Response to Major Revision Comment by Topical Editor

The length of the text is way too much, and it was quite painful to read throughout the document. This problem became much worse after the revision, although previous reviewer pointed out this issue. Of course, authors have reason to add new texts for addressing reviewer's comments as authors mentioned it in their reply letter. But, still, due to the absence of readability, it is indeed very painful to read through the document. Actually, both previous referees refused to review the revised manuscript, and this is the first case in my long experience as an editor.

I agree that authors addressed all issues raised by previous referees. It's OK. But, the manuscript should be re-organized to reduce the amount of main text substantially (like, less than half). I believe that majority of the text in the section 4 and some figures can be moved to the supplemental information.

Thank you for taking the time to go through this iteration of this manuscript. We understand the issues that you and the referees have raised, and have accordingly reduced the main text body from 21 to 11 pages, and its total length (including Figures, etc) from 45 to 27 pages. These reductions have entailed a mix of deletion and the movement of large sections of the text and Figures into the Supplement, focusing largely on Section 4, as you have helpfully pointed out. On the other hand we were somewhat perplexed by the reported refusal to referee the previous version of the manuscript. All prior specific and general comments by referees were responded to, and indeed we had substantially reduced and restructured the text as per a non-specific suggestion to do so, while as a native English speaker the first author finds the charge of 'unreadability' rather puzzling. Since this manuscript has been in review for over seven months, we were surprised to learn that we should cut the text body by a factor of two or more at this late stage, given that this was never specifically raised in such a context previously. Again, we understand and agree that the length of the document posed problems of overburden for the reader in general and the referees specifically, however we felt this necessarily reflected the very broad content of such a study, given that it appeals to both the modelling and field work communities, and would need to satisfy their rigor and interest across a large number of domains (primary production, hydrology, data, soil science and thermodynamics, and the rather more niche specificities of the inland water continuum -DOC discharge, concentration, evasion, etc.) within the context of the model's structure and function, or lack thereof. We look forward to moving on to the next iteration of this manuscript's evolution, and thank you for your patience with and participation in that process.
Title: ORCHIDEE MICT-LEAK (r5459), a global model for the production, transport and transformation of dissolved organic carbon from Arctic permafrost regions, Part 2: Model evaluation over the Lena River basin.

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Abstract

In this second part of a two-part study, we perform a simulation of the carbon and water budget of the Lena catchment with the land surface model ORCHIDEE MICT-LEAK, enabled to simulate dissolved organic carbon (DOC) production in soils and its transport and fate in high latitudes inland waters. The model results are evaluated in their ability to reproduce the fluxes of DOC and carbon dioxide (CO₂) along the soil-inland water continuum, and the exchange of CO₂ with the atmosphere, including the evasion of CO₂ from inland waters. We present simulation results over years 1901-2007, and show that the model is able to broadly reproduce observed state variables and their emergent properties across a range of interacting physical and biogeochemical processes, including: 1) Net primary production (NPP), respiration and riverine hydrologic amplitude, seasonality and inter-annual variation; 2) DOC concentrations, bulk annual flow and their volumetric attribution at the sub-catchment level; 3) High headwater versus downstream CO₂ evasion, an emergent phenomenon consistent with observations over a spectrum of high latitude observational studies. (4) These quantities obey emergent relationships with environmental variables like air temperature and topographic slope that have been described in the literature. This gives us confidence in reporting the following additional findings: Of the ~34TgC yr⁻¹ left over as input to soil matter after NPP is diminished by heterotrophic respiration, 7 TgC yr⁻¹ is leached and transported into the aquatic system. Of this, over half (3.6 TgC yr⁻¹) is evaded from the inland water surface back into the atmosphere and the remainder (3.4 TgC yr⁻¹) flushed out into the Arctic Ocean, mirroring empirically derived studies. These riverine DOC exports represent ~1.5% of NPP. DOC exported from the floodplains is dominantly sourced from recent, more ‘labile’ terrestrial production, in contrast to DOC leached from the rest of the watershed with runoff and drainage, which is mostly sourced from recalcitrant soil and litter. All else equal, both historical climate change (a spring/summer warming of 1.8°C over the catchment) and rising atmospheric CO₂ (+85.6ppm) are diagnosed from factorial simulations to contribute similar, significant increases in DOC transport via primary production, although this similarity may not hold in the future.
1 Introduction

A new branch of the high latitude-specific land surface component of the IPSL Earth System model, ORCHIDEE MICT-LEAK (r5459), was enabled to simulate new model processes of soil dissolved organic carbon (DOC) and CO₂ production, and their advective/diffusive vertical transport within a discretized soil column as well as their transport and transformation within the inland water network, in addition to improved representation of hydrological and carbon processes in floodplains. These additions, processes first coded in the model ORCHILEAK (Lauerwald et al., 2017) and implemented within the high latitude base model ORCHIDEE-MICT v8.4.1 (Guimberteau et al., 2018), were described in detail in Part 1 of this study. depicted graphically in Figure S1a,b. This second part of our study deals with the validation and application of our model. We validate simulation outputs against observation for present-day and run transient simulations over the historical period (1901-2007) using the Lena River basin as test case. The simulation setup and rationale for choice of simulation basin are outlined below.

2 Simulation Rationale

The Lena river basin, which is bounded by the region 52-72°N; 102-142°E, was chosen as the basin for model evaluation because it is the largest DOC discharge contributor amongst the Arctic rivers, according to some estimates (Raymond et al., 2007; Holmes et al., 2012), with its 2.5 million km² area (befitting our coarse-grid resolution) discharging almost 20% of the summed discharge of the largest six Arctic rivers, its large areal coverage by Podzols (DeLuca and Boisvenue, 2012), and the dominance of DOC versus particulate organic carbon (POC) with 3-6Tg DOC-C yr⁻¹ vs. 0.03-0.04 Tg POC-C yr⁻¹ (Semiletov et al., 2011) in the total OC discharge load –factors all broadly representative of the Eurasian Arctic rivers. Climatological input to the model is from the Global Soil Wetness Project Phase 3 (GSWP3) v0. data, based on 20th Century reanalysis using the NCEP land-atmosphere model and downscaled to a 0.5°, 3-hourly resolution covering the period 1901 to 2007 (Supplement, Table S1). This is then upscaled to 1° resolution and interpolated to a 30 minute timestep to comply with the timestep of ORCHIDEE’s surface water and energy balance calculation period. Precipitation was partitioned into rainfall and snowfall, and a correction for wind-induced undercatch was also applied. These are described in greater detail in Guimberteau et al. (2018). Over the simulation period under this dataset, the Lena basin experiences a mean thaw period warming of 1.8°C, while atmospheric CO₂ concentrations increase by 85.6ppm. The GSWP3 dataset was chosen due to its relative performance in simulating the inter-annual variability and seasonality of Pan-Arctic riverine discharge in ORCHIDEE-MICT (Guimberteau et al., 2018), as compared to another data-driven climate forcing product, CRUNCEP v7 (Kalnay et al., 1996; New et al., 1999). Indeed, under CRUNCEP v7, ORCHIDEE-MICT was shown to underestimate river discharge by as much as 83% over the Yukon basin. An improved floodplains area input file for the Lena basin (Tootchi et al., 2019) was used to drive the simulation of floodplain dynamics (Supplement, Table S1).

3 Simulation Setup

As detailed in Part 1 (Section 3.1), the soil carbon stock used by our model was reconstituted from a 20,000 year soil carbon spinup of an ORCHIDEE-MICT run from
4 Results and Interpretation

We refer to different simulations performed in this study according to the sensitivity factors to which they are subjected. The transient, historical climate and atmospheric CO$_2$-forced simulations are hereafter referred to as the "Control" (CTRL) scenario, for ease of interpretation. The "CLIM" and "CO$_2$" scenarios are those simulations for which climate variability and atmospheric CO$_2$ were held constant at their pre-industrial levels, respectively (Table 1). The following evaluation sections compare observations solely against the CTRL. The subsequent section will evaluate this comparison against the factorial simulations described above. The overall carbon budgets and their fluxes as generated by each of the simulations are shown in Figs. 2 and 11 and discussed in detail at the end of the evaluation. In the following we report first the broad results of model simulations with respect to the carbon cycle, and follow with an evaluation of river water and DOC discharge, DOC concentration and seasonality and river surface CO$_2$ outgassing against available empirical data. Evaluation of NPP and Soil Respiration, which are not considered primary to this study, is covered in Supplementary Text S1.

4.1 Model Output: Carbon Budget

Fig. 2 summarises the simulated components of the carbon (C) cycle across the Lena basin, averaged over the decade 1998-2007. C inputs to terrestrial ecosystems are dominated by photosynthetic input (GPP). GPP assimilates (875 Tg C yr$^{-1}$) are either used as metabolic substrate by plants and lost as CO$_2$ by plant respiration processes (376 Tg C yr$^{-1}$) or soil respiration processes (465 Tg C yr$^{-1}$), leaving behind annual growth in terrestrial C storage (net biome productivity (NBP)), an atmospheric CO$_2$ sink of 34 Tg C yr$^{-1}$. Further C inputs are delivered to the terrestrial surface via a combination of atmospheric deposition, rainwater dissolved C, and the leaching of canopy C compounds. These sum to a flux transported to the soil surface (4.6 Tg C yr$^{-1}$) by throughfall (see Part 1, Section 2.5).
DOC in the soil solution as well as a fraction of dissolved CO₂ produced in the root zone from root and microbial respiration is exported to rivers along the model’s two hydrological export vectors, surface runoff and deep drainage (Part 1, Section 2.6). For the Lena basin simulations, these fluxes of C exported from soils amount to 5.1 and 0.2 Tg yr⁻¹ for DOC and CO₂ respectively. Three water pools, representing streams, rivers and groundwater and each containing dissolved CO₂ and well as DOC of different reactivity, are routed through the landscape and between grid cells following the river network in the catchment (Part 1, Section 2.7). In addition, seasonally flooded soils located in low, flat grid cells next to the river network (see Part 1, Section 2.8) export DOC (0.57 Tg yr⁻¹) and CO₂ (1.54 Tg yr⁻¹) to the river network when their inundation occurs. Part of this leached inundated material is re-infiltrated back into the soil from the water column during floodplain recession (‘Return’ flux, 0.45 Tg yr⁻¹). During its transport through inland waters, DOC can be decomposed into CO₂ (2.1 Tg yr⁻¹) and a fraction of river CO₂ produced from DOC and transferred from soil escapes to the atmosphere (3.6 Tg yr⁻¹) through gas exchange kinetics (Part 1, Section 2.10). This flux is termed ‘CO₂ evasion’ in Fig. 2 of this study. Carbon that survives the inland water reactor is exported to the coastal ocean in the form of DOC (3.16 Tg yr⁻¹) and CO₂ (0.26 Tg yr⁻¹).

4.2.1 Model Evaluation: River Discharge

Simulated river water discharge captures the key feature of Arctic river discharge – that of a massive increase in flow to ~80,000 m³ s⁻¹ in April-June caused by melting snow and ice, but underestimates observed river discharge in late summer by around 70% (Figs. 3c, 4b). In addition, the mean spring (June) discharge peak flows are slightly underestimated or out of phase in simulations (Figs. 3c, 4b) compared to observations (Ye et al., 2009): this is caused by a large amount of water throughput being simulated in May (~10,000 m³ s⁻¹) in excess of observed rates. Finally, during the winter low-flow period, the model consistently under-estimates water flow-through volumes reaching the river main stem (see Fig. 3c, winter months). Although this underestimate is not as severe relative to annual bulk flows, the divergence is large as a percentage of observations (see right-hand axis, Fig. 3c), and may point to an issue in how ice is represented in the model, such as the fact that solid ice inclusions in the soil column are not represented, or the possibility that much slower groundwater dynamics than those represented in the model are feeding discharge. In addition to this, the presence of a dam on the Vilui tributary of the Lena has been shown to reduce main stem winter low-flow rates by up to 90% (Ye et al., 2003), similar to the discrepancy of our low-flow rates: given that our model only simulates ‘natural’ hydrological flows and thus does not include dams, we expect that this effect is also at play. Causal factors for the apparently poor performance of the hydrological module range from poor model representations (or lack thereof), climatological dataset choices and deficiencies in evaluation datasets themselves, and are covered in detail in the Supplement (Text S2).

4.2.2 Model Evaluation: DOC Annual Discharge

Our CTRL simulation shows that the yearly sum of DOC output to the Arctic Ocean has increased steadily over course of the 20th Century, from ~1.4 Tg DOC-C yr⁻¹ in 1901 to ~4 Tg DOC-C yr⁻¹ in 2007 (Fig. 4a). Smoothing the DOC discharge over a 30-year

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**Supprimé:** In the soil, DOC is produced by the decomposition of litter and soil organic carbon (SOC) pools (see Part 1, Section 4.2 and Fig. 2) and can be adsorbed or desorbed to solid particles (see Part 1 of this study, Section 2.11), while there is a continuous exchange of DOC with (solid) soil organic carbon. The interplay between decomposition and sorption leads to DOC concentration changes in the soil solution.

**Supprimé:** These fluxes and their interpretation within the context of the Land-Ocean-Aquatic Continuum (LOAC) are returned to in Section 4.8 of this study.

**Supprimé:** otherwise known as ice-out or spring freshet.

**Supprimé:** Given that DOC fluxes are almost directly proportional to river discharge in the Lena basin (Fig. 3d), this sub-optimal performance with regard to hydrology during August to October seeming to be the main cause of a substantial underestimation in simulated bulk DOC outflow. Another cause may simply be the lack of peat representation in the model, for which DOC flux concentrations in outflowing fluvial water can be very high (e.g. Frey et al., 2005; 2009: see Section 4.5.1). Deficiencies in modelled hydrology hydrology correspond to those found in Fig. 12 of Guimberteau et al. (2018), indicating that the modifications made in this model version, which focus on the DOC cycle, have not further degraded the hydrological performance of the model, the causes of which are described below. Low simulated discharge for the Lena basin, particularly during the late summer and autumn, is consistent with prior, Pan-Arctic simulations conducted by Guimberteau et al. (2018), who ran ORCHIDEE-MICT using both the GSWP3 and CRU-NCEP v7 datasets and evaluated them over the period 1981-2007. Despite the substantially better hydrological performance of ORCHIDEE under GSWP3 climate, they described a near-systematic underestimation of discharge.

**Supprimé:** It seems that

**Supprimé:** Deficiencies in modelled hydrology correspond to those found in Fig. 12 of Guimberteau et al. (2018), indicating that the modifications made in this model version, which focus on the DOC cycle, have not further degraded the hydrological performance of the model, the causes of which are described below. Low simulated discharge for the Lena basin, particularly...
running mean shows that the increasing trend (Fig. 4a) over this averaging scale is almost linear, at ~0.11 TgC per decade, or a net increase of 40% using this averaging scale. Empirically based estimates of total contemporary DOC entering the Laptev Sea from Lena river discharge vary around ~2.5-5.8 TgC-DOC (Cauwet and Sidorov, 1996; Dolman et al., 2012; Holmes et al., 2012; Lara et al., 1998; Raymond et al., 2007; Semiletov et al., 2011).

Note however that modelled aggregate DOC discharge is strongly affected by the underestimation of river water discharge. Fig. 4a shows the average simulated DOC discharge (red bar) of the last decade (1998-2007) of 3.2 TgC yr⁻¹, to be compared with estimates of 3.6 TgC yr⁻¹ (black bar) from Lara et al. (1998) and 5.8 TgC yr⁻¹ (orange bar) from Raymond et al. (2007) and 5.7 TgC yr⁻¹ from Holmes et al. (2012). The most recent and elaborate of those estimates is that of Holmes et al. (2012) who used a rating curve approach based on 17 samples collected from 2003 to 2006 and covering the full seasonal cycle, which was then applied to 10 years of daily discharge data (1999-2008) for extrapolation. Given that their estimate is also based on Arctic-GRO-1/PARTNERS data (https://www.articgreatrivers.org/data), which stands as the highest temporal resolution dataset to date, their estimate is likely the most accurate of the DOC discharge estimates. Compared to their average annual estimate of 5.7 TgC yr⁻¹, our simulated DOC export is low by around 43%, due largely to the poor performance of the hydrology module. The DOC discharge underestimate is discussed in depth in Supplement (Text S2).

### 4.3 Model Evaluation: DOC Concentrations in lateral transport

While total DOC discharge captures the integral of biogeochemical processes leading to fluvial outflow, simulations of this are highly sensitive to the performance of modelled hydrology and climatological input data. A more precise measure for the performance of the newly-introduced DOC production and transport module, which is less sensitive to reproduction of river water discharge, is DOC concentration. This is because while the total amount of DOC entering river water depends on the amount of water available as a vehicle for this flux (hydrology), the concentration of DOC depends on the rate of soil carbon leaching itself depending largely on the interaction of soil biogeochemistry with primary production and climatic factors. This we evaluate in Figure 5a. This shows that for the majority of the thaw period or growing season (April-September), which corresponds to the period during which over 90% of DOC production and transport occurs, the model largely tracks the observed seasonality of DOC concentrations in Arctic-GRO data averaged over 1999-2007. There is a large overestimate of the DOC concentration in May owing to inaccuracies in simulating the onset of the thaw period, while the months June-September underestimate concentrations by an average of 18%. On the other hand, frozen period (November-April) DOC concentrations are underestimated by between ~30-50%. This is due to deficiencies in representing wintertime soil hydrological water flow in the model, which impedes water flow when the soil is frozen, as discussed in Section 5.2. Because of this deficiency, slow-moving groundwater flows that contain large amounts of DOC leachate are under-represented. This interpretation is supported by the fact that in both observations and simulations, at low discharge rates (corresponding to wintertime), DOC concentrations exhibit a strong positive correlation with river discharge, while this relationship becomes insignificant at...
As previously discussed, the proportion of total basin-wide CO₂ evasion attributable to higher levels of river discharge (Fig. 5b). Thus wintertime DOC concentrations suffer from the same deficiencies in model representation as those for water discharge. In other words, the standalone representation of DOC leaching is satisfactory, while when it is sensitive to river discharge, it suffers from the same shortfalls identified in Sections S2 and S3. Modelled DOC concentrations in stream, river and ground water are evaluated against data and discussed in the Supplement (Text S5).

4.4 In-Stream CO₂ Production, Transport, Evasion

In our model, the fate of DOC once it enters the fluvial system is either to remain as DOC and be exported to the ocean, or to be degraded to dissolved CO₂ (CO₂(aq.)), which is itself either transported to the marine system or outgassed from the fluvial surface to the atmosphere (see Part 1, Section 2.10 and Text S6). As noted in Part 1 of this study, although the model as a whole conducts simulations at the 1 degree scale, the routing of water and carbon, as well as the evasion of the latter, occurs at the sub-grid scale, such that we are able to simulate spatially explicit rivers whose size approximates Strahler order 4, and through the ‘fast’ water pool in the model are able to simulate streams of Strahler order 1-3. The seasonality of riverine dissolved CO₂ concentrations (CO₂(aq.), mg C L⁻¹) is evaluated in Fig. 4c to compare CO₂(aq) concentrations with DOC bulk flows, since CO₂(aq) concentrations follow an inverse seasonal pattern to those of DOC, being highest during the winter baseflow period and lowest in summer due to dilution during its high discharge phase (Semiletov et al., 2011). The simulated flow of CO₂(aq) at Kusur (Fig. 4c, dashed red) reproduces the seasonality of observations from Cauwet and Sidorov (1996), who sampled the Lower Lena (Fig. 3a), but somewhat underestimates concentrations. Also included in Fig. 4c is the basin average for all non-zero values, whose shape also tracks that of observations. Thus the model represents on the one hand increasing hydrological flow mobilising increasing quantities and concentrations of DOC while on the other hand those same increasing hydrological flows increasing the flux, but decreasing the concentration, of CO₂(aq) throughput.

Evaluation of modelled CO₂ evasion is beset by problems, not least that no data on this quantity have to our knowledge been recorded for the Lena (see Text S6). Figure 6 summarises some of the results from the simulated water body CO₂ outgassing flux. Year-on-year variation in basin-wide evasion from river, stream and floodplain sources combined exhibits a marked increasing trend over the course of the 20th Century, increasing from a minimum of ~1.6 TgCO₂-C yr⁻¹ in 1901 to a maximum of ~4.4 TgCO₂-C yr⁻¹ in 2007 (+300%, Fig. 7a). Smoothing the data over a 30 year running average yields a dampened net increase in basin-wide evasion of ~30% (Fig. 7a). Thus yearly evasion flux is some 105% of yearly DOC discharge to the coast from the Lena basin and 51% of C exported from soils to headwaters as CO₂ or DOC. If we compare the mean yearly rate of increase in absolute (TgC yr⁻¹) CO₂ evasion and DOC discharge based on linear regression over the whole simulation period, it appears that the rate of increase of both fluxes has been strikingly similar over the simulated 20th Century, with mean increases of 11.1 GgC yr⁻¹ and 11.5 GgC yr⁻¹ per year for evasion and export, respectively. A summary and evaluation of the source and seasonal heterogeneity of evasion is discussed in the Text S7.

As previously discussed, the proportion of total basin-wide CO₂ evasion attributable to...
headwater streams and rivers is substantially greater than their proportion of total basin surface area. Figure 6b represents the mean monthly fractional contribution of each surface hydrological water pool to the total evasion flux (unitless) over the period 1998-2007. This shows that over the entirety of the thaw period, the stream water pool takes over from the river water pool as the dominant evasion source, particularly at the height of the freshet period, where its fractional contribution rises to >75%. The stream fraction of August outgassing is ~57% of the annual total, which is higher than the ~40% found for streams in Denfeld et al. (2013). However, the values between the two studies are not directly comparable, different basins notwithstanding due to differences between how 'streams' are defined in the model and in the field (expanded on in Text S8). Also shown in Fig. 6b is the gradual onset of evasion from the floodplain reservoir in April, as the meltwater driven surge in river outflow leads to soil inundation and the gradual increase of proportional evasion from these flooded areas over the course of the summer, with peaks in June-August as water temperatures over these flooded areas likewise peak. We stress the importance of these simulation results as they concur with large numbers of observational studies (cited above) which show smaller headwater streams’ disproportionately large contribution to total outgassing (Fig. 7c), this being due to their comparatively high outgassing rates (Fig. 7e). In addition, the contribution of floodplains to evasion, an otherwise rarely studied feature of high latitude biomes, is shown here to be significant. A Hovmöller plot (Fig. 7d) of the monthly longitude-averaged stream reservoir fraction of total evasion, allows us to infer that: (i) The dominance of stream evasion begins in the most southern upstream headwaters in the lower latitude thaw period (April-May), and trickles northward over the course of the next two months, following the riverflow. (ii) The intensity of this evasion is greatest in the lower latitude regions of the basin, which we speculate is the result of higher temperatures causing a greater proliferation of small thaw water-driven flows and evasion. (iii) Areas where the stream fraction is not dominant or only briefly dominant during the summer (58-60°N, 63-64°N, 70-71°N) are all areas where floodplain CO$_2$ evasion plays a prominent role at that latitudinal band.

We evaluate the approximate rate of modelled areal CO$_2$ efflux from the water surface against observations from Denfeld et al. (2013). (The ‘approximate’ caveat is treated in the Supplementary Text S9). The comparison of simulated results with those from Denfeld et al. (2013) are displayed in Fig. 6d, which shows boxplots for simulated CO$_2$ evasion from the stream water reservoir and river water reservoir averaged over 1998-2007. The empirical (Kolyma river) analogue of this data, from which this plot is inspired (Fig 4d in Denfeld et al., 2013), is shown in inset. Median efflux was 1.1 (6) versus 0.4 (0.8) for stream and river, respectively, in simulations (observations). Like the observations, simulated stream efflux had a substantially greater interquartile range, mean (24.6) and standard deviation (73) than total river efflux (1.3 and 7.2, respectively).

4.5. Emergent Phenomena: DOC and topographic slope, MAAT

Subsurface water infiltration fluxes and transformations of dissolved matter represent an important, if poorly understood and observationally under-represented biogeochemical pathway of DOC export to river main stems, involving the complex interplay of slope, parent material, temperature, permafrost material age and soil physical-chemical processes, such as adsorption and priming. In the Lena basin, as in...
other permafrost catchments, topographic slope has been shown to be a powerful predictor for water infiltration depth, and concentration and age of DOC (Jasechko et al., 2016; Kutscher et al., 2017; McGuire et al., 2005), with deeper flow paths and older, lower DOC-concentrated waters found as the topographic slope increases. This relationship was shown in Fig. 4 of Kutscher et al. (2017) who surveyed DOC concentrations across a broad range of slope angle values in the Lena basin and found a distinct negative relationship between the two. Comparing the Kutscher et al. (2017) values with our model output, by plotting stream and river DOC concentrations averaged per gridpoint over 1998-2007 against the topographic map used in the routing scheme (Fig. 5) we find a similar negative relationship between the two variables. The causes of this relationship and a discussion of the model’s ability to represent it are discussed in Supplementary Text S10. A positive, non-linear relationship between DOC and mean annual air temperature (MAAT), discussed in prior empirical studies, is also reproduced by the model (Fig. 7) and discussed in the Supplement Text S11.

### 4.6 DOC Reactivity Pools

Here we examine the reactivity of DOC leached from the soil and litter to different hydrological export pools. Surface runoff DOC export is dominated by refractory carbon (Fig. 3), with export rates largely following discharge rates as they drain the basin with an increasing delay when latitude increases. As the thaw period gets underway (April), the fraction of labile carbon in surface runoff DOC increases substantially from south to north, reflecting the hydrologic uptake of the previous year’s un-decomposed high-reactivity organic matter.

Refractory C-dominated drainage DOC export (Fig. 9) is highest in June through October, with refractory export rate intensities per latitudinal band during this period consistent with the fraction of inundated area (Fig. S1b) over these bands during the year. The high refractory proportion of drainage flow is expected, as drainage leaches older, relict soil and litter matter. Because of its longer residence time within the soil column, labile DOC carried downward via soil infiltration will tend to be metabolised in situ before it can be exported to the hydrological network, further increasing the proportion of refractory carbon. By contrast floodplain DOC export (Fig. 9) is composed of more nuanced mix of both reactivity classes, reflecting its relatively greater dependence on the current year’s ‘fresh’ biomass as source material (62% labile DOC versus 38% refractory DOC, year-averaged) for carbon leaching.

For both the river and stream pool, mean DOC concentrations are dominated by refractory carbon sources. When averaged over the year, the dominance of the refractory DOC carbon pool over its labile counterpart is also evident for all DOC inputs to the hydrological routing except for floodplain inputs, as well as within the ‘flowing’ stream and river pools themselves. This is shown in Table 2, where the year-averaged percentage of each carbon component of the total input or reservoir is subdivided between the ‘North’ and ‘South’ of the basin, these splits being arbitrarily imposed as the latitudinal mid-point of the basin itself (63N). This reinforces the generalised finding from our simulations that refractory carbon dominates runoff and drainage inflows to rivers (89% refractory, on average), while floodplains export mostly labile DOC to the basin (64%), these values being effectively independent of this latitudinal sub-division (Table 2). Nonetheless, there is a small consistent difference between North and South
in stream and river water DOC makeup, in that the labile portion decreases between North and South; this may be an attenuated reflection of the portion of labile DOC that is decomposed to CO$_2$ within the water column during its transport northward, affecting the bulk average proportions contained within the water in each ‘hemisphere’.

5 Discussion

5.1 Land-Ocean Aquatic Continuum (LOAC)

5.1.1 LOAC Fluxes

Overall, our simulation results show that dissolved carbon entering the Lena river system is significantly transformed during its transport to the ocean. Taking the average throughput of carbon into the system over the last ten years of our simulation, our results show that whereas 7 TgC yr$^{-1}$ (after reinfiltration following flooding of 0.45 TgC yr$^{-1}$; see Fig. 2 ‘Return’ flux) of carbon enters the Lena from terrestrial sources as dissolved carbon and CO$_2$ only 3.4 TgC yr$^{-1}$ is discharged into the Laptev Sea and beyond from the river mouth. The remainder (3.6 TgC yr$^{-1}$) is metabolised in the water column during transport and evaded to the atmosphere (bottom panel, Fig. 10). The terrestrial DOC inflow estimate is comparable to that made by Kicklighter et al. (2013), who estimated in a modelling study terrestrial dissolved carbon loading of the Lena is $\sim$7.7 TgC yr$^{-1}$.

The relative quantities of carbon inflow, evasion and outflow in the river system that are presented for the Lena in Fig. 10 can be compared to the same relative quantities – that is, the ratios of evasion:in and out:in, where ‘in’ refers to dissolved terrestrial input, – from the global study by Cole et al. (2007), who estimated these fluxes from empirical or empirically-derived data at the global scale. This is shown in the top panel of Fig. 10, where we simplify the Cole et al. (2007) data to exclude global groundwater CO$_2$ flux from the coast to the ocean (because our basin mask has a single coastal pixel whereas coastal groundwater seepage is distributed along the entire continental boundary) and the POC fraction of in-river transport and sedimentation (since ORCHIDEE MICT lacks a POC erosion/sedimentation module) from their budget.

This gives global terrestrial dissolved carbon input of 1.45 PgC yr$^{-1}$, 0.7 PgC of which is discharged to the ocean, and the other 0.75 PgC evaded to the atmosphere. Taking the previously mentioned [evasion:in] and [out:in] ratios as a percentage, the outflow and evasion fluxes for the Lena versus the global aggregate are remarkably similar, at 48.6 vs. 48.3% and 51.4 vs 51.7%, for the two respective flows. Thus our results agree with the proposition that the riverine portion of the ‘land-ocean aquatic continuum’ (Regnier et al., 2013) or ‘boundless carbon cycle’ (Battin et al., 2009) is indeed a substantial reactor for matter transported along it. The drivers of changes in CO$_2$ and DOC export from the soil over the simulation period (temperature and precipitation versus CO$_2$), which we extract from our constant climate and CO$_2$ factorial simulations discussed in the Simulation Setup, are similar, if somewhat dominated by temperature (Text S12).

5.1.2 LOAC export flux considerations

Despite our simulations’ agreement with observations regarding the proportional fate of terrestrial DOC inputs as evasion and marine export (Fig. 10), our results suggest
substantial and meaningful differences in the magnitude of those fluxes relative to NPP in the Lena, compared to those estimated by other studies in temperate or tropical biomes. Our simulations’ cumulative DOC and CO₂ export from the terrestrial realm into inland waters is equivalent to ~1.5 % of NPP.

This is considerably lower than Cole et al. (2007) and Regnier et al. (2013) who find lateral transfer to approximate ~5% (1.9PgC yr⁻¹) of NPP at the global scale, while Lauerdal et al. (2017) found similar rates for the Amazon. The cause of this discrepancy with our results is beyond the scope of this study to definitively address, given the lack of tracers for carbon source and age in our model. Nonetheless, our analysis leads us to hypothesise the following.

Temperature limitation of soil microbial respiration at the end of the growing season (approaching zero by October, SI Fig. S5d) makes this flux negligible from November through May (SI Fig. S4d). In late spring, mobilisation of organic carbon is performed by both microbial respiration and leaching of DOC via runoff and drainage water fluxes. However, because the latter are controlled by the initial spring meltwater flux period, which occurs before the growing season has had time to produce litter or new soil carbon (May-June, Fig. 4b), aggregate yearly DOC transport reactivity is characterised by the available plant matter from the previous year, which is overwhelmingly derived from recalcitrant soil matter (Fig. 3) and is itself less available for leaching based on soil carbon residence times.

This causes relatively low leaching rates and riverine DOC concentrations (e.g. Fig. 7), as compared to the case of leaching from the same year’s biological production. Highlighting this point is floodplain domination by labile carbon sourced from that year’s production with a mean DOC concentration of 12.4 mgC L⁻¹ (1998-2007 average), with mean riverine DOC concentrations around half that value (6.9 mgC L⁻¹). Nonetheless the May-June meltwater pulse period dominates aggregate DOC discharge. As this pulse rapidly subsides by late July, so does the leaching and transport of organic matter. Warmer temperatures come in conjunction with increased primary production and the temperature driven soil heterotrophic degradation of contemporary and older matter (via active layer deepening). These all indicate that transported dissolved matter in rivers, at least at peak outflow, is dominated by sources originating in the previous year’s primary production, that was literally ‘frozen out’ of more complete decomposition by soil heterotrophs.

Further, we infer from the fact that all of our simulation grid cells fall within areas of low (< -2°C) MAAT, far below the threshold MAAT (>3°C) proposed by Laundon et al. (2012) for soil respiration-dominated carbon cycling systems (Fig. 7), that the Lena is hydrologically-limited with respect to DOC concentration and its lateral flux. Indeed, the seasonal discharge trend of the Lena –massive snowmelt-driven hydrological and absolute DOC flux, coupled with relatively low DOC concentrations at the river mouth (Fig. 4b, simulation data of Fig. 7), are in line with the Laundon et al. (2012) typology.

We therefore suggest that relatively low lateral transport relative to primary production rates (e.g. as a percentage of net primary production, (%NPP)) in our simulations compared to the lateral transport : NPP percentages reported from the literature in other biomes is driven by meltwater (vs. precipitation) dominated DOC mobilisation,
which occurs during a largely pre-litter deposition period of the growing season. DOC is then less readily mobilised by being sourced from recalcitrant matter, leading to low leaching concentrations relative to those from labile material. As discharge rates decline, the growing season reaches its peak, leaving carbon mobilisation of fresh organic matter to be overwhelmingly driven by in situ heterotrophic respiration.

While we have shown that bulk DOC fluxes scale linearly to bulk discharge flows (Fig. 3d), DOC concentrations (mg C L\(^{-1}\)) hold a more complex and weaker positive relationship with discharge rates, with correlation coefficients (R\(^2\)) of 0.05 and 0.25 for river and stream DOC concentrations, respectively (Fig. 13). This implies that while increasing discharge reflects increasing runoff and an increasing vector for DOC leaching, particularly in smaller tributary streams, by the time this higher input of carbon reaches the river main stem there is a confounding effect of dilution by increased water fluxes which reduces DOC concentrations, explaining the difference between stream and river discharge vs. DOC concentration regressions in the Figure. Thus, and as a broad generalisation, with increasing discharge rates we can also expect somewhat higher concentrations of terrestrial DOC input to streams and rivers. Over the floodplains, DOC concentrations hold no linear relationship with discharge rates (R\(^2\)=0.003, SI Fig. S11), largely reflecting the fact that DOC leaching is here limited by terrestrial primary production rates more than by hydrology. To the extent that floodplains fundamentally require flooding and hence do depend on floodwater inputs at a primary level, we hypothesise that DOC leaching rates are not limited by that water input, at least over the simulated Lena basin.

As discussed above simulated DOC and CO\(_2\) export as a percentage of simulated NPP over the Lena basin was 1.5% over 1998-2007. However, this proportion appears to be highly dynamic at the decadal timescale. As shown in Fig. S12, all lateral flux components in our simulations increased their relative throughput at a rate double to triple that of NPP or respiration fluxes over the 20\(^{th}\) century, also doing so at a rate substantially higher than the rate increase in discharge. In addition, differentials of these lateral flux rates with the rates of their drivers (discharge, primary production) have on average increased over the century (Fig. S12). This suggests that there are potential additive effects of the production and discharge drivers of lateral fluxes that could lead to non-linear responses to changes in these drivers as the Arctic environment transforms, as suggested by the Laudon et al. (2012) data plotted in Fig. 4. Acceleration of the hydrological cycle compounded by temperature and CO\(_2\) -driven increases in primary production could therefore increase the amount of matter available for leaching, increase the carbon concentration of leachate, and increase the aggregate generation of runoff to be used as a DOC transport vector. Given that these causal dynamics apply generally to permafrost regions, both low lateral flux as %NPP and the hypothesised response of those fluxes to future warming may be a feature particular to most high latitude river basins.

6. Conclusion

This study has shown that the new DOC-representing high latitude model version of ORCHIDEE, ORCHIDEE MICT-LEAK, is able to reproduce with reasonable accuracy modern concentrations, rates and absolute fluxes of carbon in dissolved form, as well as the relative seasonality of these quantities through the year. When combined with a
reasonable reproduction of real-world stream, river and floodplain dynamics, we
demonstrate that this model is a potentially powerful new tool for diagnosing and
reproducing past, present and potentially future states of the Arctic carbon cycle. Our
simulations show that of the 34 TgC yr⁻¹ remaining after GPP is respired autotrophically
and heterotrophically in the Lena basin, over one-fifth of this captured carbon is
removed into the aquatic system. Of this, over half is released to the atmosphere from
the river surface during its period of transport to the ocean, in agreement with previous
empirically-derived global-scale studies. Both this transport and its transformation are
therefore non-trivial components of the carbon system at these latitudes that we have
shown are sensitive to changes in temperature, precipitation and atmospheric CO₂
concentration. Our results, in combination with empirical data, further suggest that
changes to these drivers—in particular climate—may provoke non-linear responses in
the transport and transformation of carbon across the terrestrial-aquatic system’s
interface as change progresses in an Arctic environment increasingly characterised by
amplified warming.

**Code and data availability**

The source code for ORCHIDEE MICT-LEAK revision 5459 is available via

Primary data and scripts used in the analysis and other supplementary information that
may be useful in reproducing the author’s work can be obtained by contacting the
responding author.

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**Authors’ contribution**

SB coded this model version, conducted the simulations and wrote the main body of the
paper. RL gave consistent input to the coding process and made numerous code
improvements and bug fixes. BG advised on the inclusion of priming processes in the
model and advised on the study design and model configuration; DZ gave input on the
modelled soil carbon processes and model configuration. PR contributed to the
interpretation of results and made substantial contributions to the manuscript text. MG,
AT and AD contributed to improvements in hydrological representation and floodplain
forcing data. PC oversaw all developments leading to the publication of this study. All
authors contributed to suggestions regarding the final content of the study.

**Competing interests**

The authors declare no competing financial interests.

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References:


Lara, R. J., Rachold, V., Katner, G., Hubberten, H. W., Guggenberger, G., Skoog, A. and


Tables and Figures:

Table 1: Summary describing of the factorial simulations undertaken to examine the relative drivers of lateral fluxes in our model.

<table>
<thead>
<tr>
<th>Simulation Name</th>
<th>Abbreviation</th>
<th>Historical Input Data</th>
<th>Input* Held Constant</th>
</tr>
</thead>
<tbody>
<tr>
<td>Control</td>
<td>CTRL</td>
<td>Climate, CO2, Vegetation</td>
<td>None</td>
</tr>
<tr>
<td>Constant Climate</td>
<td>CLIM</td>
<td>CO2, Vegetation</td>
<td>Climate</td>
</tr>
<tr>
<td>Constant CO2</td>
<td>CO2</td>
<td>Climate, Vegetation</td>
<td>CO2 (Pre-industrial)</td>
</tr>
</tbody>
</table>

*Historically-variable input
Table 2: Summary of the average carbon reactivity types comprising the hydrological inputs to rivers and streams (runoff, drainage and floodplain inputs), and within the rivers and streams themselves, subdivided between the 'North' and 'South' of the Lena basin (greater or less than 63N, respectively).

<table>
<thead>
<tr>
<th>Hydrological Source</th>
<th>Model Carbon Reactivity Pool</th>
<th>North</th>
<th>South</th>
</tr>
</thead>
<tbody>
<tr>
<td>Runoff Input</td>
<td>Refractory</td>
<td>81%</td>
<td>83%</td>
</tr>
<tr>
<td></td>
<td>Labile</td>
<td>19%</td>
<td>17%</td>
</tr>
<tr>
<td>Drainage Input</td>
<td>Refractory</td>
<td>96%</td>
<td>94%</td>
</tr>
<tr>
<td></td>
<td>Labile</td>
<td>4%</td>
<td>6%</td>
</tr>
<tr>
<td>Flood Input</td>
<td>Refractory</td>
<td>36%</td>
<td>37%</td>
</tr>
<tr>
<td></td>
<td>Labile</td>
<td>64%</td>
<td>63%</td>
</tr>
<tr>
<td>Streams</td>
<td>Refractory</td>
<td>91%</td>
<td>89%</td>
</tr>
<tr>
<td></td>
<td>Labile</td>
<td>9%</td>
<td>11%</td>
</tr>
<tr>
<td>Rivers</td>
<td>Refractory</td>
<td>92%</td>
<td>90%</td>
</tr>
<tr>
<td></td>
<td>Labile</td>
<td>8%</td>
<td>10%</td>
</tr>
</tbody>
</table>

Figure 1: Flow diagram illustrating the step-wise stages required to set up the model, up to and including the historical period. The two stages that refer to the inverted reading of restart soil profile order point to the fact that the restart inputs from ORCHIDEE-MICT are read by our model in inverse order, so that one year must be run in
which an activated flag reads it properly, before the reading of soil profile restarts is re-inverted for all subsequent years.

**Figure 2:** Schematic diagrams detailing the major yearly carbon flux outputs (TgC yr\(^{-1}\)) from the Control simulation averaged over the period 1998-2007 as they are transformed and transported across the land-aquatic continuum.
River water discharge,
Log (m$^3$ s$^{-1}$)

APRIL
JUNE
SEPTEMBER

Kusun
Zhigansk

Verbovansk Range

Yakutsk

Kodar Range
Suntar-Khayata Range

Lena Delta

Mean Basin Slope (%)

River Discharge Differential (m$^3$ s$^{-1}$)

Difference (Sim. - Obs.) River Discharge
% Diff. (Sim. - Obs.) River Discharge

River Discharge Differential (%)

0%
50%
100%
150%
200%
250%

-250%
-200%
-150%
-100%
-50%
0%
2000
6000
10000
14000
18000
22000
Figure 3: Map of the Lena (a) with the scale bar showing the mean grid cell topographic slope from the simulation, and the black line the satellite-derived overlay of the river main stem and sub-basins. Mountain ranges of the Lena basin are shown in orange. Green circles denote the outflow gridcell (Kusur) from which our simulation outflow data are derived, as well as the Zhigansk site, from which out evaluation against data from Raymond et al. (2007) are assessed. The regional capital (Yakutsk) is also included for geographic reference. Coastal outline and inland water bodies are shown as dashed red and solid black lines, respectively. (b) Maps of river water discharge (log(m³s⁻¹)) in April, June and September, averaged over 1998-2007. (c) The mean monthly river discharge differential between observed discharge for the Lena (Ye et al., 2009) and simulated discharge averaged over 1998-2007, in absolute (m³ s⁻¹) and percentage terms. (d) Regression of simulated monthly DOC discharge versus simulated river discharge at the river mouth (Kusur) over the entire simulation period (1901-2007). (e) Summed yearly lateral flux versus NPP values for DOC discharge, CO₂ discharge and CO₂ evasion (FCO₂) over the entire simulation period, with linear regression lines shown.
Empirical Raymond et al. (2007)
Sim. DOC Discharge (Aug. 1998-2007)
Empirical Lara et al. (1998)
Sim. DOC Discharge (30-Yr. Run. Avg.)
Sim. DOC Discharge (Yearly)
Empirical est. Dolman et al. (2012)
Empirical Holmes et al. (2012)
Figure 4: (a) Yearly DOC discharged from the Lena river into the Laptev sea is shown here in tC yr⁻¹, over the entire simulation period (dashed red line), with the smoothed, 30-year running mean shown in asterisk. Observation based estimates for DOC discharge from Lara et al. (1998), Raymond et al. (2007), Dolman et al. (2012) and Holmes et al. (2012) are shown by the horizontal black, green triangle, blue diamond and yellow circle line colours and symbols, respectively, and are to be compared against the simulated mean over the last decade of simulation (1998-2007, horizontal red line), with error bars added in grey displaying the standard deviation of simulated values over that period. (b) Average monthly DOC discharge (solid red, tC month⁻¹) and water discharge (dashed red, m³ s⁻¹) to the Laptev Sea over the period averaged for 1901-1910 (circles) and 1997-2007 (squares) are compared, with modern maxima closely tracking observed values. Observed water discharge over 1936-2000 from R-ArcticNet v.4 (Lammers et al., 2001) and published in Ye et al. (2009) are shown by the dashed black line. (c) Observed (black) and simulated (red) seasonal DOC fluxes (solid lines) and CO₂ discharge concentrations (dashed lines). Observed DOC discharge as published in Raymond et al. (2007) from 2004-2005 observations at Zhigansk, a site ~500km upstream of the Lena delta. This is plotted against simulated discharge for: (i) the Lena delta at Kusur (red circles) and (ii) the approximate grid pixel corresponding to the Zhigansk site (red squares) averaged over 1998-2008. Observed CO₂ discharge from a downstream site (Cauwet & Sidorov, 1996; dashed black), and simulated from the outflow site (dashed circle) and the basin average (dashed square) are shown on the log-scale right-hand axis for 1998-2008.
Figure 5: (a) Simulated and observed (Arctic-GRO/Holmes et al., 2012) DOC concentration seasonality for the Lena basin over the period 1999-2007. (b) Plots of DOC concentration versus river discharge as in observations (Raymond et al., 2007) and simulations, where simulations data points are monthly averages taken over the period 1999-2007.
Figure 6: CO$_2$ evasion from stream, river, flood reservoirs. (a) Timeseries of total yearly CO$_2$ evasion (tC yr$^{-1}$) summed over the three hydrological pools (red line) with
the 30-year running mean of the same variable overlain in thick red (asterisk). Error bars give the standard deviation of each decade (e.g. 1901-1910) for each data point in that decade. (b) The fraction of total CO$_2$ evasion emitted from each of the hydrological pools for the average of each month over the period 1998-2007 is shown for river, flood and stream pools (blue, green and red lines, respectively), with error bars depicting the standard deviation of data values for each month displayed. (c) Hovmöller diagram showing the monthly evolution of the stream pool fraction (range 0-1) per month and per latitudinal band, averaged over the period 1998-2007. (d) Boxplot for approximate (see text) simulated CO$_2$ evasion (gC m$^{-2}$ d$^{-1}$) from the streamwater reservoir and river water reservoir averaged over 1998-2007. Coloured boxes denote the first and third quartiles of the data range, internal black bars the median. Whiskers give the mean (solid red bar) and standard deviation (dashed red bar) of the respective data. Empirical data on these quantities using the same scale for rivers, streams and mainstem of the Kolyma river from Denfeld et al., 2013 are shown inset.

**Figure 7:** Mean summertime DOC concentrations (mg C L$^{-1}$) plotted against mean annual air temperature (MAAT, °Celsius) for simulated pixels over the Lena river basin (red circles), and observations for largely peat-influenced areas in western Siberia as reported in Frey et al., 2009 (black crosses), and observations from a global non-peat temperate and high latitude meta-analysis (black circles) reported in Laudon et al. (2012). The blue region represents permafrost-affected areas, while the orange region represents permafrost-free areas. The green region bounds the area of overlap in MAAT between the observed and simulated datasets. The dark red shaded area corresponds to the MAAT ‘zone of optimality’ for DOC production and transport proposed by Laudon et al.
al. (2012). Regression curves of DOC against MAAT for each of the separate datasets are shown for each individual dataset.

**Figure 8:** Variation of DOC concentrations versus topographic slope in Kutscher et al., 2017 (black triangles) and (red dots) as simulated and averaged for the summer months (JJA) over 1998-2007; observed values were measured during June and July 2012-2013.

(a)
Figure 9: The mean monthly fraction of each hydrological pool's (runoff, drainage, floodplains) carbon reactivity constituents (labile and refractory) averaged across the simulation area over 1998-2008.
**Figure 10:** Simplified ‘leaky pipe’ diagram representing the transport and processing of DOC within the land-ocean hydrologic continuum. The scheme template is taken from Cole et al. (2007), where we reproduce their global estimate of DOC and non-groundwater discharge portion of this flow in the top panel (PgC yr⁻¹), and the equivalent flows from our Lena basin simulations in TgC yr⁻¹ in the bottom panel. Thus easy comparison would look at the relative fluxes within each system and compare them to the other.

**Figure 13:** Simulated basin-mean annual DOC concentrations (mg L⁻¹) for the stream and river water pools regressed against mean annual simulated discharge rates (m³ s⁻¹) at Kusur over 1901-2007. Linear regression plots with corresponding R² values are shown.

Supprimé: (b-c): Schematic diagrams detailing the major yearly carbon flux outputs from simulations averaged over the period 1998-2007 as they are transformed and transported across the land-aquatic continuum. Figures (b) and (c) give the same fluxes as a percentage difference from the Control (CTRL-Simulation), for the constant climate and CO₂ simulations, respectively.
Given that DOC fluxes are almost directly proportional to river discharge in the Lena basin (Fig. 3d), this sub-optimal performance with regard to hydrology during August to October seeming to be the main cause of a substantial underestimation in simulated bulk DOC outflow. Another cause may simply be the lack of peat representation in the model, for which DOC flux concentrations in outflowing fluvial water can be very high (e.g. Frey et al., 2005; 2009: see Section 4.5.1).

Deficiencies in modelled hydrology correspond to those found in Fig. 12 of Guimberteau et al. (2018), indicating that the modifications made in this model version, which focus on the DOC cycle, have not further degraded the hydrological performance of the model, the causes of which are described below. Low simulated discharge for the Lena basin, particularly during the late summer and autumn, is consistent with prior, Pan-Arctic simulations conducted by Guimberteau et al. (2018), who ran ORCHIDEE-MICT using both the GSWP3 and CRU-NCEP v7 datasets and evaluated them over the period 1981-2007. Despite the substantially better hydrological performance of ORCHIDEE under GSWP3 climate, they described a near-systematic underestimation of summer/autumn discharge rates for both datasets over the Yukon, Mackenzie, Lena and Kolyma basins. Furthermore, the discrepancy of model output between climatological datasets was almost as large as the discrepancy between model output and observational data in that study, which analysed this in great depth, suggesting that the source of error is both a covariate of model process representation and parameterisation, as well as the climatological datasets themselves. Model hydrological representation and empirically derived climate input data are then subject to interaction with modelled soil (e.g. infiltration), vegetation (e.g. canopy interception) and thermodynamics (e.g. freezing and consequent partitioning of water transport) from which river discharge is computed, confounding full interpretation of sources of bias, briefly described below.

Model process deficiency in this regard was identified by Guimberteau et al. (2018) as residing in an overly restrictive representation of water impermeability through frozen topsoil, which decreases the residence time of running water by directing it to surface runoff rather than subsurface flow, and in the process increases the susceptibility of the total water volume to evapotranspiration from incoming shortwave radiation. This would bias both the timing (over-partitioning of water to high runoff periods) and volume of water (low bias) reaching the river stem and its eventual discharge into the ocean, respectively, as demonstrated by model output. Guimberteau et al. (2018) suggest that representation of sub-grid-scale infiltration mechanisms under frozen conditions, such as
soil freezing-drying that would enhance infiltration, be included in future, yet-to-be implemented iterations of ORCHIDEE. Furthermore, we suggest that the lack of representation of lakes in ORCHIDEE, which serve to increase the time lag between precipitation/melt and oceanic discharge, may likewise be a powerful source of bias in the timing of discharge fluxes represented by the model.

Unsurprisingly, simulated surface runoff has been shown to be strongly affected by differences in precipitation between datasets (Biancamaria et al., 2009; Fekete et al., 2004), while biases in these and evapotranspiration datasets that are used to both drive and evaluate the hydrological models, are a powerful source of water balance biases in high-latitude basins (Wang et al., 2015). Indeed, climatological dataset estimates for the spatial distribution of high latitude winter snowfall are generally problematic, owing to the low density of meteorological stations (Burke et al., 2013), wind-related issues with in-field collection and measurement that lead to systematic underestimates of snowfall rates (Yang et al., 2005), creating biases in the climatological datasets that only show up when the integrator of their model input -in this case river discharge -is modelled. In addition, the wintertime partitioning of precipitation between rain and snow, a function of 2m air temperatures in the forcing datasets, strongly affects the volume and timing of runoff (Guimberteau et al., 2018; Haddeland et al., 2011). Indeed, 69% of the spatial variance of the spring freshet has been attributed to snow water-equivalent bias during the pre-melt season (Rawlins et al., 2007). In addition, errors in forcing of soil evaporation due to inaccuracies in incoming shortwave radiation, as well as biases in the parameterisation of canopy interception -a function of simulated LAI -can lead to upward biases in evapotranspiration rates (Guimberteau et al., 2018).

Firstly, there is a quasi-linear positive relationship between DOC discharge and river discharge (Fig. 3d). This relation is common to Arctic rivers, as DOC loading experiences disproportionately large increases with increases in discharge (Fig. 4, Raymond et al., 2007), owing largely to the ‘flushing’ out of terrestrially fixed carbon from the previous year’s production by the massive runoff generated by ice and snow melt during the spring thaw. Comparing simulated annual mean discharge rate (m\(^3\) s\(^{-1}\)) with long-term observations (Ye et al. 2003) over years 1940-2000 (Fig. 4c) shows that though absolute discharge rates are underestimated by simulations, their interannual variation reasonably tracks the direction and magnitude of observations. Linear regressions through each trend yield very similar yearly increases of 29 vs 38 m\(^3\) s\(^{-1}\) yr\(^{-1}\) for simulations and observations, respectively. The observed vs. simulated mean annual water discharge differential hovers at 36% (Figs. 3d, 4c), close to the 43% differential between observed and simulated DOC discharge, giving some indication that, given the linear relationship between water and DOC discharge, most of the DOC discrepancy can be explained by the performance of the hydrology and not the DOC module, the latter of which was the subject of developments added in ORCHIDEE M-L. Applying the regression slope of the relationship in Fig. 3d (9E-06 mgC per m\(^3\)s\(^{-1}\)) to the mean river discharge discrepancy of 36%, we find that 84% of the
differential between observed and simulated discharge can be explained by the underperformance of the hydrology module.

Further sources of error are process exclusion and representation/forcing limitations. Indeed, separate test runs carried out using a different set of climatological input forcing show that changing from the GSWP3 input dataset to input from bias-corrected projections from the IPSL Earth System Model under the second Inter-Sectoral Impact Model Intercomparison Project (ISIMIP2b (Frieler et al., 2017; Lange, 2016, 2018)) protocol increases DOC discharge to the ocean to 4.14 TgC yr⁻¹ (+37%), largely due to somewhat higher precipitation rates in that forcing dataset (see Table S3). Thus, the choice of input dataset itself introduces a significant degree of uncertainty to model output.

In addition, this model does not include explicit peatland formation and related dynamics, which is the subject of further model developments (Qiu et al., 2018) yet to be included in this iteration. With peatlands thought to cover ~17% of the Arctic land surface (Tarnocai et al., 2009), and with substantially higher leaching concentrations, this may be a significant omission from our model. The remaining biases likely arise from errors in the interaction of simulated NPP, respiration and DOC production and decomposition, which will impact on the net in and out-flow of dissolved carbon to the fluvial system. However, the DOC relationship with these variables is less clear-cut than with river discharge. Indeed, regressions (Fig. 3e) of annual DOC versus NPP (TgC yr⁻¹) show that DOC is highly sensitive to increases in NPP, but is less coupled to it (more scattered, R²=0.42) than other simulated fluvial carbon variables shown, i.e. aquatic CO₂ evasion and soil CO₂ export to the river network. The differences in correlation and slope of the variables in Fig. 3e are expected: aquatic CO₂ evasion is least sensitive yet most tightly coupled to NPP (R²=0.52), while CO₂ export to rivers is intermediate between the two (R²=0.43). The greater correlation with NPP of DOC compared to evasion is understandable, given that DOC leaching is a covariate of both NPP and runoff, whereas evasion flux is largely dependent on organic inputs (production) and temperature (see Part 1).

4.2.3 Model Evaluation: DOC Discharge Seasonality

Figure 4b shows that the bulk of the DOC outflow occurs during the spring freshet or snow/ice-melting period of increased discharge, accounting for ~50-70% of the total Arctic outflow (Lammers et al., 2001; Ye et al., 2009), with peak water discharge rates in June of ~80,000 m³ s⁻¹. DOC concentrations increase, as meltwater flushes out DOC accumulated from the previous year’s litter and SOC generation (Raymond et al., 2007; Kutscher et al., 2017). This is reproduced in our simulations, since DOC discharge peak occurs at the onset of the growing season, meaning it is generated from a temporally prior stock of organic carbon. Simulation of the hydrological dynamic is presented in maps of river discharge through the basin in Fig. 3b, which show low-flows in April with substantial hydrographic flow from upstream mountainous headwaters and Lake Baikal inflow in the south, peak flow in June dominated by headwaters, and little headwater input in September.

In Fig. 4b we observe the following: (i) DOC discharge fluxes closely track hydrological fluxes. (ii) The simulated modern river discharge peak approximates the historical
observed discharge peak, but slightly overestimates spring fluxes and substantially underestimates fluxes in the autumn, as explained above. (iii) The difference between the first and last decades of the simulation in Fig. 4b is mostly attributable to a large increase in the DOC flux mobilised by spring freshet waters. This suggests both greater peaks in simulated DOC flux and a shift to earlier peak timing, owing to an increase in river discharge, indicative of an earlier spring and a progressively warmer environment over the 20th Century. (iv) The maximum modelled modern monthly DOC flux rate of ~1.3 TgC month$^{-1}$ is comparable to the mean maximum DOC flux rate measured in a recent study (1.75 TgC month$^{-1}$, Kutscher et al., 2017, Fig. 2).

We compare the Raymond et al. (2007) modern DOC outflow (Fig. 4d, solid black line) from the Lena river at Zhigansk (Raymond et al., 2007) against simulated DOC outflow from both Zhigansk and Kusur (Fig. 4d). Simulated DOC flux is underestimated for both sites. Peakflow at Zhigansk seems to be attenuated over May and June in simulations, as opposed to May peakflow in observations. Peakflow at Kusur is definitively in June. This suggests that simulated outflow timing at Zhigansk may slightly delayed, causing a split in peak discharge when averaged in the model output. Thus the aggregation of model output to monthly averages from calculated daily and 30 minute timesteps can result in the artificial imposition of a normative temporal boundary (i.e. month) on a continuous series. This may cause the less distinctive ‘sharp’ peak seen in Fig. 4d, which is instead simulated at the downstream Kusur site, whose distance some 500km away from Zhigansk more clearly explains the delay difference in seasonality. We further evaluate our DOC discharge at the sub-basin scale, to test whether the fractional contribution of different DOC flows from each sub-basin correspond to those in their observed correlates from Kutscher et al., (2017). This comparison is depicted in Fig. S2, where the observed and simulated percentage DOC contributions of the Aldan, Vilui, and Upper and Lower Lena sub-basins to total flux rates are 19 (24)%, 20(10%), 33 (38%) and 30 (28)% in simulations (observations) for the four sub-basins, respectively. While deviations between simulated and observed DOC fluxes can be expected, the nearly twofold value mismatch of the Vilui basin is due to its real-word damming, not represented here. On the other hand, we cannot explain the ~5% discrepancies in other sub-basin fluxes, particularly for the Aldan.

The spatial distribution of DOC concentrations are shown in maps of mean monthly DOC concentration for stream water, river water and groundwater (Fig. 6a,b,c, respectively) in April, June and September. For both the stream and river water reservoirs, DOC concentrations appear to have spatio-temporal gradients correlated with the flux of water over the basin during the thaw period, with high concentrations of 10-15 mgC L$^{-1}$ driven by April meltwaters upstream of the basin, these high concentrations moving northward to the coldest downstream regions of the basin in June. Lower DOC concentrations of ~5 mgC L$^{-1}$ dominate the basin in September when the bulk of simulated lateral flux of DOC has dissipated into the Laptev Sea. In contrast, groundwater DOC concentrations are generally stable with time, although some pixels appear to experience some ‘recharge’ in their concentrations during the first two of the three displayed thaw months. Significantly, highest groundwater DOC concentrations of up to 20 mgC L$^{-1}$ are focussed on the highest
elevation areas of the Lena basin on its Eastern boundary, which are characterized by a dominance of Podzols (SI, Fig. 2b). This region, the Verkhoyansk range, is clearly visible as the high groundwater DOC concentration (2-20 mg C L⁻¹) arc (in red) in Fig. 6a, as well as other high elevation areas in the south-western portion of the basin (Fig. 3a), while the low-lying central basin shows much smaller groundwater DOC concentrations (0-2 mg C L⁻¹). The range of simulated groundwater DOC concentration comes close to those aggregated from the empirical literature by Shvartsev (2008), which finds from >9,000 observations that groundwater in permafrost regions exhibit a mean concentration of ~10 mg C L⁻¹ after peatlands and swamps (not simulated here) are removed (Table 2).

To our knowledge, no direct measurements for CO₂ evasion from the surface of the Lena river are available in the literature. We refer to Denfeld et al. (2013) for evaluating our evasion flux results, since their basin of study, the Kolyma River, is the most geographically proximate existing dataset to the Lena, despite biogeographical differences between the two basins – namely that the Kolyma is almost entirely underlain by continuous permafrost. The Kolyma River CO₂ evasion study measured evasion at 29 different sites along the river basin (~158-163°E; 68-69.5°N), with these sites distinguished from one another as ‘main stem’, ‘inflowing river’ or ‘stream’ on the basis of reach length. The study showed that during the summer low-flow period (August), areal river mainstem CO₂ evasion fluxes were ~0.35 g C m⁻² d⁻¹, whereas for streams of stream order 1-3 (widths 1-19 m), evasion fluxes were up to ~7 g C m⁻² d⁻¹, and for non-mainstem rivers (widths 20-400 m) mean net fluxes were roughly zero (Table 3 of Denfeld et al., 2013). Thus, while small streams have been observed to contribute to roughly 2% of the Kolyma basin surface area, their measured percentage contribution to total basin-wide CO₂ evasion ~40%, whereas for the main stem the surface area and evasion fractions were ~80% and 60%, respectively. Likewise, mean annual evasion rates of <0.8 up to around 7 g C m⁻² d⁻¹ have been found for the Ob and Pur rivers in Western Siberia (Serikova et al., 2018).

Results such as these, in addition to permafrost soil incubation experiments (e.g. Drake et al., 2015; Vonk et al., 2013, 2015b, 2015a) suggest that small streams, which represent the initial (headwater) drainage sites of these basins, rapidly process hydrologically leached carbon to the atmosphere, and that this high-reactivity carbon is a mix of recently thawed ancient permafrost material, as well as decomposing matter from the previous growth year. This is given as evidence that the total carbon processing of high-latitude rivers is significantly underestimated if only mainstem carbon concentrations are used in the accounting framework, since a large amount of carbon is metabolised to the atmosphere before reaching the site of measurement.

The heterogeneity of CO₂ evasion from different sources in the model is most evident in
terms of their geographic distribution and relative intensity, as shown in the evasion flux rate maps over stream and river areas in April, June and September (Fig. 8a-b). Stream evasion (Fig.8a), tends to be broadly distributed over the whole basin, representing the fact that small streams and their evasion are the main hydrologic connectors outside of the main river and tributary grid cells, whereas river evasion (Fig. 8b) is clearly linked to the hydrographic representation of the Lena main stem itself, with higher total quantities in some individual grid cells than for the stream reservoir, yet distributed amongst a substantially smaller number of grid cells. Whereas the stream reservoir has greatest absolute evasion flux rates earlier in the year (April-May), maximum evasion rates occur later in the year and further downstream for the river reservoir, reflecting the fact that headwaters are first-order integrators of soil-water carbon connectivity, whereas the river mainstem and tributaries are of a secondary order.

The spatio-temporal pattern of increasing evasion over the simulation period is shown in Fig. 7b as a Hovmöller difference plot, between the last and first decade, of log-scale average monthly evasion rates per latitudinal band. This shows that the vast majority of outgassing increase occurs between March and June, corresponding to the progressive onset of the thaw period moving northwards over this timespan. Although relatively small, outgassing increases are apparent for most of the year, particularly at lower latitudes. This would suggest that the change is driven most acutely by relatively greater temperature increases at higher latitudes ('Arctic amplification’ of climate warming, e.g. Bekryaev et al., 2010) while less acute but more temporally homogenous evasion is driven by seasonal warming at lower latitudes.

This is because in ORCHIDEE MICT-L, the ‘stream’ water reservoir is water routed to the river network for all hydrologic flows calculated to not cross a 0.5 degree grid cell boundary (the resolution of the routing module, explained in Part 1, Section 2.6), which may not be commensurate with long, <20m width streams in the real-world, that were used in the Denfeld et al. (2013) study. In addition, this ‘stream’ water reservoir in the model does not include any values for width or area in the model, so we cannot directly compare our stream reservoir to the <20m width criterion employed by Denfeld et al. (2013) in their definition of an observed stream. Thus our ‘stream’ water reservoir encompasses substantially greater surface area and hydrologic throughput than that in the Denfeld et al. study.

refers to the fact that model output doesn’t define a precise surface area for the stream water reservoir, which is instead bundled into a single value representing the riverine fraction of a grid cell’s total surface area. To approximate the areal outgassing for the stream versus river water reservoirs, we weight the total non-floodplain inundated area of each grid cell by the relative total water mass of each of the two hydrological pools, then divide the total daily CO₂ flux simulated by the model by this value. The per-pool areal
estimate is an approximation since it assumes that rivers and streams have the same surface area: volume relationship. This is clearly not the case, since streams are generally shallow, tending to have greater surface area per increment increase in depth than rivers. Thus, our areal approximations are likely underestimated (overestimated) for streams (rivers), respectively.

Note that from ~700 non-zero simulation datapoints, 7 were omitted as 'outliers' from the stream reservoir efflux statistics described below, because very low stream:river reservoir values skewed the estimation of total approximate stream surface area values very low, leading to extreme efflux rate values of 1-3000 gC m\(^{-2}\) d\(^{-1}\) and are thus considered numerical artefacts of the areal approximation approach used here.

4.5 Emergent Phenomena

4.5.1 DOC and mean annual air temperature

A key emergent property of DOC concentrations in soils and inland waters should be their positive partial determination by the temperature of the environment under which their rates of production occur, as has been shown in the literature on permafrost regions, most notably in Frey & Smith (2005) and Frey & McClelland (2009).

Increasing temperatures should lead to greater primary production, thaw, decomposition and microbial mobilisation rates, and hence DOC production rates, leading to (dilution effects notwithstanding) higher concentrations of DOC in thaw and so stream waters. Looking at this emergent property allows us to evaluate the soil-level production of both DOC and thaw water at the appropriate biogeographic and temporal scale in our model. This provides a further constraint on model effectiveness at simulating existing phenomena at greater process-resolution.

Figure 9 compares three datasets (simulated and two observational) of riverine DOC concentration (in mgC L\(^{-1}\)) plotted against mean annual air temperature (MAAT). The simulated grid-scale DOC versus MAAT averaged over July and August (for comparability of DOC with observational sampling period) of 1998-2007 is shown in red, and observed data compiled by Laudon et al. (2012) and Frey and Smith (2005) for sites in temperate/cold regions globally and peatland-dominated Western Siberia, respectively. The Laudon et al. (2012) data are taken from 49 observations including MAAT over the period 1997-2011 from catchments north of 43°N, and aggregated to 10 regional biogeographies, along with datapoints from their own sampling; those in the Frey and Smith study are from 55-68°N and ~65-85°E (for site locations, see Laudon et al. (2012), Table 1 and 2; Frey and Smith (2005), Fig. 1).
Fig. 9 can be interpreted in a number of ways. First, this MAAT continuum spans the range of areas that are both highly and moderately permafrost affected and permafrost free (Fig. 9, blue and green versus orange shading, respectively), potentially allowing us a glimpse of the behaviour of DOC concentration as the environment transitions from the former to the latter. Simulated Lena DOC concentrations, all in pixels with MAAT < -2°C and hence all bearing continuous or discontinuous permafrost (‘permafrost-affected’ in the figure), only exhibit a weakly positive response to MAAT on the scale used (y=6.05e-0.03MAAT), although the consistent increase in DOC minima with MAAT is clearly visible. Second, the Laudon et al. (2012) data exhibit an increasing then decreasing trend over the range of MAAT (-2°C to 10°C) in their dataset, which they propose reflects an ‘optimal’ MAAT range (0-3°C) for the production and transport of DOC (Fig. 9, red shading). Below this optimum range, DOC concentrations may be limited by transport due to freezing, and above this, smaller soil carbon pools and temperature-driven decomposition would suppress the amount of DOC within rivers. Third, the lower end of the Laudon et al. (2012) MAAT values correspond to a DOC concentration in line with DOC concentrations simulated by our model. Fourth, DOC concentrations in the Frey and Smith (2005) data exhibit a broad scattering in permafrost-affected sites, with concentrations overlapping those of our simulations (Fig. 9, green shading), before rapidly increasing to very high concentrations relative to the Laudon et al. (2012) data, as sites transition to permafrost-free (red shading, y=3.6MAAT+29.4).

Their data highlight the difference in DOC concentration regime between areas of high (Frey and Smith, 2005) and low (Laudon et al., 2012) peatland coverage and the different response of these to temperature changes. Fifth, because our simulation results largely correspond with the observed data where the MAAT ranges overlap (green shading), and because our model lacks peatland processes, we should expect our model to follow the polynomial regression plotted for the Laudon et al. (2012) data as temperature inputs to the model increase. Figure 9 implies that this increase should be on the order of a doubling of DOC concentration as a system evolves from a MAAT of -2°C to 2°C. With warming, we expect the response of DOC concentrations to reflect a mix of both observationally-derived curves, as a function of peatland coverage.

This relationship was found in temperate rivers by Lauerwald et al. (2012), and in a recent Pan-Arctic synthesis paper Connolly et al. (2018). The reasoning for the negative slope-DOC concentration relationship is that as elevation increases, temperature and primary production decreases. This leads to a thinner organic soil layer, meaning that mineral soil plays a stronger role in shallow hydrologic flowpaths, allowing for deeper infiltration and shorter residence time in a given soil layer. Further, steeper terrain leads to a lower soil water residence time and lower moisture than in flat areas. As a result, a given patch of soil matter will be exposed to leaching for less (residence) time, while the organic matter that is leached is thought to be adsorbed more readily to mineral soil particles, leading to either their re-stabilisation in the soil column or shallow retention and subsequent heterotrophic respiration in situ, cumulatively resulting in lower DOC concentrations in the hydrologic export (Kaiser and Kalbitz, 2012; Klaminder et al., 2011). This line of reasoning was
recently shown to apply also to deep organic permafrost soils (Zhang et al., 2017), although the degree to which this is the case in comparison to mineral soils is as yet unknown.

In addition, and as described in Part 1 (Section 2.5) of this study, MICT-L contains a provision for increased soil column infiltration and lower decomposition rates in areas underlain by Podzols and Arenosols. The map from the Harmonized World Soil Database (Nachtergaele, 2010), which is used as the input to this criterion, shows areas underlain by these soils in the Lena basin to also be co-incident with areas of high topographic slope (Fig. 3a, SI, Fig S3b). The 'Podzol effect' is to increase the rate of decomposition and infiltration of DOC, relative to all other soil types, thus also increasing the rate of DOC flux into groundwater (see Part 1 of this study, Section 2.5). Thus, our modelling framework explicitly resolves the processes involved in these documented dynamics – soil thermodynamics, solid vertical flow (turbation), infiltration as a function of soil textures and types, adsorption as a function of soil parameters (see Part 1 of this study, Section 2.11), DOC respiration as a function of soil temperature and hence depth (Part 1, Section 2.12), and lagging of DOC vertical flow behind hydrological drainage flow (summary Figure in Part 1, Fig. 1). We thus have some confidence in reporting that the simulated negative relationship of DOC concentration with topographic slope may indeed emerge from the model.

5.1.2 LOAC drivers

The constant climate (CLIM) and constant CO2 (CO2) simulations described in Section 3 were undertaken to assess the extent – and the extent of the difference – to which these two factors are drivers of model processes and fluxes. These differences are summarised in Figs. 12(b-c), in which we show the same 1998-2007 – averaged yearly variable fluxes as in the CTRL simulation, expressed as percentages of the CTRL values given in Fig. 2. A number of conclusions can be drawn from these diagrams.

First, all fluxes are lower in the factorial simulations, which can be expected due to lower carbon input to vegetation from the atmosphere (constant CO2) and colder temperatures (constant climate) inhibiting more vigorous growth and carbon cycling. Second, broadly speaking, both climate and CO2 appear to have similar effects on all fluxes, at least within the range of climatic and CO2 values to which they have subjected the model in these historical runs. With regard to lateral export fluxes in isolation, variable climate (temperature increase) is a more powerful driver than CO2 increase (see below). Third, the greatest difference between the constant climate and CO2 simulation carbon fluxes appear to be those associated with terrestrial inflow of dissolved matter to the aquatic network, these being more sensitive to climatic than CO2 variability. This is evidenced by a 49% and 32% decline in CO2 and DOC export, respectively, from the land to rivers in the constant climate simulation, versus a 27% and 23% decline in these same variables in the constant CO2 simulation. Given that the decline in primary production and respiration in both factorial simulations was roughly the same, this difference in terrestrial dissolved input is
attributable to the effect of climate (increased temperatures) on the hydrological cycle, driving changes in lateral export fluxes.

This would imply that at these carbon dioxide and climatic ranges, the modelled DOC inputs are slightly more sensitive to changes in the climate rather than to changes in atmospheric carbon dioxide concentration and the first order biospheric response to this. However, while the model biospheric response to carbon dioxide concentration may be linear, thresholds in environmental variables such as MAAT may prove to be tipping points in the system’s emergent response to change, as implied by Fig. 9, meaning that the Lena, as with the Arctic in general, may soon become much more temperature-dominated with regard to the drivers of its own change.

Table 2: Mean observed groundwater CO$_2$ and DOC concentrations for global permafrost regions subdivided by biogeographic province and compiled by Shvartsev (2008) from over 9000 observations.

<table>
<thead>
<tr>
<th>Permafrost Groundwater Provinces</th>
<th>Swamp</th>
<th>Tundra</th>
<th>Taiga</th>
<th>Average</th>
<th>Average (-Swamp)</th>
</tr>
</thead>
<tbody>
<tr>
<td>CO$_2$ (mgC L$^{-1}$)</td>
<td>12.3</td>
<td>14</td>
<td>10.8</td>
<td>12.4</td>
<td>12.4</td>
</tr>
<tr>
<td>DOC (mgC L$^{-1}$)</td>
<td>17.6</td>
<td>10.1</td>
<td>9.3</td>
<td>12.3</td>
<td>9.7</td>
</tr>
</tbody>
</table>
Mean annual discharge rate (m$^3$s$^{-1}$)

Year

Simulated Discharge

Observed Discharge (Ye et al., 2003)
Figure 6: Maps of (a) DOC concentrations (mgC L\(^{-1}\)) in groundwater ('slow' water pool), (b) stream water pool, (c) river water pool in April, June and September (first to third rows, respectively), averaged over the period 1998-2007. The coastal boundary and a water body overlay have been applied to the graphic in red and black, respectively, and the same scale applies to all diagrams. All maps have the Lena basin area shaded in the background.
Modern increase CO$_2$ Evasion Flux, Log (g m$^{-2}$ d$^{-1}$)

(a) 'Stream' CO$_2$ Evasion (mgC m$^{-2}$ d$^{-1}$)
(b) 'River' CO$_2$ Evasion (mgC m$^{-2}$ d$^{-1}$)
Figure 8: Maps of CO₂ evasion from the surface of the two fluvial hydrological pools in the model, (a) streams and (b) rivers in April, June and September. All maps use the same (log) scale in units of (mgC m⁻² d⁻¹).
Constant CO₂ (% of CTRL)

δCO₂=67%
δDOC=63%
δCO₂=70%
δCO₂=73%

Plant Resp.

Throughfall Flux

Soil Respiration

CO₂ Evasion

CO₂ δDOC=77%

GPP δCO₂=73%

DOC δCO₂=77%

Runoff δGPP=67%

Drainage

[CO₂]

DOC δDOC=67%

DOC δDOC=76%

DOC δDOC=70%

Flooded Soil δCO₂=71%

Terrestrial

Aquatic

Marine
Supplement of

ORCHIDEE MICT-LEAK (r5459), a global model for the production, transport and transformation of dissolved organic carbon from Arctic permafrost regions, Part 2: Model evaluation over the Lena River basin.

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Text S1: Evaluation of Simulated NPP and Soil Respiration

Rates of yearly net primary production (NPP) for Russian and Siberian forests have been inferred in situ from eddy flux and inventory techniques to range from 123-250 gC m⁻² yr⁻¹ (Beer et al., 2006; Lloyd et al., 2002; Roser et al., 2002; Schulze et al., 1999; Shvidenko and Nilsson, 2003). We likewise simulate a broad range of NPP carbon uptake rates, of 61-469 gC m⁻² yr⁻¹ averaged per grid cell over the Lena basin, with a mean value of 210 gC m⁻² yr⁻¹. NPP is heterogeneously distributed over space and between PFTs (SI, Fig. S4c), with forests averaging 90 gC m⁻² yr⁻¹ and grasslands averaging 104 gC m⁻² yr⁻¹ over the basin as a whole. Low values tended to originate in basin grid cells with elevated topography or high mean slope, while the maximum value was standalone, exceeding the next greatest by ~100 gC m⁻² yr⁻¹, and is most likely caused by the edge effects of upscaling a coastal gridcell’s small fraction of terrestrial area where high productivity occurs in a small plot, to the grid cell as a whole. By evaluating NPP we are also evaluating at a secondary level litter production, which is at a third level a major component of DOC production.

Taken as a whole, gross primary production (GPP) was performed under simulations by four PFT groups, with the largest basin-wide bulk contributions coming from boreal needleleaf summer-green trees and C3 grasses (SI, Fig. S4a), the highest GPP uptake rates (3 TgC pixel⁻¹ yr⁻¹) generated by boreal needleleaf evergreen trees, and the remainder of GPP contributed by Boreal broad-leaved summer-green trees (SI, Fig. S4a).

Soil respiration rates, of combined soil heterotroph and plant root respiration in our Control simulation, averaged 208 gC m⁻² yr⁻¹ (0.57 gC m⁻² d⁻¹) over the Lena basin over the period 1990-2000, which is somewhat higher than those found by Elberling (2007) in tundra soils on Svalbard, of 103-176 gC m⁻² yr⁻¹ (0.29-0.48 gC m⁻² d⁻¹), Sawamoto, et al. (2000) measured in situ summertime soil respiration over the central Lena basin and found rates of 1.6-34 gC m⁻² d⁻¹, while Sommerkorn (2008) observed rates of 0.1-3.9 gC m⁻² d⁻¹ at higher latitudes, varying to vary with vegetation and fire history.
water table depth and temperature. Mean heterotrophic respiration rates of 1.6 gC m$^{-2}$ d$^{-1}$ are simulated here during July and August, in the range 0.5-2.2 gC m$^{-2}$ d$^{-1}$ for each of the above PFT groups. The spatial distribution of, and difference in respiration rates between PFT groups largely mirrors those for NPP (SI Fig. S4c), with maximum rates of 1.4 gC m$^{-2}$ d$^{-1}$ over forested sites, versus a maximum of 2.2 gC m$^{-2}$ d$^{-1}$ over grassland/tundra sites (SI, Fig. S4b).

Aggregated over the basin, results show that increases over the course of the 20th Century were simulated for NPP, GPP, River Discharge, DOC, CO$_2$(aq.), autotrophic and heterotrophic respiration and CO$_2$ evasion, with percentage changes in the last versus first decade of +25%, +27%, +38%, +73%, +60%, +30%, +33% and +63%, respectively. (Fig. S12) It thus appears that rising temperatures and CO$_2$ concentrations disproportionately favoured the metabolism of carbon within the soil and its transport and mineralisation within the water column, fed by higher rates of primary production and litter formation as well as an accelerated hydrological cycle.

Text S2: Deficiencies in Modelled Hydrology

Deficiencies in modelled hydrology correspond to those found in Fig. 12 of Guimberteau et al. (2018), indicating that the modifications made in this model version, which focus on the DOC cycle, have not further degraded the hydrological performance of the model, the causes of which are described below. Low simulated discharge for the Lena basin, particularly during the late summer and autumn, is consistent with prior, Pan-Arctic simulations conducted by Guimberteau et al. (2018), who ran ORCHIDEE-MICT using both the GSWP3 and CRU-NCEP v7 datasets and evaluated them over the period 1981-2007. Despite the substantially better hydrological performance of ORCHIDEE under GSWP3 climate, they described a near-systematic underestimation of summer/autumn discharge rates for both datasets over the Yukon, Mackenzie, Lena and Kolyma basins. Furthermore, the discrepancy of model output between climatological datasets was almost as large as the discrepancy between model output and observational data in that study, which analysed this in great depth, suggesting that the source of error is both a covariate of model process representation and parameterisation, as well as the climatological datasets themselves. Model hydrological representation and empirically derived climate input data are then subject to interaction with modelled soil (e.g. infiltration), vegetation (e.g. canopy interception) and thermodynamics (e.g. freezing and consequent partitioning of water transport) from which river discharge is computed, confounding full interpretation of sources of bias, briefly described below.

Model process deficiency in this regard was identified by Guimberteau et al. (2018) as residing in an overly restrictive representation of water impermeability through frozen topsoil, which decreases the residence time of running water by directing it to surface runoff rather than subsurface flow, and in the process increases the susceptibility of the total water volume to evapotranspiration from incoming shortwave radiation. This would bias both the timing (over-partitioning of water to high runoff periods) and volume of water (low bias) reaching the river stem and its eventual discharge into the ocean, respectively, as demonstrated by model output. Guimberteau et al. (2018) suggest that representation of sub-grid-scale infiltration mechanisms under frozen conditions, such as soil freezing-drying that would enhance infiltration, be included in
future, yet-to-be implemented iterations of ORCHIDEE. Furthermore, we suggest that the lack of representation of lakes in ORCHIDEE, which serve to increase the time lag between precipitation/melt and oceanic discharge, may likewise be a powerful source of bias in the timing of discharge fluxes represented by the model.

Unsurprisingly, simulated surface runoff has been shown to be strongly affected by differences in precipitation between datasets (Biancamaria et al., 2009; Fekete et al., 2004), while biases in these and evapotranspiration datasets that are used to both drive and evaluate the hydrological models, are a powerful source of water balance biases in high-latitude basins (Wang et al., 2015). Indeed, climatological dataset estimates for the spatial distribution of high latitude winter snowfall are generally problematic, owing to the low density of meteorological stations (Burke et al., 2013). Wind-related issues with in-field collection and measurement that lead to systematic underestimates of snowfall rates (Yang et al., 2005), creating biases in the climatological datasets that only show up when the integrator of their model input -in this case river discharge -is modelled. In addition, the wintertime partitioning of precipitation between rain and snow, a function of 2m air temperatures in the forcing datasets, strongly affects the volume and timing of runoff (Guimberteau et al., 2018; Haddeland et al., 2011). Indeed, 69% of the spatial variance of the spring freshet has been attributed to snow water-equivalent bias during the pre-melt season (Rawlins et al., 2007). In addition, errors in forcing of soil evaporation due to inaccuracies in incoming shortwave radiation, as well as biases in the parameterisation of canopy interception -a function of simulated LAI -can lead to upward biases in evapotranspiration rates (Guimberteau et al., 2018).

Text S3: Deficiencies in Modelled DOC Discharge

Firstly, there is a quasi-linear positive relationship between DOC discharge and river discharge (Fig. 3d). This relation is common to Arctic rivers, as DOC loading experiences disproportionately large increases with increases in discharge (Fig. 4, Raymond et al., 2007), owing largely to the 'flushing' out of terrestrially fixed carbon from the previous year's production by the massive runoff generated by ice and snow melt during the spring thaw. Comparing simulated annual mean discharge rate (m$^3$/s) with long-term observations (Ye et al. 2003) over years 1940-2000 (Fig. S3) shows that though absolute discharge rates are underestimated by simulations, their interannual variation reasonably tracks the direction and magnitude of observations. Linear regressions through each trend yield very similar yearly increases of 29 vs 38 m$^3$/s yr$^{-1}$ for simulations and observations, respectively. The observed vs. simulated mean annual water discharge differential hovers at 36% (Figs. 3d, 4c), close to the 43% differential between observed and simulated DOC discharge, giving some indication that, given the linear relationship between water and DOC discharge, most of the DOC discrepancy can be explained by the performance of the hydrology and not the DOC module, the latter of which was the subject of developments added in ORCHIDEE M-L. Applying the regression slope of the relationship in Fig. 3d ($9E-06$ mgC per m$^3$/s), to the mean river discharge discrepancy of 36%, we find that 84% of the differential between observed and simulated discharge can be explained by the underperformance of the hydrology module.

Further sources of error are process exclusion and representation/forcing limitations. Indeed, separate test runs carried out using a different set of climatological input forcing
show that changing from the GSWP3 input dataset to input from bias-corrected projections from the IPSL Earth System Model under the second Inter-Sectoral Impact Model Intercomparison Project (ISIMIP2b [Frieler et al., 2017; Lange, 2016, 2018]) protocol increases DOC discharge to the ocean to 4.14 TgC yr$^{-1}$ (+37%), largely due to somewhat higher precipitation rates in that forcing dataset (see Table S3). Thus, the choice of input dataset itself introduces a significant degree of uncertainty to model output.

In addition, this model does not include explicit peatland formation and related dynamics, which is the subject of further model developments (Qiu et al., 2018) yet to be included in this iteration. With peatlands thought to cover ~17% of the Arctic land surface (Tarnocai et al., 2009), and with substantially higher leaching concentrations, this may be a significant omission from our model. The remaining biases likely arise from errors in the interaction of simulated NPP, respiration and DOC production and decomposition, which will impact on the net in and out-flow of dissolved carbon to the fluvial system. However, the DOC relationship with these variables is less clear-cut than with river discharge. Indeed, regressions (Fig. S2) of annual DOC versus NPP (TgC yr$^{-1}$) show that DOC is highly sensitive to increases in NPP, but is less coupled to it (more scattered, $R^2$=0.42) than other simulated fluvial carbon variables shown, i.e aquatic CO$_2$ evasion and soil CO$_2$ export to the river network. The differences in correlation and slope of the variables in Fig. S2 are expected: aquatic CO$_2$ evasion is least sensitive yet most tightly coupled to NPP ($R^2$=0.52), while CO$_2$ export to rivers is intermediate between the two ($R^2$=0.43). The greater correlation with NPP of DOC compared to evasion is understandable, given that DOC leaching is a covariate of both NPP and runoff, whereas evasion flux is largely dependent on organic inputs (production) and temperature (see Part 1).

**Text S4: Model Evaluation: DOC Discharge Seasonality**

Figure 4b shows that the bulk of the DOC outflow occurs during the spring freshet, accounting for ~50-70% of the total Arctic outflow (Lammers et al., 2001; Ye et al., 2009), in which DOC concentrations increase as meltwater flushes out DOC accumulated from the previous year’s litter and SOC generation (Raymond et al., 2007; Kutscher et al., 2017), reproduced in Fig. 4b. Simulation of the hydrological dynamic is presented in maps of river discharge through the basin in Fig. 3b, which show low-flows in April with substantial hydrographic flow from upstream mountainous headwaters and Lake Baikal inflow in the south, peak flow in June dominated by headwaters, and little headwater input in September.

In Fig. 4b we observe the following: (i) DOC discharge fluxes closely track hydrological fluxes. (ii) The simulated modern river discharge peak approximates the historical observed discharge peak, but slightly overestimates spring fluxes and substantially underestimates fluxes in the autumn, as explained above. (iii) The difference between the first and last decades of the simulation in Fig. 4b is mostly attributable to a large increase in the DOC flux mobilised by spring freshet waters. This suggests both greater peaks in simulated DOC flux and a shift to earlier peak timing, owing to an increase in river discharge indicative of an earlier spring and a progressively warmer environment over the 20th Century. (iv) The maximum modelled modern monthly DOC flux rate of ~1.3 TgC month$^{-1}$ is comparable to the mean maximum DOC flux rate measured in a
recent study (1.75 Tg C month⁻¹; Kutscher et al., 2017, Fig. 2).

We compare the Raymond et al. (2007) modern DOC outflow (Fig. 4d, solid black line) against simulated DOC outflow from both Zhigansk and Kusur (Fig. 4d). Simulated DOC flux is underestimated for both sites. Peakflow at Zhigansk seems to be attenuated over May and June in simulations, as opposed to May peakflow in observations. Peakflow at Kusur is definitively in June. This suggests that simulated outflow timing at Zhigansk may slightly delayed, causing a split in peak discharge when averaged in the model output. Thus the aggregation of model output to monthly averages from calculated daily and 30 minute timesteps can result in the artificial imposition of a normative temporal boundary (i.e. month) on a continuous series. This may cause the less distinctive ‘sharp’ peak seen in Fig. 4c, which is instead simulated at the downstream Kusur site, whose distance some 500km away from Zhigansk more clearly explains the delay difference in seasonality. We further evaluate our DOC discharge at the sub-basin scale, to test whether the fractional contribution of different DOC flows from each sub-basin correspond to those in their observed correlates from Kutscher et al., (2017). This comparison is depicted in Fig. 5S, where the observed and simulated percentage DOC contributions of the Aldan, Vilui, and Upper and Lower Lena sub-basins to total flux rates are 19 (24%), 20 (10%), 33 (38%) and 30 (20%) in simulations (observations) for the four sub-basins, respectively. While deviations between simulated and observed DOC fluxes can be expected, the nearly twofold value mismatch of the Vilui basin is due to its real-word damming, not represented here. On the other hand, we cannot explain the ~5% discrepancies in other sub-basin fluxes, particularly for the Aldan.

**Text S5**: Evaluation of Modelled DOC Concentrations in Stream, River, Ground Water

The spatial distribution of DOC concentrations are shown in maps of mean monthly DOC concentration for stream water, river water and groundwater (Fig. S6a,b,c, respectively) in April, June and September. For both the stream and river water reservoirs, DOC concentrations appear to have spatio-temporal gradients correlated with the flux of water over the basin during the thaw period, with high concentrations of 10–15 mg C L⁻¹ driven by April meltwaters upstream of the basin, these high concentrations moving northward to the coldest downstream regions of the basin in June. Lower DOC concentrations of ~5 mg C L⁻¹ dominate the basin in September when the bulk of simulated lateral flux of DOC has dissipated into the Laptev Sea. In contrast, groundwater DOC concentrations are generally stable with time, although some pixels appear to experience some ‘recharge’ in their concentrations during the first two of the three displayed thaw months. Significantly, highest groundwater DOC concentrations of up to 20 mg C L⁻¹ are focussed on the highest elevation areas of the Lena basin on its Eastern boundary, which are characterized by a dominance of Podzols (SI, Fig. S9b). This region, the Verkhoyansk range, is clearly visible as the high groundwater DOC concentration (2-20mg C L⁻¹) are (in red) in Fig. S6a, as well as other high elevation areas in the south-western portion of the basin (Fig. 3a), while the low-lying central basin shows much smaller groundwater DOC concentrations (0-2mg C L⁻¹). The range of simulated groundwater DOC concentration comes close to those aggregated from the empirical literature by Shvarts(2000), which finds from >9,000 observations that groundwater in permafrost regions exhibit a mean concentration of ~10 mg C L⁻¹ after peatlands and swamps (not simulated here) are removed (Table 2).
The high groundwater reservoir DOC concentrations simulated in high altitude regions by ORCHIDEE MICT-L is related to the fact that, in the model, DOC is rapidly produced and infiltrated deep into soil above the permafrost table, to the point that it reaches the simulated groundwater pool relatively quickly, allowing it to enter this reservoir before being metabolised through the soil column—hence allowing for the relatively high groundwater concentrations found in mountain areas. Because of the prevailing low temperatures, this DOC is not quickly decomposed by microbes and instead feed the groundwater DOC pool.

**Text S6: Riverine CO₂ Evasion**

In our model, the fate of DOC once it enters the fluvial system is either to remain as DOC and be exported to the ocean, or to be degraded to dissolved CO₂ (CO₂\(_{\text{aq.}}\)) and itself either also transported to the marine system or outgassed from the fluvial surface to the atmosphere (see Part 1, Section 2.10). The latter two outcomes also apply to CO₂\(_{\text{aq.}}\) produced in the soil by organic matter degradation and subsequently transported by runoff and drainage flows to the water column. As shown in Fig. 2, a large proportion of DOC (38%, 2.1 TgC yr\(^{-1}\)) that enters the water column is degraded to CO₂\(_{\text{aq.}}\) during transport, which adds to the 1.65 TgC yr\(^{-1}\) of direct CO₂\(_{\text{aq.}}\) input from the terrestrial land surface. Of this bulk CO₂ exported into and generated within the water column, 3.6 TgC yr\(^{-1}\) evades from the water surface to the atmosphere before reaching the river delta. In what follows, we evaluate first inputs of CO₂\(_{\text{aq.}}\) to the water column in terms of their seasonality, before evaluating CO₂ evasion rates and the relation of this to smaller and larger water bodies (river versus stream).

To our knowledge, no direct measurements for CO₂ evasion from the surface of the Lena river are available in the literature. We refer to Denfeld et al. (2013) for evaluating our evasion flux results, since their basin of study, the Kolyma River, is the most geographically proximate existing dataset to the Lena, despite biogeo graphical differences between the two basins—namely that the Kolyma is almost entirely underlain by continuous permafrost. The Kolyma River CO₂ evasion study measured evasion at 29 different sites along the river basin (~158-163°E; 68-69.5°N), with these sites distinguished from one another as ‘main stem’, ‘inflowing river’ or ‘stream’ on the basis of reach length. The study showed that during the summer low-flow period (August), areal river mainstem CO₂ evasion fluxes were ~0.35 gC m\(^{-2}\) d\(^{-1}\), whereas for streams of stream order 1-3 (widths 1-19m), evasion fluxes were up to ~7 gC m\(^{-2}\) d\(^{-1}\), and for non-mainstem rivers (widths 20-400m) mean net fluxes were roughly zero (Table 3 of Denfeld et al., 2013). Thus, while small streams have been observed to contribute to roughly 2% of the Kolyma basin surface area, their measured percentage contribution to total basin-wide CO₂ evasion ~40%, whereas for the main stem the surface area and evasion fractions were ~80% and 60%, respectively. Likewise, mean annual evasion rates of <0.8 up to around 7 gC m\(^{-2}\) d\(^{-1}\) have been found for the Ob and Pur rivers in Western Siberia (Serikova et al., 2018).

Results such as these, in addition to permafrost soil incubation experiments (e.g. Drake et al., 2015; Vonk et al., 2013, 2015b, 2015a) suggest that small streams, which represent the initial (headwater) drainage sites of these basins, rapidly process hydrologically leached carbon to the atmosphere, and that this high-reactivity carbon is
a mix of recently thawed ancient permafrost material, as well as decomposing matter from the previous growth year. This is given as evidence that the total carbon processing of high-latitude rivers is significantly underestimated if only mainstream carbon concentrations are used in the accounting framework, since a large amount of carbon is metabolised to the atmosphere before reaching the site of measurement.

**Text S7: Spatio-Temporal Heterogeneity in CO2 Evasion**

The heterogeneity of CO2 evasion from different sources in the model is most evident in their geographic distribution and relative intensity, as shown in the evasion flux rate maps over stream and river areas in April, June and September (Fig. S8a-b). Stream evasion (Fig. S8a), tends to be broadly distributed over the whole basin, representing the fact that small streams and their evasion are the main hydrologic connectors outside of the main river and tributary grid cells, whereas river evasion (Fig. S8b) is clearly linked to the hydrographic representation of the Lena main stem itself, with higher total quantities in some individual grid cells than for the stream reservoir, yet distributed amongst a substantially smaller number of grid cells. Whereas the stream reservoir has greatest absolute evasion flux rates earlier in the year (April-May), maximum evasion rates occur later in the year and further downstream for the river reservoir, reflecting the fact that headwaters are first-order integrators of soil-water carbon connectivity, whereas the river mainstem and tributaries are of a secondary order.

The spatio-temporal pattern of increasing evasion over the simulation period is shown in Fig. S7 as a Hovmöller difference plot, between the last and first decade, of log-scale average monthly evasion rates per latitudinal band. This shows that the vast majority of outgassing increase occurs between March and June, corresponding to the progressive onset of the thaw period moving northwards over this timespan. Although relatively small, outgassing increases are apparent for most of the year, particularly at lower latitudes. This would suggest that the change is driven most acutely by relatively greater temperature increases at higher latitudes ('Arctic amplification' of climate warming, e.g. Bekryaev et al., 2010) while less acute but more temporally homogenous evasion is driven by seasonal warming at lower latitudes.

**Text S8:**

This is because in ORCHIDEE MICT-L, the 'stream' water reservoir is water routed to the river network for all hydrologic flows calculated to not cross a 0.5 degree grid cell boundary (the resolution of the routing module, explained in Part 1, Section 2.6), which may not be commensurate with long, <20m width streams in the real-world, that were used in the Denfeld et al. (2013) study. In addition, this 'stream' water reservoir in the model does not include any values for width or area in the model, so we cannot directly compare our stream reservoir to the <20m width criterion employed by Denfeld et al. (2013) in their definition of an observed stream. Thus our 'stream' water reservoir encompasses substantially greater surface area and hydrologic throughput than that in the Denfeld et al. study.

**Text S9:**
The 'approximate' caveat refers to the fact that model output doesn’t define a precise surface area for the stream water reservoir, which is instead bundled into a single value representing the riverine fraction of a grid cell’s total surface area. To approximate the areal outgassing for the stream versus river water reservoirs, we weight the total non-floodplain inundated area of each grid cell by the relative total water mass of each of the two hydrological pools, then divide the total daily CO₂ flux simulated by the model by this value. The per-pool areal estimate is an approximation since it assumes that rivers and streams have the same surface area: volume relationship. This is clearly not the case, since streams are generally shallow, tending to have greater surface area per increment increase in depth than rivers. Thus, our areal approximations are likely underestimated (overestimated) for streams (rivers), respectively. Note that from ~700 non-zero simulation datapoints used to generate Fig. 6d were omitted as ‘outliers’ from the stream reservoir efflux statistics described below, because very low stream:river reservoir values skewed the estimation of total approximate stream surface area values very low, leading to extreme efflux rate values of 1-3000 g C m⁻² d⁻¹ and are thus considered numerical artefacts of the areal approximation approach used here.

Text S10: Emergent Phenomena: DOC and Topographic Slope

This relationship was found in temperate rivers by Lauerwald et al. (2012), and in a recent Pan-Arctic synthesis paper Connolly et al. (2018). The reasoning for the negative slope-DOC concentration relationship is that as elevation increases, temperature and primary production decreases. This leads to a thinner organic soil layer, meaning that mineral soil plays a stronger role in shallow hydrologic flowpaths, allowing for deeper infiltration and shorter residence time in a given soil layer. Further, steeper terrain leads to a lower soil water residence time and lower moisture than in flat areas. As a result, a given patch of soil matter will be exposed to leaching for less (residence) time, while the organic matter that is leached is thought to be adsorbed more readily to mineral soil particles, leading to either re-stabilisation in the soil column or shallow retention and subsequent heterotrophic respiration in situ, cumulatively resulting in lower DOC concentrations in the hydrologic export (Kaiser and Kalbitz, 2012; Klaminder et al., 2011). This line of reasoning was recently shown to apply also to deep organic permafrost soils (Zhang et al., 2017), although the degree to which this is the case in comparison to mineral soils is as yet unknown.

In addition, and as described in Part 1 (Section 2.5) of this study, MICT-L contains a provision for increased soil column infiltration and lower decomposition rates in areas underlain by Podzols and Arenosols. The map from the Harmonized World Soil Database (Nachtergaele, Freddy, Harrý van Vethuiizen, Luc Verelst, N. H. Batjes, Koos Dijkshoorn, V. W. P. van Engelen, Guenther Fischer, Arwyn Jones, Luca Montanarella, Monica Petri, Sylvia Prieler-B, Xuezheng Shi, Edmar Teixera and David Wiberg, 2010), which is used as the input to this criterion, shows areas underlain by these soils in the Lena basin to also be co-incident with areas of high topographic slope (Fig. 3a, SI, Fig S9b). The 'Podzol effect' is to increase the rate of decomposition and infiltration of DOC, relative to all other soil types, thus also increasing the rate of DOC flux into groundwater (see Part 1 of this study, Section 2.5). Thus, our modelling framework explicitly resolves the processes involved in these documented dynamics – soil thermodynamics, solid vertical flow (turbation), infiltration as a function of soil textures and types, adsorption as a
function of soil parameters (see Part 1 of this study, Section 2.11), DOC respiration as a function of soil temperature and hence depth (Part 1, Section 2.12), and lagging of DOC vertical flow behind hydrological drainage flow (summary Figure in Part 1, Fig. 1). We thus have some confidence in reporting that the simulated negative relationship of DOC concentration with topographic slope may indeed emerge from the model.

Text S11: Emergent Phenomena: DOC and Mean Annual Air Temperature

A key emergent property of DOC concentrations in soils and inland waters should be their positive partial determination by the temperature of the environment under which their rates of production occur, as has been shown in the literature on permafrost regions, most notably in Frey & Smith (2005) and Frey & McClelland (2009). Increasing temperatures should lead to greater primary production, thaw, decomposition and microbial mobilisation rates, and hence DOC production rates, leading to (dilution effects notwithstanding) higher concentrations of DOC in thaw and so stream waters. Looking at this emergent property allows us to evaluate the soil-level production of both DOC and thaw water at the appropriate biogeographic and temporal scale in our model. This provides a further constraint on model effectiveness at simulating existing phenomena at greater process-resolution.

Figure 7 compares three datasets (simulated and two observational) of riverine DOC concentration (in mg C L$^{-1}$) plotted against mean annual air temperature (MAAT). The simulated grid-scale DOC versus MAAT averaged over July and August (for comparability of DOC with observational sampling period) of 1998-2007 is shown in red, and observed data compiled by Laudon et al. (2012) and Frey and Smith (2005) for sites in temperate/cold regions globally and peatland-dominated Western Siberia, respectively. The Laudon et al. (2012) data are taken from 49 observations including MAAT over the period 1997-2011 from catchments north of 43°N, and aggregated to 10 regional biogeographies, along with datapoints from their own sampling; those in the Frey and Smith study are from 55-68°N and ~65-85°E (for site locations, see Laudon et al. (2012), Table 1 and 2; Frey and Smith (2005), Fig 1).

Fig. 7 can be interpreted in a number of ways. First, this MAAT continuum spans the range of areas that are both highly and moderately permafrost affected and permafrost free (Fig. 7, blue and green versus orange shading, respectively), potentially allowing us a glimpse of the behaviour of DOC concentration as the environment transitions from the former to the latter. Simulated Lena DOC concentrations, all in pixels with MAAT < -2°C and hence all bearing continuous or discontinuous permafrost (‘permafrost-affected’ in the figure), only exhibit a weakly positive response to MAAT on the scale used ($y=6.05e^{0.03MAAT}$), although the consistent increase in DOC minima with MAAT is clearly visible. Second, the Laudon et al. (2012) data exhibit an increasing then decreasing trend over the range of MAAT (-2°C to 10°C) in their dataset, which they propose reflects an ‘optimal’ MAAT range (0-3°C) for the production and transport of DOC (Fig. 7, red shading). Below this optimum range, DOC concentrations may be limited by transport due to freezing and above this, smaller soil carbon pools and temperature-driven decomposition would suppress the amount of DOC within rivers. Third, the lower end of the Laudon et al. (2012) MAAT values correspond to a DOC concentration in line with DOC concentrations simulated by our model. Fourth, DOC concentrations in the Frey and Smith (2005) data exhibit a broad scattering in
permafrost-affected sites, with concentrations overlapping those of our simulations (Fig. 7, green shading), before rapidly increasing to very high concentrations relative to the Laudon et al. (2012) data, as sites transition to permafrost-free (red shading, \( y=3.6\text{MAAT}+29.4 \)).

Their data highlight the difference in DOC concentration regime between areas of high (Frey and Smith, 2005) and low (Laudon et al., 2012) peatland coverage and the different response of these to temperature changes. Fifth, because our simulation results largely correspond with the observed data where the MAAT ranges overlap (green shading), and because our model lacks peatland processes, we should expect our model to follow the polynomial regression plotted for the Laudon et al. (2012) data as temperature inputs to the model increase. Figure 7 implies that this increase should be on the order of a doubling of DOC concentration as a system evolves from a MAAT of -2°C to 2°C. With warming, we expect the response of DOC concentrations to reflect a mix of both observationally-derived curves, as a function of peatland coverage.

**Text S12: LOAC drivers**

The constant climate (CLIM) and constant CO\(_2\) (CO\(_2\)) simulations described in Section 3 were undertaken to assess the extent –and the extent of the difference –to which these two factors are drivers of model processes and fluxes. These differences are summarised in Figs. S10 (a-b), in which we show the same 1998-2007–averaged yearly variable fluxes as in the CTRL simulation, expressed as percentages of the CTRL values given in Fig. 2. A number of conclusions can be drawn from these diagrams.

First, all fluxes are lower in the factorial simulations, which can be expected due to lower carbon input to vegetation from the atmosphere (constant CO\(_2\)) and colder temperatures (constant climate) inhibiting more vigorous growth and carbon cycling. Second, broadly speaking, both climate and CO\(_2\) appear to have similar effects on all fluxes, at least within the range of climatic and CO\(_2\) values to which they have subjected the model in these historical runs. With regard to lateral export fluxes in isolation, variable climate (temperature increase) is a more powerful driver than CO\(_2\) increase (see below). Third, the greatest difference between the constant climate and CO\(_2\) simulation carbon fluxes appear to be those associated with terrestrial inflow of dissolved matter to the aquatic network, these being more sensitive to climatic than CO\(_2\) variability. This is evidenced by a 49% and 32% decline in CO\(_2\) and DOC export, respectively, from the land to rivers in the constant climate simulation, versus a 27% and 23% decline in these same variables in the constant CO\(_2\) simulation. Given that the decline in primary production and respiration in both factorial simulations was roughly the same, this difference in terrestrial dissolved input is attributable to the effect of climate (increased temperatures) on the hydrological cycle, driving changes in lateral export fluxes.

This would imply that at these carbon dioxide and climatic ranges, the modelled DOC inputs are slightly more sensitive to changes in the climate rather than to changes in atmospheric carbon dioxide concentration and the first order biospheric response to this. However, while the model biospheric response to carbon dioxide concentration may be linear, thresholds in environmental variables such as MAAT may prove to be tipping points in the system’s emergent response to change, as implied by Fig. 7.
meaning that the Lena, as with the Arctic in general, may soon become much more temperature-dominated with regard to the drivers of its own change.

### Supplementary Tables and Figures

**Table S1**: Data type, name and sources of data files used to drive the model in the study simulations.

<table>
<thead>
<tr>
<th>Data Type</th>
<th>Name</th>
<th>Source</th>
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<tr>
<td>Vegetation Map</td>
<td>ESA CCI Land Cover Map</td>
<td>(Bontemps et al., 2013)</td>
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<tr>
<td>Topographic Index</td>
<td>STN-30p</td>
<td>(Vörösmarty et al., 2000)</td>
</tr>
<tr>
<td>Stream flow direction</td>
<td>STN-30p</td>
<td>Vörösmarty et al., 2000</td>
</tr>
<tr>
<td>River surface area</td>
<td>(Lauerwald et al., 2015)</td>
<td></td>
</tr>
<tr>
<td>Soil texture class</td>
<td>(Reynolds et al., 1999)</td>
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<tr>
<td>Climatology</td>
<td>GSWP3 v0, 1 degree</td>
<td><a href="http://hydro.iis.u-tokyo.ac.jp/GSWP3/">http://hydro.iis.u-tokyo.ac.jp/GSWP3/</a></td>
</tr>
<tr>
<td>Potential floodplains</td>
<td>Multi-source global wetland maps</td>
<td>(Tootchi et al., 2019)</td>
</tr>
<tr>
<td>Poor soils</td>
<td>Harmonized World Soil Database map</td>
<td>(Nachtergaele, Freddy, Harrij van Velthuizen, Luc Verelst, N. H. Batjes, Koos Dijkshoorn, V. W. P. van Engelen, Guenther Fischer, Arwyn Jones, Luca Montanarella, Monica Petri, Sylvia Prieler-B, Xuezhoug Shi, Edmar Teixera and David Wilberg, 2010)</td>
</tr>
<tr>
<td>Spinup Soil Carbon Stock</td>
<td>20ky ORCHIDEE-MICT soil carbon spinup</td>
<td>(Guimberteau et al., 2018)</td>
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**Table S2**: Literature sources for empirical evaluation of model output.

<table>
<thead>
<tr>
<th>Empirical Evaluation Sources</th>
<th>DOC Discharge</th>
<th>Water Discharge</th>
<th>DOC concentration</th>
<th>NPP</th>
<th>Soil Respiration</th>
<th>CO2 Evasion</th>
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<tr>
<td>Cauwet and Sidorov (1996); Dolman et al. (2012); Holmes et al. (2012); Lara et al. (1998); Raymond et al. (2007); Semiletov et al. (2011); Kutscher et al. (2017).</td>
<td>Ye et al. (2009); Lammers et al. (2001)</td>
<td>Shvartsev (2008); Denfeld et al. (2013); Mann et al. (2015); Raymond et al. (2007); Semiletov et al. (2011); Arctic-GRO/PARTNERS (Holmes et al., 2012)</td>
<td>Beer et al. (2000); Lloyd et al. (2002); Roser et al. (2002); Schulze et al. (1999); Shvidenko and Nilsson, (2003)</td>
<td>Eberling (2007); Sawamoto et al. (2000); Sommerkorn (2008).</td>
<td>Denfeld et al. (2013); Serikova et al. (2018).</td>
<td></td>
</tr>
</tbody>
</table>

**Table S3**: Observed versus simulated DOC discharge (1998-2007), where we compare the output of two separate climatological datasets used as input to the model (GSWP3 and ISIMIP 2b). Also shown are the simulated versus observed DOC discharge for the six largest Arctic rivers (the "Big Six") and for the Pan-Arctic as a whole.
<table>
<thead>
<tr>
<th>Simulated DOC to Ocean</th>
<th>Simulated DOC to Ocean</th>
<th>Observations [Holmes et al., 2012]</th>
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<tr>
<td>GSWP3</td>
<td>SIMIP 2b</td>
<td>PARTNERS/Arctic-GRO</td>
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<td>Lena</td>
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<td>4.14</td>
</tr>
<tr>
<td>Big 6</td>
<td>19.36</td>
<td>18.11</td>
</tr>
<tr>
<td>Pan-Arctic</td>
<td>32.06</td>
<td>34.04</td>
</tr>
</tbody>
</table>

Table S4: Mean observed groundwater CO$_2$ and DOC concentrations for global permafrost regions subdivided by biogeographic province and compiled by Shvartsev (2008) from over 9000 observations.

<table>
<thead>
<tr>
<th>Permafrost Groundwater Provinces</th>
<th>Swamp</th>
<th>Tundra</th>
<th>Taiga</th>
<th>Average</th>
<th>Average [-Swamp]</th>
</tr>
</thead>
<tbody>
<tr>
<td>CO$_2$ (mg C L$^{-1}$)</td>
<td>12.3</td>
<td>14</td>
<td>10.8</td>
<td>12.4</td>
<td>12.4</td>
</tr>
<tr>
<td>DOC (mg C L$^{-1}$)</td>
<td>17.6</td>
<td>10.1</td>
<td>9.3</td>
<td>12.3</td>
<td>9.7</td>
</tr>
</tbody>
</table>
Figure S1: (a-b) Carbon and water flux map for core DOC elements in model structure relating to DOC transport and transformation, first published in Part 1 of this study. (a) Summary of the differing extent of vertical discretisation of soil and snow for different processes calculated in the model. Discretisation occurs along 32 layers whose thickness increases geometrically from 0-38m. N refers to the number of layers, SWE=snow water equivalent, $S_n=$ Snow layer n. Orange layers indicate the depth to which diffusive carbon (turbation) fluxes occur. (b) Conceptual map of the production, transfer and transformation of carbon in its vertical and lateral (i.e., hydrological) flux as calculated in the model. Red boxes indicate meta-reservoirs of carbon, black boxes the actual pools as they exist in the model. Black arrows indicate carbon fluxes between pools, dashed red arrows give carbon loss as CO$_2$, green arrows highlight the fractional distribution of DOC to SOC (no carbon loss incurred in this transfer), a feature of this model. For a given temperature (5°C) and soil clay fraction, the fractional fluxes between pools are given for each flux, while residence times for each pool (τ) are in each box. The association of carbon dynamics with the hydrological module are shown by the blue arrows. Blue coloured boxes illustrate the statistical sequence which activates the boolean floodplains module. Note that for readability, the generation and lateral flux of dissolved CO$_2$ is omitted from this diagram, but is described at length in the Methods section. (c) (Left) Soil carbon concentrations per depth level for each soil carbon reactivity pool at the end of the spinup period. (Right) Evolution of each soil carbon pool over the course of the 400-year spinup quasi-equilibration period.
Figure S2: Summed yearly lateral flux versus NPP values for DOC discharge, CO$_2$ discharge and CO$_2$ evasion (FCO$_2$) over the entire simulation period, with linear regression lines shown.
Figure S3: Observed versus simulated mean annual water discharge from the Lena river, where observations are taken from (Ye et al., 2003).

(a)

(b)

(c)
Figure S4: (a) Absolute yearly gross primary productivity (GPP, TgC yr⁻¹) for the four relevant PFT groups over the Lena basin, averaged over 1998-2007. (b) Mean July and August soil heterotrophic respiration rates (g m⁻² d⁻¹) for the same PFT groups as in (a), during the period 1998-2007. (c) Average yearly NPP (gC m⁻² yr⁻¹) averaged over the period 1998-2007. All maps have the Lena basin area shaded in the background. (d)
Mean monthly carbon uptake (GPP) versus its heterotrophic respiration from the soil (Het Resp) in TgC per month, over the period 1998-2007.

Figure S5: Map adapted from Fig. 2 in Kutscher et al. (2017) showing proportional sub-basin contributions of TOC outflow to total TOC discharge in June and July (designated as their sampling period 'p-1') of 2012-2013, as observed in Kutscher et al., 2017 (black arrows), and DOC export contributions as simulated over the period 1998-2007 by ORCHIDEE MICT-L (red boxes). Simulation pixels used in the calculation are correlates of the real-world sampling locations unless the site coordinates deviated from a mainstem hydrographic flowpath pixel—in which case a nearest 'next-best' pixel was used. Here the percentages are out of the summed mean bulk DOC flow of each tributary, not the mean DOC discharge from the river mouth, because doing so would negate the in-stream loss of DOC via degradation to CO2 while in-stream.
Figure S6: Maps of (a) DOC concentrations (mg C L$^{-1}$) in groundwater (‘slow’ water pool), (b) stream water pool, (c) river water pool in April, June and September (first to third rows, respectively) averaged over the period 1998-2007. The coastal boundary and a water body overlay have been applied to the graphic in red and black, respectively, and the same scale applies to all diagrams. All maps have the Lena basin area shaded in the background.
Figure S7: Log-scale Hovmöller diagram plotting the longitudinally-averaged difference (increase) in total CO₂ evaded from the Lena River basin between the average of the periods 1998-2007 and 1901-1910, over each monthly timestep, in (log) gC m⁻² d⁻¹. Thus as the river drains northward the month-on-month difference in water-body CO₂ flux, between the beginning and end of the 20th Century is shown.

Figure S8: Maps of CO₂ evasion from the surface of the two fluvial hydrological pools in the model, (a) streams and (b) rivers in April, June and September. All maps use the same (log) scale in units of (mgC m⁻² d⁻¹).
Figure S9: (a) Maximum floodable fraction of grid cells for the Lena basin per the input map from Tootchi et al. (2018). (b) Podzol and Arenosol map (Nachtergaele, Freddy, Harrij van Velthuizen, Luc Verelst, N. H. Batjes, Koos Dijkshoorn, V. W. P. van Engelen, Guenther Fischer, Arwyn Jones, Luca Montanarella, Monica Petri, Sylvia PrielerB, Xuezheng Shi, Edmar Teixera and David Wiberg, 2010) used as input to the ‘poor soils’ module, basin mask in the background.
Constant Climate (% of CTRL)

- Plant Respiration: $\delta\text{CO}_2 = 67\%$, $\delta\text{DOC} = 61\%$, $\delta\text{CO}_2 = 70\%$
- Soil Respiration: $\delta\text{CO}_2 = 72\%$
- CO2 Evasion: $\delta\text{CO}_2 = 81\%$
- DOC Respiration: $\delta\text{DOC} = 67\%$
- DOC Return: $\delta\text{DOC} = 61\%$, $\delta\text{CO}_2 = 70\%$
- DOC Evasion: $\delta\text{CO}_2 = 73\%$

Constant CO2 (% of CTRL)

- Plant Respiration: $\delta\text{CO}_2 = 67\%$, $\delta\text{DOC} = 63\%$, $\delta\text{CO}_2 = 70\%$
- Soil Respiration: $\delta\text{CO}_2 = 73\%$
- CO2 Evasion: $\delta\text{CO}_2 = 83\%$
- DOC Respiration: $\delta\text{DOC} = 77\%$
- DOC Return: $\delta\text{DOC} = 77\%$, $\delta\text{CO}_2 = 76\%$
- DOC Evasion: $\delta\text{CO}_2 = 76\%$

Terrestrial Aquatic Marine

DOC CO2 Evasion Throughfall Flux Plant Resp. DOC Resp. DOC Return DOC Evasion Runoff

Flooded Soil

$\delta\text{DOC} = 74\%$, $\delta\text{CO}_2 = 81\%$

$\delta\text{DOC} = 66\%$, $\delta\text{CO}_2 = 71\%$

$\delta\text{DOC} = 68\%$, $\delta\text{CO}_2 = 67\%$

$\delta\text{DOC} = 67\%$, $\delta\text{CO}_2 = 70\%$

$\delta\text{DOC} = 67\%$, $\delta\text{CO}_2 = 67\%$

$\delta\text{DOC} = 67\%$, $\delta\text{CO}_2 = 75\%$

$\delta\text{DOC} = 67\%$, $\delta\text{CO}_2 = 70\%$

$\delta\text{DOC} = 74\%$, $\delta\text{CO}_2 = 76\%$
Figure S10: (a-b): Schematic diagrams detailing the major yearly carbon flux outputs from simulations averaged over the period 1998-2007 as they are transformed and transported across the land-aquatic continuum. Figures (b) and (c) give the same fluxes as a percentage difference from the Control (CTRL-Simulation), for the constant climate and CO₂ simulations, respectively.

Figure S11: Simulated basin-mean annual DOC concentrations (mg L⁻¹) for the floodplain water pool regressed against mean annual simulated discharge rates at Kusur (m³ s⁻¹) over 1901-2007. A linear regression with R² is plotted.
Figure S12: Time series showing the decadal-mean fractional change in carbon fluxes normalised to a 1901-1910 average baseline (=1 on the y-axis) for NPP, GPP, autotrophic and heterotrophic respiration, DOC inputs to the water column, CO₂ inputs to the water column, CO₂ evasion from the water surface (FCO₂), and discharge.

References


Kutscher, L., Möth, C. M., Porcelli, D., Hirst, C., Maximov, T. C., Petrov, R. E. and


Roser, C., Montagnani, L., Schulze, E.-D., Mollicone, D., Kolle, O., Meroni, M., Papale, D.,


Text S1: Groundwater DOC Concentrations

The high groundwater reservoir DOC concentrations simulated in high altitude regions by ORCHIDEE MICT-L is related to the fact that, in the model, DOC is rapidly produced and infiltrated deep into soil above the permafrost table, to the point that it reaches the simulated groundwater pool relatively quickly, allowing it to enter this reservoir before being metabolised through the soil column – hence allowing for the relatively high groundwater concentrations found in mountain areas. Because of the prevailing low temperatures, this DOC is not quickly decomposed by microbes and instead feed the groundwater DOC pool.

Text S2: Evaluation of Simulated NPP and Soil Respiration

Rates of yearly net primary production (NPP) for Russian and Siberian forests have been inferred in situ from eddy flux and inventory techniques to range from 123-250 gC m\(^{-2}\) yr\(^{-1}\) (Beer et al., 2006; Lloyd et al., 2002; Roser et al., 2002; Schulze et al., 1999; Shvidenko and Nilsson, 2003). We likewise simulate a broad range of NPP carbon uptake rates, of 61-469 gC m\(^{-2}\) yr\(^{-1}\) averaged per grid cell over the Lena basin, with a mean value of 210 gC m\(^{-2}\) yr\(^{-1}\). NPP is heterogeneously distributed over space and between PFTs (SI, Fig. S5c), with forests averaging 90 gC m\(^{-2}\) yr\(^{-1}\) and grasslands averaging 104 gC m\(^{-2}\) yr\(^{-1}\) over the basin as a whole. Low values tended to originate in basin grid cells with elevated topography or high mean slope, while the maximum value was standalone, exceeding the next greatest by \~100 gC m\(^{-2}\) yr\(^{-1}\), and is most likely caused by the edge effects of upscaling a coastal gridcell’s small fraction of terrestrial area where high productivity occurs in a small plot, to the grid cell as a whole. By evaluating NPP we are also evaluating at a secondary level litter production, which is at a third level a major component of DOC production.

Taken as a whole, gross primary production (GPP) was performed under simulations by four PFT groups, with the largest basin-wide bulk contributions coming from boreal needleleaf summer-green trees and C3 grasses (SI, Fig. S5a), the highest GPP uptake rates (3 TgC pixel\(^{-1}\) yr\(^{-1}\)) generated by boreal needleleaf evergreen trees, and the remainder of GPP contributed by Boreal broad-leaved summer-green trees (SI, Fig. S5a).

Soil respiration rates, of combined soil heterotroph and plant root respiration in our Control simulation, averaged 208 gC m\(^{-2}\) yr\(^{-1}\) (0.57 gC m\(^{-2}\) d\(^{-1}\)) over the Lena basin over the period 1990-2000, which is somewhat higher than those found by Elberling (2007) in tundra soils over Svalbard, of 103-176 gC m\(^{-2}\) yr\(^{-1}\) (0.28-0.48 gC m\(^{-2}\) d\(^{-1}\)). Sawamoto, et al. (2000) measured in situ summertime soil respiration over the central Lena basin and found rates of 1.6-34 gC m\(^{-2}\) d\(^{-1}\), while Sommerkorn (2008) observed rates of 0.1-3.9 gC m\(^{-2}\) d\(^{-1}\) at higher latitudes, these appearing to vary with vegetation and fire history, water table depth and temperature. Mean heterotrophic respiration rates of 1.6 gC m\(^{-2}\) d\(^{-1}\) are simulated here during July and August, in the range 0.05-2.2 gC m\(^{-2}\) d\(^{-1}\) for each of the above PFT groups. The spatial distribution of, and difference in respiration rates between PFT groups largely mirrors those for NPP (SI Fig. S5c), with maximum rates of 1.4 gC m\(^{-2}\) d\(^{-1}\) over forested sites,
versus a maximum of 2.2 gC m² d⁻¹ over grassland/tundra sites (SI, Fig. S5b).

Aggregated over the basin, results show that increases over the course of the 20th Century were simulated for NPP, GPP, River Discharge, DOC, CO₂(aq.), autotrophic and heterotrophic respiration and CO₂ evasion, with percentage changes in the last versus first decade of +25%, +27%, 38%, +73%, +60%, +30%, +33% and +63%, respectively. (Fig. S7). It thus appears that rising temperatures and CO₂ concentrations disproportionately favoured the metabolisation of carbon within the soil and its transport and mineralisation within the water column, fed by higher rates of primary production and litter formation as well as an accelerated hydrological cycle.

Figure S2: Map adapted from Fig. 2 in Kutscher et al. (2017) showing proportional sub-basin contributions of TOC outflow to total TOC discharge in June and July (designated as their sampling period 'p-1') of 2012-2013, as observed in Kutscher et al., 2017 (black arrows), and DOC export contributions as simulated over the period 1998-2007 by ORCHIDEE MICT-L (red boxes). Simulation pixels used in the calculation are correlates of the real-world sampling locations unless the site coordinates deviated from a mainstem hydrographic flowpath pixel – in which case a nearest ‘next-best’ pixel was used. Here the percentages are out of the summed mean bulk DOC flow of each tributary, not the mean DOC discharge from the river mouth, because doing so would negate the in-stream loss of DOC via degradation to CO₂ while in-stream.
Figure S4: Groundwater DOC concentrations over the Lena basin for April, June and September averaged over 1998-2007, with mean observed concentrations for permafrost groundwater inset.

(a)

(b)
Figure SS: (a) Absolute yearly gross primary productivity (GPP, TgC yr\(^{-1}\)) for the four relevant PFT groups over the Lena basin, averaged over 1998-2007. (b) Mean July and August soil heterotrophic respiration rates (g m\(^2\) d\(^{-1}\)) for the same PFT groups as in (a), during the period 1998-2007. (c) Average yearly NPP (gC m\(^2\) yr\(^{-1}\)) averaged over the period 1998-2007. All maps have the Lena basin area shaded in the background. (d) Mean monthly carbon uptake (GPP) versus its heterotrophic respiration from the soil (Het_Resp) in TgC per month, over the period 1998-2007.