPP with the MODIS dataset on the individual site level (r² = .72, Figure S6), although it should be noted that our model performs better than the MODIS product when comparing to eddy-covariance based magnitude and seasonal variations of GPP for the two sites we analysed (Figure S5). Modelled annual GPP in climate + CO₂ is 3-fold lower on average than in the simulations Nutrients5x (Figure S7), emphasizing that soil nutrient levels are limiting GPP in climate + CO₂ and provide a strong constraint for reproducing observed GPP at the site level.

Our model performs well overall in terms of reproducing peak and seasonal variation of site-level GPP over multi-year means for both case-study sites with respect to the eddy-covariance estimates (Figure 3a,b). The weekly averaged simulated GPP peak is reached during week 27 on average at both case study sites (end of June), at magnitudes of 4.9 μmol m⁻² s⁻¹ (Cherskiy) and 1.6 μmol m⁻² s⁻¹ (Samoylov). Over all Arctic sites, the GPP peak is also reached at week 27 on average, at an Arctic-average GPP level peak of 3.7 μmol m⁻² s⁻¹. After reaching this peak, GPP at both case study sites sharply declines in both the eddy-covariance products observations and in the model (Figure 3a,b). The seasonality found in the monthly averaged MODIS dataset also suggests a GPP peak early in the summer (Figure S5). Our model also indicates that GPP decreases past the peak owing to depletion of soil N, since modelled soil N availability cannot meet plant N demand at week 26 at Cherskiy and at week 27 at Samoylov over the multi-year average (Figure S7).

Averaged over all high-Arctic sites, plant N demand exceeds N soil availability at week 26 (Figure S8). This is in concurrence with declines of both simulated and observed GPP.
Both observational data and model simulations show seasonal thawing maxima succeeds the GPP maxima by several weeks at both case study sites, as well as for the Arctic average (Figure 3a,c). In our multi-year averaged observational data, where we derived the soil thaw depth through interpolation of the temperature between the coarsely measured layers, we find maximum thaw depths of 0.7 and 0.6 m at weeks 34 and 33 and at Cherskiy and Samoylov, respectively. The model reflects this pattern, simulating the same maximum thaw depth at weeks 35 and 34, respectively. The timing of maxima in mineralization, or N release, in the deep soil (50–100 cm), occurs in concurrence with or closely following the thaw depth maxima in the model (Figure 3a,c).
3.3 | Perturbation of active layer depth and increased deep-soil nutrient availability

Our observation-based forcing dataset prescribes rising atmospheric CO₂ levels, from 312 ppm in 1960 to 409 ppm in 2018, as well as annual mean surface temperatures, rising by 2.3 and 2.5°C over 1960–2018 for the Cherskiy and Samoylov sites, respectively (Figure 4a,d,h). The perturbation of heat exchange between atmosphere and soil surface is also reflected through warming rates in simulated annual mean upper-soil temperatures (30 cm), which show increases of 1.6 and 2.0°C at the both sites. Over the same time period and all Arctic sites, atmospheric temperatures rise by 2.4°C, leading to a soil temperature increase of 2.1°C at 30 cm depth. The warming impacts the depth of the permanently frozen layer in our simulations, with the active layer depth increasing by 0.19 and 0.15 m over 1960–2018 due to increased thaw at Cherskiy and Samoylov, respectively, and by 0.20 m at the high Arctic average (Figure 4b,f,j). Based on SOC contents averaged for the period 1950–1970 in climate + CO₂, these increases in active layer depth would expose 1.4–6.9 kg C m⁻² year⁻¹ to potential degradation over 1960–2018 (low value for Samoylov, high value for Cherskiy), of which 71–338 g C m⁻² year⁻¹ is fast-degrading. Following exposure of permafrost carbon to non-freezing temperatures, soil respiration increases by 105 and 26 g C m⁻² year⁻¹ at Cherskiy and Samoylov, with the deep soil (50–100 cm) responsible for 21 and 5 g C m⁻² year⁻¹ of the increase. Over all Arctic sites, the soil respiration increase is of 66 g C m⁻² year⁻¹, with deep-soil respiration increasing by 13 g C m⁻² year⁻¹. The shallowest layer of the deep-soil vertical range (~50 cm) is only exposed to above-zero temperatures for 2–6 weeks of the year (low value for Samoylov, high value for Cherskiy) over 1998–2018 on average.

Soil thaw simulated by the model also exposes 5.9 and 21 g N m⁻² year⁻¹ and 2.0 and 9.4 g P m⁻² year⁻¹ of fast-degrading organic N and P at Cherskiy and Samoylov, respectively, to temperatures above zero, and thus mineralisation. Over the Arctic mean, the increases in active layer depth over the 1960–2018 time period increase organic N and P mineralization by 0.6 and 0.2 g N m⁻² year⁻¹ and 0.01–0.03 g P m⁻² year⁻¹ in the deep soil aggregated over the time span at Cherskiy and Samoylov, respectively. Averaged over all high Arctic sites.

![Figure 4](image-url) Temporal evolution of annual mean air surface temperature (left column, panels a,e,i), simulated average August thaw depth (middle left, panels b,f,j), simulated annual N mineralization from 50 to 100 cm (deep-soil N min., middle right, panels c,g,k) and simulated annual Gross primary productivity (GPP) (right column, panels d,h,l) over 1960 to 2018. Deep-soil min. and GPP are shown for simulations considering CO₂ fertilization and climate perturbations (climate + CO₂), as climate perturbations alone (climate) and climate perturbations without permafrost carbon and nutrient initialisations (climate + withoutpermafrostCNP) at Cherskiy (CHE), Samoylov (SAM) and averaged over all high Arctic sites.
all sites, we compute an increase of 0.6 g N m\(^{-2}\) year\(^{-1}\) and 0.04 g P m\(^{-2}\) year\(^{-1}\) of organic N and P mineralization for the same time period. In the case of N, this means a steep 2- to 3-fold increase with respect to 1950–1970 organic N mineralization levels, which were close to zero (Figure 4c,g,k). In the case of P, the net transformation of mineral-bound P to biologically available P increases by 0.06–0.19 g P m\(^{-2}\) year\(^{-1}\) in the deep soil at the Arctic average.

As a result of the deepening of the active layer, annual mean root carbon content undergoes a strong relative increase at depths of 50–100 cm (+0.3 g C m\(^{-2}\), or +156% in relative terms). Overall, however, the root carbon distribution only experiences minor changes in our simulations when comparing the beginning of the 1960s with present-day conditions, with around 85% of roots still simulated in the upper 20 cm of the soil (Figure 5a–c) for 1998–2018. Root growth past 50 cm is inhibited to 2–6 weeks during the year due to freezing temperatures in the deep soil.

Plant nutrient uptake in the deep soil is enhanced by 0.6 and 0.1 g N m\(^{-2}\) year\(^{-1}\), and 0.06 and 0.09 g P m\(^{-2}\) year\(^{-1}\) at Cherskiy and Samoylov, respectively, over the simulation time period (Figure 5d–f). Averaged over all sites, deep-soil plant nutrient uptake increases by 0.3 g N m\(^{-2}\) year\(^{-1}\). This leads to a total mismatch in 1998–2018 deep-soil N mineralization of 1.2 and 0.2 g N m\(^{-2}\) year\(^{-1}\); and 1998–2018 plant N uptake of 0.3 and 0.1 g N m\(^{-2}\) year\(^{-1}\); at Cherskiy and Samoylov, respectively (Figure 5d–f). Averaged over the high-Arctic scale, mean 1998–2018 deep-soil N mineralisation is 1.1 g N m\(^{-2}\) year\(^{-1}\) and plant N uptake is 0.6 g N m\(^{-2}\) year\(^{-1}\). Our results suggest that, averaged over all Arctic sites, around 40%–60% of N mineralised in the deep soil is taken up through plant uptake over the 1998–2018 time slice. Due to the mismatch of N and P release and plant uptake in the deep soil, our model also simulates a vertical gradient in nutrient availability increasing towards the permafrost front for 1998–2018 (Figure S9).

### 3.4 Implications for GPP compared with other perturbations

Our simulations indicate an increase in GPP for all case-study sites over 1960–2018 in the climate + CO\(_2\) simulation, as result of all perturbations (Figure 4d,h,l). Increased nutrient availability from the thawing of permafrost thereby contributes to GPP increases of 9 and 5 g C m\(^{-2}\) year\(^{-1}\) aggregated over the entire 1960–2018 timespan at the Cherskiy and Samoylov sites, and 10 g C m\(^{-2}\) year\(^{-1}\) on average over all Arctic sites (Figures 6 and 9). Other climate-induced changes account for increases in the GPP of 58 and 32 g C m\(^{-2}\) year\(^{-1}\) at both sites, and 57 g C m\(^{-2}\) year\(^{-1}\) at the Arctic scale. Alone, CO\(_2\) fertilization accounts for rising the GPP by only 18 g C m\(^{-2}\) year\(^{-1}\) at Cherskiy, and a non-significant effect at Samoylov. Over the high-Arctic average, GPP increases by 11 g C m\(^{-2}\) year\(^{-1}\). At the Cherskiy and Samoylov sites, GPP trends arising from all perturbations account for a total increase in GPP of 84 and 37 g C m\(^{-2}\) year\(^{-1}\) aggregated, respectively. Averaged over all Arctic sites, the GPP increase resulting from all perturbations is of 79 g C m\(^{-2}\) year\(^{-1}\). These changes represent increases of 30% and 27% with respect to 1950–1970 mean levels at our case study sites, and a 36% increase at the high Arctic scale. Climate-induced effects (including increased nutrient availability due to permafrost thaw) overall account for over 70% of the total GPP change in all cases at the high Arctic scale, while increased nutrient availability following permafrost thaw, and CO\(_2\) fertilization, account for around 15% of the GPP increase each.
3.5 Implications for N loss pathways and N\textsubscript{2}O emissions

Since newly mineralized N from previously frozen organic soil layers is not completely removed from the deep soil through plant uptake (Figure 5), solute N concentrations increase at the permafrost front (Figure 59). This leads to increasing deep-soil loss pathways of N (i.e., nitrification, denitrification, and transport) in the CO\textsubscript{2} + climate simulation. At Cherskiy and Samoylov, we compute increases in nitrification of 166 (+10%) and 95 (+25%) mg N m\textsuperscript{-2} year\textsuperscript{-1}, and of 77 (+44%) and 14 (+40%) mg N m\textsuperscript{-2} year\textsuperscript{-1} in denitrification over the simulation time period, respectively. At the high Arctic average, we compute an increase in nitrification of 138 mg N m\textsuperscript{-2} year\textsuperscript{-1} (+12%) and an increase in denitrification of 43 mg N m\textsuperscript{-2} year\textsuperscript{-1} (+25%). Both increases in nitrification and denitrification occur in concurrence to the period of deepest thaw depths in the model (Figure 7a–c). The increases in nitrification and denitrification cause a simulated rise in N\textsubscript{2}O fluxes of 1.6 and 0.4 mg N m\textsuperscript{-2} year\textsuperscript{-1} to the atmosphere at Cherskiy and Samoylov, respectively, leading to present-day fluxes of 5.2 and 1.1 mg N m\textsuperscript{-2} year\textsuperscript{-1} for 1998–2018 (Figure 8b–d). Over all high-Arctic sites the 2.2 mg N m\textsuperscript{-2} year\textsuperscript{-1} mean increase in N\textsubscript{2}O emissions lead to contemporary emissions of 3.5 mg N m\textsuperscript{-2} year\textsuperscript{-1}. This suggests an increase of more than 50% over all sites over 1960–2018 at the Arctic average. The growing season emissions computed by our model compare well to median emissions observed by Marushchak et al. (2021) for the Yedoma domain in Eastern Siberia, and is situated within the wide range of estimates made until now for the Arctic (Figure 8e).

N\textsubscript{2}O emissions and their trends over 1960–2018 are higher in climate + CO\textsubscript{2} than in climate + without permafrost CNP, which does not consider increased N availability via the mobilization of permafrost pools, at both case study sites and at the high Arctic average (Figures 8b–d and 9c). While denitrification and nitrification are enhanced shortly before or following soil thawing peaks, increased N\textsubscript{2}O emissions occur later when the deep-soil re-freezes during weeks 40 to 42. This is also the timeframe when the deep soil is aerated again in the model during freezing (Figure S2). Overall, the months of September and October account for over half of total yearly N\textsubscript{2}O emissions averaged over all sites for 1998–2018.

The magnitude of N\textsubscript{2}O emission rise shows an increasing relation with prescribed total soil N (Figure 10a). The magnitude in the site-level N\textsubscript{2}O emissions trend is also dependent on degree of soil thaw, with increasing soil thaw seemingly leading to increasing N\textsubscript{2}O emissions (Figure 10b). Indeed, our model shows a stronger response of N\textsubscript{2}O emissions to increasing thaw depths than for GPP (Figure 10b). We also show significant correlation in the simulated year-to-year magnitude of the N\textsubscript{2}O fluxes shows with surface air temperature (2 m) at Cherskiy (0.39) and Samoylov (0.5).

4 DISCUSSION

4.1 Modelling the terrestrial biosphere of high latitudes in QUINCY

Estimating human-caused changes in greenhouse gas exchanges over time is presently strongly dependent on global land surface models (Canadell et al., 2022; Friedlingstein et al., 2020). The representation of high-latitude biophysical processes in these models is, however, still associated with strong limitations. Models have, for one, been lacking in terms of their vertical representation of physical and biogeochemical features in permafrost-affected soils (Schädel et al., 2016). Second, nutrient cycles in high latitudes are either omitted or strongly simplified in these models (Chadburn et al., 2017), even though both N and P impose important constrains for microbial and vegetation dynamics in the high latitudes (Finger et al., 2016; Haag, 1974; Natali et al., 2012; Pedersen et al., 2020; Schaller et al., 2019; Yang et al., 2021). Through the full vertical discretization of both physical and biogeochemical variables, our extended QUINCY model version permits the analysis of changes associated with vertical permafrost thaw and their impact on vegetation and soil dynamics, while also taking into account a full representation of C, N and P cycles. Our model representation of dynamic changes in root distributions following soil thaw offers a simple, yet improved formulation to assess impacts of changing long-term and seasonal soil conditions on vegetation growth and increased nutrient availability in the deep soil. Until now, only strongly simplified impacts of N release from the permafrost have been assessed using bucket-type approaches (Beermann et al., 2017), whereas we offer an improved dynamic representation of the interaction of permafrost
thaw, increased C-N-P mobilization following the thaw, and vegetative uptake and microbial dynamics.

The model presented here still has clear limitations in its biophysical representation of high-latitude permafrost ecosystems. Strong simplifications are made in both physical soil processes, as well as vegetation dynamics. The parameterizations implemented here are often of first order, with organic matter mineralization for instance being constrained by a singular rate constant in combination with a temperature and soil moisture scaling factor. While parameters in our model have been assessed in terms of representing important land surface variables at the global scale (e.g., Thum et al., 2019), as well as the assessments for permafrost-affected soils we provide in this study, improvements in quantifying both vegetative and soil processes both from observational and modelling perspectives are still needed in the high Arctic.

Important challenges remain in terms of representing soil moisture and energy fluxes associated with snow cover. These are largely community-scale problems, since land surface models struggle to represent the observed local conditions of these variables. In our study, we assume fixed constants for snow density, specific heat conductivity, and heat capacity, which are partly computed dynamically in other models (Chadburn et al., 2015; Guimberteau et al., 2018). We also assume soil heat conductivity and specific heat capacity also do not change dynamically as a response to soil water content and soil organic matter, which could lead to certain bias in soil temperatures. A further source of uncertainty arises from the handling of species composition in the model. Assuming homogenous grassland coverage could cause biases in GPP as shown for instance at the Samoylov site, where GPP is reported to be majorly affected by mosses (Holl et al., 2019). Such biases could potentially be reduced by improving the representation of Arctic plant functional types in global vegetation models (Sulman et al., 2021).

Our model only represents limited spatial heterogeneity, both in the vertical and horizontal plane, which is partly linked with the polygonal tundra landscape found at many sites. The heterogeneity in soil variables such as heat conductivity, even at the local scale (Boike et al., 2019; Göckede et al., 2019), could strongly impact the degree of thaw at these scales. Our results should thus be interpreted as the bulk mean of the spatial fluxes with a strong degree of local uncertainty, and also omitting non-linear effects such as landscape collapse following deep-soil thaw (e.g., Yang et al., 2021).

4.2 | Changes in vegetation growth due to increased nutrient availability from the permafrost compared with increases caused by other drivers

Over the past century, Arctic regions have been subject to disproportionate warming due to climate change (e.g., Serreze & Barry, 2011), which has been reported to already affect dynamics of vegetation in the high latitudes. Through our set of simulations, we can decompose total perturbations caused by increased nutrient and carbon availability through permafrost thaw, other climate effects (e.g., earlier growth season, higher atmospheric and soil temperatures) and CO₂ fertilization alone. Our computed GPP increase of around 40% for the 1960–2018 period is
FIGURE 8  Panel (a) shows trends in N₂O emissions over all Arctic sites individually, aggregated over the 1960–2018 time span. Panel (b) shows the evolution of annual mean N₂O emissions over 1960–2018 at Cherskiy (b), Samoylov (c) and all high Arctic sites (d). Panel (e) shows averaged growing season (June–August) emission rates averaged over all Arctic sites in CO₂ + climate versus observed estimate means (solid lines) and upper and lower ranges (dashed lines) reported by Wilkerson et al. (2019) and Marushchak et al. (2021).

FIGURE 9  Simulated response of Gross primary productivity (GPP) and total ecosystem respiration (TER, panel a), net ecosystem production (NEP, panel b) and N₂O emissions (panel c) to climate effects other than increased permafrost nutrients and carbon (other climate, green), permafrost C, N, and P (orange) and CO₂ fertilization (khaki), all averaged over all sites.
consistent with increases in greenness of around 10% per decade since the 1980s derived from satellite estimates (Winkler et al., 2019).

Our model suggests that increased nutrient input from permafrost thaw is responsible for less than 15% of the total GPP change over 1960–2018 at the high-Arctic scale. This effect could be increased in the future, with increased thaw meaning further increases in nutrient supply, as well as longer growing season periods of potential vegetation uptake (Pedersen et al., 2020). Other climate effects have the largest effect on the simulated GPP trend over our investigated time period. Our results also suggest a trend towards a slightly earlier start and peak of the growing season with our simulations showing a 2-weeks earlier growing season peak of GPP at Cherskiy, and 1-week earlier peak at Samoylov and at the Arctic scale over the 1960 to 2018 period (Figure S10). This feature is in agreement with the study of Park et al. (2019), who report a historical trend in start and peak of vegetation growth of 1 day per decade derived from a satellite-based approach. CO₂ fertilization plays only a small role in explaining the GPP increase at the Arctic scale in our model, which likely owes to vegetation growth being strongly N limited at our case-study sites, which is also confirmed by the simulated GPP trend being dependent on total soil N.

4.3 Root dynamics as response to deep-soil thaw and incomplete use of deep-soil N

Ecosystems in the Arctic tundra are reported to be strongly nutrient-limited during the growing season (Martin et al., 2022; Schuur et al., 2015), owing to shallow active layers meaning a limited nutrient pool accessible for plant roots and lower mineralization rates due to thermal constraints. In accordance to this, nutrients added to the active layer in fertilization experiments have been shown to strongly enhance vegetation growth (Natali et al., 2012; Pedersen et al., 2020; Street et al., 2018). In our simulations, historical increases in permafrost organic matter mineralization led to the release of additional N to the active layer, especially in late summer. This N fertilization effects is, however, only minorly impacted the change in GPP, which our results suggest is a consequence of incomplete plant use of the N released in the deep soil.

The incomplete use of increased N supply firstly owes to the mismatch in the timing of simulated peak vegetation growth and soil thaw. Both our model and collected observations show that peak vegetation growth occurs 4–8 weeks earlier than the timing of maximum permafrost thaw, the time period when new nutrients in the deep active soil are mineralized to biologically available compounds. In addition to this, soil conditions only allow for root growth in the deep soil for several weeks in the late summer before the deep soil re-freezes, greatly limiting the potential for deep-soil build-up of root mass. Due to this temporal and spatial mismatch, plants only use around half of increased N mineralised in the deep soil in our model, leading to accumulation of N in the deep soil. Treat et al. (2016) suggest a similar mismatch due to increased N availability in the fall owing to deep-soil thaw, leading to higher biologically available N in the deep soil in this season.

Few studies have assessed seasonal and long-term dynamics of root growth and plant uptake in relation the dynamics of soil thaw. In our model, the vast majority (>85%) of roots are simulated in the top 20–30 cm of the soil, which is in accordance to observed vertical root distribution profiles (Blume-Werry et al., 2019; Iversen et al., 2015; Keuper et al., 2020). Since the vegetation is already in senescence by the time the seasonal thaw maximum is reached, the vegetation’s investment into root growth is low in the model. However, it remains unclear if these relatively small root fractions in the deep soil (<5% of total root mass) could lead to disproportionate uptake of nutrients in the deeper soil. In the case of Eriophorum vaginatum, adapted uptake in the late summer could give the species a competitive advantage in the case of active layer deepening (Blume-Werry et al., 2019). Uptake of deep-soil N in late autumn has also been demonstrated for Salix arctica by Pedersen et al. (2020)
through the use of stable isotopes, but overall, the magnitude that a competitive advantage of individual species exhibits at the larger scale still remains enigmatic, since most plant species found in the Arctic Tundra seldomly root past 40 cm depth (Iversen et al., 2015).

Finger et al. (2016) postulate in their analysis of N availability over permafrost thaw gradients that vegetation uses additional N made available through permafrost thaw over time-scales of years to centuries. Thus longer time periods may be needed for plants to adapt to increasing N availability in the deeper permafrost-affected soil. A key uncertainty in our study is that we cannot resolve with the present observational basis, is the deep-soil uptake rate of N by plants, which could be dictated by adaptation abilities of the vegetation to increase their rooting depth with increasing soil thaw.

4.4 Implications for N loss pathways and N$_2$O emissions

Due to the incomplete use of deep active layer nutrients by plants, our model shows accumulation of both biologically available N near the permafrost table. Ultimately, three pathways are plausible following increased N concentrations in the deep soil. Either they are mixed into the active soil layer and thus made available for plant uptake over longer timescales (Finger et al., 2016), exported laterally (Treat et al., 2016), or used microbially for denitrification and outgassed as either N$_2$O or N$_2$ (Voigt et al., 2020).

Our model suggests an important increase in N loss pathways through nitrification and denitrification, which results in an enhancement of N$_2$O emissions over the high Arctic. The magnitude in the N$_2$O emissions increase is dependent on local total soil N, meaning the accumulation is larger with higher N mobilization from the permafrost. Our model also simulates higher loss of N and higher site-level N$_2$O emission trends with larger trends with deep thawing on average. The limited accessibility of roots to increasingly deeper soil layers could lead to a less N efficient uptake of the deep-soil N, and thus exposing larger amounts of N to other loss pathways. This is illustrated by Cherskiy having a reduced relative response in GPP but higher relative N$_2$O increase than at Samoylov, with a higher degree of thaw. In turn, our model even suggests a decrease in emissions only at Kangerlussuaq, Greenland (DK-KAN, Figure 52), where thaw depths do not exceed 50 cm, thus providing poor conditions accumulation of N and for production of N$_2$O.

Since lateral exports remove biogeochemical compounds from permafrost systems, it could affect the potential use of nutrients originating from permafrost thaw by plants and microbes (Treat et al., 2016). Lateral water and biogeochemical exports from the permafrost are, however, still poorly constrained (Lacroix, Ilyina, Mathis, et al., 2021; Lacroix et al., 2020), and difficult to assess in our model-based study. These lateral exports could however have important implications for high-Arctic freshwater systems (Sanders et al., 2022) and Arctic shelves (Dai et al., 2022; Lacroix, Ilyina, Laruelle, et al., 2021; Terhaar et al., 2021).

4.5 High Arctic as a weak, but emergent source of N$_2$O gas

N$_2$O is a strongly potent greenhouse gas and its global budget could impact the global climate in the future (Tian et al., 2020). The magnitude of N$_2$O emissions from high latitudes has, however, been largely disregarded due to low active layer nutrient contents in these regions (Marushchak et al., 2021). Recent studies that have measured N$_2$O fluxes at site-level, have reported weak emissions per area in permafrost-affected ecosystems, that could add up to an important component of the global N$_2$O budget (Voigt et al., 2020), in addition to hot-spot emissions that could disproportionately affect the N$_2$O budget in the high latitudes (Marushchak et al., 2021; Wilkerson et al., 2019). The overall magnitude of our present-day N$_2$O emission average at the high Arctic scale (4 mg N m$^{-2}$ year$^{-1}$) is still around three-fold lower than the reported estimate of 11 mg N m$^{-2}$ year$^{-1}$ of Voigt et al. (2020). However, our averaged growing season emissions at the Arctic scale are close to the median measured over the Yedoma domain by Marushchak et al. (2021), with estimates of Wilkerson et al. (2019) being larger for August emissions for the Alaskan Tundra. Our model shows that high-Arctic N$_2$O emissions are threefold higher when considering permafrost nutrients and carbon contents, and thus are already driven by historical thaw of the permafrost. The results also suggest pathways leading to N$_2$O emissions could increase disproportionally with future deeper soil thawing because Arctic plants could have limited access to the newly released nutrients in the deeper soil.

The simulated seasonally changing soil hydrological conditions, and thus changes in soil aeration that are needed for nitrification and denitrification, impact the magnitude of the fluxes. Short aerobic periods are a requisite for nitrification, whereas longer anaerobic timespans are needed for denitrification (Voigt et al., 2020), both of which are simulated at our case studies. However, there remains strong uncertainty in our model in terms of accurately representing soil moisture (Figure 3), a notable problem for terrestrial biosphere models that remains to be resolved. Our model does, however, reproduce the time of thawing, and ensuing anaerobic conditions in the deep soil. Our results suggest an increase in total nitrification and denitrification following deep-soil thaw in the late summer. The emission peak only takes place later during soil re-freezing, when the soil is aerated again in the model. These simulated temporal dynamics entail substantial uncertainty, and further work is needed to improve observational constraints on magnitudes and seasonality of N fluxes in permafrost-affected soils.

5 Conclusions

Our model results show that N release from permafrost thaw since the 1960s leads to a fertilization effect on plant growth and vegetation C uptake. These effects are low in comparison to other
climate-caused perturbations due to two factors backed by both our model simulations and collected observational data:

1. Deep-soil N mineralization and vegetation growth peaks are temporarily decoupled, meaning that N released in the deep soil is not available during the time period of the largest plant N demand.

2. Arctic vegetation presently does not invest large resources into deep-soil root growth and uptake, likely owing to the short time window of growth in the deep soil prior to re-freezing. It is, however, debated how well individual species adapt to seasonal and long-term variations in thaw depths and nutrient release in the deep soil.

Both these effects lead to N accumulation near the permafrost table following permafrost thaw, opening N loss pathways in our model, which is of particular importance with respect to N$_2$O emissions in the high Arctic. Through the mismatch of peak vegetative growth and deep-soil N release, our results offer a mechanistic explanation for a considerable present-day N$_2$O flux despite generally low biologically-available N concentrations in the active layers of permafrost-affected soils. Our results suggest that N$_2$O emissions are already driven by the mobilization of permafrost material and could increase disproportionately with the extent of deepening of the active layer, leading to a positive non-carbon climate feedback. Since thawing of permafrost is projected to substantially increase over the next century, our results call for increased efforts to improve the understanding of the fate of N upon permafrost thaw and to improve observational constraints of C-N-P dynamics, and especially of N$_2$O emissions in the high Arctic.

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**CONFLICT OF INTEREST**

All authors declare no conflict of interests.

**DATA AVAILABILITY STATEMENT**

Model results and code, as well as constructed eddy-covariance GPP seasonality, are available under https://doi.org/10.5281/zenodo.6832235.

**ORCID**

Fabrice Lacroix https://orcid.org/0000-0003-4749-2826
Sönke Zaehle https://orcid.org/0000-0001-5602-7956
Silvia Caldararu https://orcid.org/0000-0001-5839-6480
Jörg Schaller https://orcid.org/0000-0003-1996-0127
Peter Stimmer https://orcid.org/0000-0002-8599-5486
David Holl https://orcid.org/0000-0002-9269-7030
Mathias Göckede https://orcid.org/0000-0003-2833-8401

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**SUPPORTING INFORMATION**

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