Simulated single-layer forest canopies delay Northern Hemisphere snowmelt

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Abstract.

Single-layer vegetation schemes have been found to overestimate diurnal cycles in longwave radiation beneath forest canopies. This study derives a correction from forest stand-scale simulations, which reduces diurnal cycles of sub-canopy longwave radiation. Correction factors are subsequently implemented in land-only simulations of the Community Land Model version 4.5 (CLM4.5) in order to assess the impact on snow cover. Nighttime underestimations of sub-canopy longwave radiation outweigh daytime overestimations, which leads to underestimated averages over the snow cover season. As a result, snow temperatures are underestimated and snowmelt is delayed in CLM4.5 across evergreen boreal forests. Increasing insolation and day length change the impact of overestimated diurnal cycles on daily average sub-canopy longwave radiation throughout the snowmelt season. Consequently, delay in snowmelt is more substantial where winters are warm and snowmelt occurs early, which results in a shortened snowmelt duration across boreal forests.

1 Introduction

Forest canopy cover modulates longwave radiation received by the ground, which consequently differs from atmospheric longwave radiation. This process is called longwave enhancement and has been shown to result in substantial postive net longwave radiation of the surface when snow cover is prevalent, especially under clear skies and during snowmelt (Webster et al., 2016). In contrast, net longwave radiation fluxes are typically negative for snow under clear-sky conditions in unforested areas as has been observed for evergreen Canadian boreal forests (Harding and Pomeroy, 1996; Ellis et al., 2010). Moreover, forest cover has been reported to enhance snowmelt for a subarctic open woodland during overcast days and early in the snowmelt season (Woo and Giesbrecht, 2000). However, the impact of forest coverage on snowmelt varies regionally as a function of forest density and meteorological conditions, with the importance of shortwave and longwave radiation changing throughout the snowmelt season (Sicart et al., 2004; Lundquist et al., 2013).

Meteorological conditions control longwave enhancement as clear skies increase insolation and thereby vegetation temperature while radiative temperature of the sky is reduced (Sicart et al., 2004; Lundquist et al., 2013; Todt et al., 2018). Therefore, values for longwave enhancement, i.e. the ratio of below-canopy to above-canopy longwave radiation, are higher under clear skies but close to 1 or even smaller for overcast conditions due to similar vegetation temperature and radiative temperature of the

sky. Vegetation density impacts longwave enhancement by scaling the respective contributions of vegetation and atmosphere to sub-canopy longwave radiation as well as by governing the impact of meteorological forcing on vegetation temperatures (Todt et al., 2018). While observations have shown trunks heating up due to insolation and emission of longwave radiation consequently increasing (Rowlands et al., 2002; Pomeroy et al., 2009), diurnal variations in tree temperatures depend on exposure

5   to insolation and thus vegetation density (Webster et al., 2016).

About a fifth of seasonally snow-covered land over the Northern Hemisphere is covered by boreal forests (Rutter et al., 2009), indicating the process of longwave enhancement affects a substantial fraction of global snow cover. Considerable challenges persist in the representation of snow cover and snowmelt in the current generation of climate models as historical simulations by the suite of models from Climate Model Intercomparison Project's fifth phase (CMIP5) underestimate observed trend

10   and interannual variability of spring snow cover extent (SCE) (Derksen and Brown, 2012; Brutel-Vuilmet et al., 2013; Rupp et al., 2013; Mudryk et al., 2014; Thackeray et al., 2016). Snow Model Intercomparison Project's second phase (SnowMIP2) identified less skill in modelling snow for forested than for open sites, which was attributed to complex processes between atmosphere, snow, and vegetation (Essery et al., 2009; Rutter et al., 2009).

Among models displaying deficiencies in simulating snow cover evolution across boreal forests is the Community Land

15   Model (CLM) version 4 and its parent Community Climate System Model version 4 (Thackeray et al., 2014, 2015). CLM uses a one-layer vegetation scheme and CLM version 4.5 (CLM4.5) has been found to show deficiencies in simulation of sub-canopy longwave radiation and longwave enhancement with overestimated diurnal cycles under clear-sky conditions (Todt et al., 2018). Similar issues have been mitigated in CLM4.5 by subdividing the roughness layer (Bonan et al., 2018) and in SNOWPACK, a one-dimensional snow cover model, by partitioning the canopy into two layers with separate energy balances

20   and consequently separate vegetation temperatures (Gouttevin et al., 2015), which results in different longwave radiation fluxes emitted upward and downward from the vegetation.

In order to avoid implementing multiple canopy layers in a global land model and associated computational costs, this study presents an alternative guided by the effect of separate vegetation layers on sub-canopy longwave radiation. A correction to sub-canopy longwave radiation is implemented in CLM4.5 to reduce overestimated diurnal cycles, damping variations in

25   longwave radiation emitted downward and, consequently, increasing variations in longwave radiation emitted upward. While simulation of sub-canopy longwave radiation and longwave enhancement by land surface models had so far been assessed using forest stand-scale forcing and evaluation data, this study uses land-only simulations of CLM4.5 to assess the impact of overestimated diurnal cycles in sub-canopy longwave radiation on global snow cover and snowmelt. Therefore, this study has three objectives:

30   i. To develop a correction of sub-canopy longwave radiation simulated by single-layer vegetation in CLM4.5;

ii. To demonstrate the effect of correction on diurnal cycles of and daily average sub-canopy longwave radiation; and

iii. To quantify the impact of corrected sub-canopy longwave radiation on snow cover and snowmelt across the Northern Hemisphere.

Treatment of sub-canopy longwave radiation in CLM4.5, implementation of correction, and land-only simulations are described in Sect. 2. Calculation of correction factors is detailed in Sect. 3. Results are presented in Sect. 4 and discussed in Sect. 5.

2 Methods

2.1 Sub-canopy longwave radiation in CLM4.5

Vegetation in CLM4.5 is parameterized as a single layer using a "big-leaf" approach (Oleson et al., 2013). Sub-canopy longwave radiation is calculated as the sum of atmospheric longwave radiation LW_{atm} and longwave radiation emitted by vegetation LW_{veg}, weighted by vegetation emissivity ε_v:

$$LW_{sub} = (1-\varepsilon_v)\, LW_{atm} + \varepsilon_v\, \sigma\, T_v^4 \tag{1}$$

using the Stefan-Boltzmann law with Stefan-Boltzmann constant σ = 5.67 10^{-8} $\mathrm{Wm^{-2}K^{-4}}$ and vegetation temperature T_v. Vegetation emissivity depends on Leaf Area Index (LAI) and Stem Area Index (SAI) and is calculated as

$$\varepsilon_v = 1 - e^{-(LAI+SAI)}. \tag{2}$$

CLM4.5 subdivides grid cells based on land units, e.g. vegetated, glacier, or urban, and vegetated land units based on Plant Functional Types (PFTs), with up to 16 possible PFTs as well as bare soil. Sub-canopy longwave radiation is calculated for each PFT present in a grid cell, with separate values of LAI, SAI, and vegetation temperature for each PFT. All PFTs within one vegetated land unit share a single column of snow and soil, so that fluxes from vegetation to the ground are weighted averages over all PFTs. Consequently, changes in fluxes from an individual PFT affect snow cover beneath every PFT in a particular vegetated land unit.

2.2 Correction of sub-canopy longwave radiation in CLM4.5

For this study, a correction factor f_{corr} was implemented in CLM4.5 to reduce unphysical diurnal variations in sub-canopy longwave radiation. As atmospheric longwave radiation is an input variable to CLM4.5, from either forcing datasets or the atmospheric component of CESM (Community Atmosphere Model, CAM), correction factors were used to scale longwave radiation emitted from vegetation:

$$LW_{sub} = (1-\varepsilon_v)\, LW_{atm} + \varepsilon_v\, \sigma\, T_v^4\, f_{corr}. \tag{3}$$

Conceptually, correction factors represent a vegetation consisting of multiple individual layers, so that longwave radiation fluxes emitted upward and downward from the vegetation are no longer equal by design. In a multi-layer canopy scheme, the uppermost layer contributes most to longwave radiation emitted upward and is directly exposed to atmospheric forcing. Conversely, the lowest layer contributes most to longwave radiation emitted downward, but is shaded and sheltered by canopy layers above.

Using multi-layer canopy schemes as a guideline, longwave radiation was redistributed in CLM4.5 resulting in asymmetrical upward and downward longwave radiation fluxes and consequently, above-canopy longwave radiation was calculated as

$$LW_{above} = (1 - \varepsilon_v)(1 - \varepsilon_g)(1 - \varepsilon_v) LW_{atm} + \varepsilon_v \left((2 - f_{corr}) + (1 - \varepsilon_v)(1 - \varepsilon_g) f_{corr} \right) \sigma T_v^4 + (1 - \varepsilon_v) \varepsilon_g \sigma T_g^4 \quad (4)$$

with emissivity of the ground ε_g and temperature of the ground T_g. In Eq. (4), the first term represents atmospheric longwave

5 radiation transmitted through the vegetation, reflected by the ground, and transmitted through the vegetation to the atmosphere; the second term represents longwave radiation emitted from the vegetation reaching the atmosphere; and the third term represents longwave radiation emitted by the ground and transmitted through the vegetation to the atmosphere. The second term combines longwave radiation emitted by the vegetation directly to the atmosphere (first term in parentheses) and longwave radiation emitted downward from the vegetation, reflected by the ground, and transmitted through the vegetation to the atmo-

10 sphere (second term in parentheses). For $f_{corr} > 1$, LW_{above} decreases as reduction in the first term in parentheses $(2 - f_{corr})$ outweighs increase in the second term in parentheses $(1 - \varepsilon_v)(1 - \varepsilon_g) f_{corr}$, while LW_{sub} in Eq. (3) increases. For $f_{corr} < 1$, LW_{sub} decreases and LW_{above} increases. Note that the sum of LW_{sub} and LW_{above} was not changed by the introduction of f_{corr}, which guaranteed conservation of energy. The calculation of vegetation temperature in CLM4.5 was not altered by this approach, so that the temperature of the single vegetation layer represented an average of multiple (theoretical) layers

15 suggested by asymmetrical upward and downward longwave radiation fluxes.

2.3 Global offline simulations with CLM4.5

Offline simulations of CLM4.5 were forced by prescribed atmospheric data, using the CRUNCEP version 7 data set covering 1981 to 2016 and thus snow seasons 1981/82 to 2015/16 (Viovy, 2018). The impact of correction factors on longwave enhancement, snow cover, and snowmelt was assessed by comparing two simulations, a control run (henceforth, CTRL) and a run in

20 which correction factors were implemented (henceforth, CORR). Correction factors were applied to evergreen needleleaf trees in CLM4.5 as given in Eq. (3) and Eq. (4). Two PFTs, Needleleaf Evergreen Boreal Trees (NEBTs) and Needleleaf Evergreen Temperate Trees (NETTs), represent evergreen forests across snow-covered areas in CLM4.5 and grid-cell coverage by these two PFTs is shown in Fig. 1a. Plant Area Index (PAI), the sum of LAI and SAI, is shown in Fig. 1b as a weighted average of NEBTs and NETTs.

25 3 Calculation of correction factors

Todt et al. (2018) created a "toy model", which utilized forest stand-scale forcing data to evaluate sub-canopy longwave radiation in CLM4.5 and revealed systematic simulation errors that depend on meteorological conditions. These meteorological conditions were categorized via insolation and cloudiness represented by effective emissivity of the sky, which is calculated as

30 $$\varepsilon_{sky} = \frac{LW_{atm}}{\sigma T_{air}^4} \quad (5)$$

Table 1. Forest stands used for calculation of correction factors based on simulations by Todt et al. (2018). Vegetation density is given as Plant Area Index (PAI), the one-sided area of plant components per unit ground surface area including stems, branches, and leaves or needles. Abisko and Cherskiy feature deciduous vegetation, so that trees were leafless throughout the simulation periods and PAI values do not consider leaves or needles.

Location	Abisko, Sweden	Alptal, Switzerland	Cherskiy, Russia	Seehornwald, Switz.	Sodankylä, Finland
Latitude [°N]	68.4	47.1	68.7	46.8	67.4
Longitude [°E]	18.8	8.8	161.4	9.9	26.6
Snowmelt Season	2011	2004-07	2017	2008-12	2012
Days of Evaluation	9	41, 57, 73, 85	51	116, 90, 106, 83, 116	37
Tree Type	birch	fir & spruce	larch	fir & spruce	pine
Tree Height [m]	3.5	25	5	25	18
PAI [$m^2 m^{-2}$]	0.44	4.1	0.67	5.1	1.14

using air temperature T_{air}. Based on those stand-scale simulations, this study calculated correction factor f_{corr} from ε_{sky} and insolation SW_{in} as

$$f_{corr}^{-1} = b_0 + b_1\,\varepsilon_{sky} + b_2\,SW_{in} + b_3\,SW_{in}\,\varepsilon_{sky}. \tag{6}$$

Coefficients $b_{0,...,3}$ relate to intercept, ε_{sky}, insolation, and interaction of ε_{sky} and insolation, respectively, and were calculated
5   via multiple linear regression from stand-scale simulation errors expressed as ratios (Fig. 2) and observations of ε_{sky} and insolation at forest stands listed in Table 1. Consequently, correction factors were calculated as inverses of these ratios to scale longwave radiation in CLM4.5. For example, if stand-scale simulations revealed an overestimation of longwave radiation by 25% for particular values of ε_{sky} and SW_{in}, correction factors in global simulations would be $1.25^{-1} = 0.8$ for the same meteorological conditions. As CLM4.5 only simulates longwave radiation emitted from vegetation, simulation errors were
10 calculated for LW_{veg} that was derived from sub-canopy longwave radiation via Eq. (1) and Eq. (2) using measurements of atmospheric longwave radiation and PAI given in Table 1. Error ratios as a function of ε_{sky} and insolation as well as estimates based on regression coefficients are shown in Fig. 2. Nighttime estimates are a linear function of ε_{sky} as insolation is zero, while daytime estimates involve interaction of effective emissivity of the sky and insolation. Both daytime and nighttime simulation errors generally increase in magnitude with clearer skies. However, insolation is necessary in addition to ε_{sky} in
15 order to determine daytime regression slopes.

Regression coefficients as outlined in Eq. (6) are shown in Fig. 3 for every site and season, differentiated for day and night. Intercept b_0 and regression coefficient for ε_{sky} b_1 agree in sign for all sites and agree in magnitude for all sites except Abisko

(panels a, b, d, e), with little interannual variability for the two sites with multiple years of data, Alptal and Seehornwald. In contrast to Abisko, b_0 and b_1 for the deciduous forest at Cherskiy are similar to those for evergreen sites Alptal, Seehornwald, and Sodankylä despite featuring different vegetation type, structure, and density. Regression coefficients involving insolation agree in sign but differ in magnitude among Alptal, Cherskiy, Seehornwald, and Sodankylä (panels c and f), with similar values

5   for the latter two sites due to little interannual variability for Seehornwald. In contrast, interannual variability is large for Alptal with higher magnitudes for all four years combined compared to Seehornwald and Sodankylä, while magnitudes are smallest for Cherskiy. For Abisko, five out of six regression coefficients display smallest magnitudes due to deciduous vegetation and consequently low vegetation density as well as smaller simulation errors compared to other sites (Todt et al., 2018). Overall, uncertainties are largest for Abisko due to a short evaluation period, with no regression coefficient being significantly different

10   from one, as in the case of intercept b_0, or zero.

For implementation in global simulation CORR, regression coefficients were calculated based on one season each of Alptal, Cherskiy, Seehornwald, and Sodankylä in order to balance dense and sparser sites. Despite featuring a deciduous PFT, Cherskiy was included as regression coefficients are similar to evergreen sites. Individual seasons for Alptal, 2005, and Seehornwald, 2009, were chosen based on similarity of regression coefficients to those for all years combined of the respective site. Regres-

15   sion coefficients for these four sites combined are shown as red lines in Fig. 3. Estimates of simulation errors based on these regression coefficients are shown in Fig. 2 and explain 60% of variance in nighttime errors and 59% of variance in daytime errors.

4   Effect of correction in global simulations of CLM4.5

4.1   Sub-canopy longwave radiation – case study Alptal, Switzerland

20   In contrast to other forest stands used to estimate regression coefficients, grid cell and forest stand feature similarly high vegetation densities (PAIs of 3.7 $\mathrm{m^2m^{-2}}$ and 4.1 $\mathrm{m^2m^{-2}}$, respectively) and thus similar vegetation emissivities ε_v (0.975 and 0.983, respectively) for the location of Alptal. This allows for a comparison of diurnal cycles of sub-canopy longwave radiation as well as longwave enhancement between offline simulations and stand-scale measurements. Implementation of correction factors in CLM4.5 results in decreased sub-canopy longwave radiation during daytime and increased sub-canopy longwave radiation

25   during nighttime, thereby reducing diurnal cycles. For the grid cell representing Alptal, diurnal ranges decrease from about 70 $\mathrm{Wm^{-2}}$ to about 30 $\mathrm{Wm^{-2}}$ (Fig. 4a and Fig. 4b). Observations at the forest stand show an average diurnal range of about 15 $\mathrm{Wm^{-2}}$. Simulations and observations display a similar range of intraseasonal variability but do not agree in evolution and daily average of sub-canopy longwave radiation. Implementation of correction factors increases average sub-canopy longwave radiation, seen in Fig. 4b, for two reasons. Firstly, daytime correction depends on insolation, which changes throughout the

30   snow cover season so that daytime correction varies to a higher degree than nighttime correction. Secondly, nights are longer than days prior to the boreal spring equinox, which leads to nighttime increases outweighing daytime decreases. Consequently, correction results in increased average sub-canopy longwave radiation even for equal magnitudes of daytime overestimation and nighttime underestimation.

Comparison of simulated and measured longwave enhancement is shown in Fig. 4c and Fig. 4d for Alptal. As for sub-canopy longwave radiation, the diurnal cycle of simulated longwave enhancement is reduced by implementation of correction factors with increased enhancement at night and decreased enhancement at daytime. Reduction of daytime longwave enhancement increases throughout the snowmelt season, which is due to increasing insolation and thus increasing reduction of sub-canopy

5 longwave radiation during daytime. Longwave enhancement values vary between 1.1 and 1.4 in CTRL, which is predominately driven by diurnal cycles. The diurnal cycle of longwave enhancement is reduced by more than 50% in CORR, resulting in a diurnal range similar to observations and increased daily average longwave enhancement. Simulated longwave enhancement displays little intraseasonal variability with variations mostly due to the overestimated diurnal cycle, which indicates intraseasonal variability in sub-canopy longwave radiation is largely due to variations in atmospheric longwave radiation. In

10 contrast, measured longwave enhancement values range from less than 1 to more than 1.6 and display little diurnal variability but high variability on synoptic timescales, which results in a different daily average of longwave enhancement compared to simulations. Moreover, lower average longwave enhancement for observations indicates more overcast conditions, which lead to smaller diurnal cycles in sub-canopy longwave radiation compared to simulations. Therefore, correction factors improve the realism of diurnal cycles of sub-canopy longwave radiation and longwave enhancement, encouraging usage for evaluation of

15 impact on snow cover.

The contrast in variability between simulated and observed longwave enhancement can be seen in Fig. 5. Observations show a large range of longwave enhancement values that are closely tied to effective emissivity of the sky, which represents clear-sky (low ε_{sky}) and overcast (high ε_{sky}) conditions. Observed longwave enhancement increases for decreasing ε_{sky} as the contrast between vegetation temperatures, increasing due to higher insolation, and effective temperature of the sky increases. Spread

20 in observed longwave enhancement is small throughout the range of ε_{sky}, indicating little diurnal variability and the process of longwave enhancement depending on meteorological conditions. Simulations display a narrow range of ε_{sky}, which causes the lack of intraseasonal variability seen in Fig. 4c. The spread in simulated longwave enhancement values is substantially larger compared to observations for the respective range in ε_{sky} representing overestimated diurnal cycles. Implementation of correction factors reduces the spread in longwave enhancement values and increases average longwave enhancement (see

25 Fig. 4d), however, spread in longwave enhancement is still overestimated and average longwave enhancement is underestimated in CORR compared to observations for the respective range in ε_{sky}.

4.2 Longwave enhancement and limited spatial variability in ε_{sky}

Offline simulations feature a lack of variability in ε_{sky}, as seen for the grid cell of Alptal, across the Northern Hemisphere (not shown here), which results in correction factors being similar spatially and largely dependent on insolation. However,

30 variability in both insolation and diurnal ranges in atmospheric longwave radiation indicate small variations in meteorological forcing that are not represented by ε_{sky}. Therefore, ε_{sky} in simulations may indicate clear-sky conditions even when insolation and atmospheric longwave radiation suggest more overcast conditions, resulting in overestimated correction factors and overcorrection of sub-canopy longwave radiation. This overcorrection results in larger nighttime than daytime values of sub-canopy longwave radiation in contrast to atmospheric longwave radiation and occurs mostly along continental coasts (Fig. 6).

Consequently, a contour line is shown in maps depicting the impact of correction factors in order to denote an overcorrection for 10% of days.

Maps of longwave enhancement beneath evergreen needleleaf forests in CLM4.5 are shown in Fig. 7a and Fig. 7b. Averages over boreal winter and spring show an enhancement of longwave radiation beneath canopies by about 20% to 30% and display

5   little differences across boreal forests, which is due to small spatial variability in both ε_{sky} and vegetation density (Fig. 1). CORR displays increased average longwave enhancement north of 40°N with an additional enhancement of longwave radiation of up to 5% beneath dense boreal forests. Changes in longwave enhancement generally increase with latitude as daytime correction factors vary with insolation while nighttime correction factors are independent of latitude. A higher increase in longwave enhancement can be seen for higher vegetation density within regions covered by boreal forests (Fig. 1b) due to

10   weighting of contributions to subcanopy longwave radiation (Eq. (3)).

4.3 Snow cover and snowmelt

Changes in sub-canopy longwave radiation affect the surface energy balance, which can be seen for grid cell-averaged snow surface temperature (Fig. 7c and Fig. 7d). Simulated average snow surface temperatures are determined by latitude, topography, and continentality, reaching values of less than -40°C in the mountainous regions of northeastern Siberia (Fig. 1c), and range

15   between -20°C and -15°C for boreal forests, the outlines of which can be seen in central Siberia and central North America. The impact of correction factors is limited to grid cells for which vegetation is dominated by evergreen needleleaf trees and implementation results in an increase in average snow surface temperature of up to 2°C. The lack of spatial variability in change due to correction factors is caused by little spatial variability in meteorological conditions as well as high vegetation density and similarly high PFT coverage across boreal forests (Fig. 1a and Fig. 1b).

20   Cold content, the energy required to raise snow temperatures to 0°C, is used to quantify the impact of correction factors on the entire snow column. Average cold content simulated by CLM4.5 mostly reaches values of up to 4 MJm^{-2} and exceeds 5 MJm^{-2} only in glaciated grid cells (Fig. 7e). In CTRL, simulated average cold content ranges between 1.5 MJm^{-2} and 3 MJm^{-2} across boreal forests, with lowest values in northeastern Europe and highest values in eastern Siberia, western Canada, and Quebec. Relative changes in cold content from CTRL to CORR display spatial differences with cold content generally

25   decreasing across boreal forests (Fig. 7f). Reductions in average cold content reach up to 30% in northeastern Europe and western North America and up to 20% in central North America. Across Siberian boreal forests, relative reductions decrease from west to east from more than 20% to about 10%. Spatial differences in relative reductions correspond to spatial differences in average cold content, with higher relative reductions for smaller averages, representing a more even spatial pattern of absolute reductions in cold content as indicated by changes in snow surface temperature (Fig. 7d).

30   Spatial patterns in snow off date are similar to those in cold content as higher cold content corresponds to later meltout (Fig. 7g and Fig. 7h). Changes in snow off date from CTRL to CORR display stark spatial contrasts with meltout happening up to 10 days earlier in central Europe and on the western coast of North America. Meltout is advanced by about 5 days for boreal forests in northeastern Europe and western Siberia and slightly less for boreal forests in central North America. In contrast, meltout is delayed in mountains of southeastern Siberia (Fig. 1c), where meltout occurs late among boreal forests.

Reasons for spatial differences in changes of snow off date across Siberian boreal forests are explored in Fig. 8. Snow off dates are similar spatially in CTRL, likely caused by higher elevations in southeastern Siberia compensating for less cold content, and meltout generally occurs past the boreal spring equinox in northwestern and southeastern Siberia. However, higher insolation for southeastern Siberia results in higher reductions of daytime sub-canopy longwave radiation by correction factors

5 and consequently smaller increases in daily average sub-canopy longwave radiation prior to the boreal spring equinox compared to northwestern Siberia. Although changes in sub-canopy longwave radiation are still positive in southeastern Siberia accumulated over the snow season, causing a decrease in cold content, reduction in daily average sub-canopy longwave radiation by correction factors past the boreal spring equinox cancels out the previous increase and consequently, snowmelt is slightly delayed. In contrast to southeastern Siberia, meltout is slightly accelerated in central North America although both lat-

10 itude and meltout date are similar, as relative reductions in cold content are generally higher. However, differences in changes in meltout date between central North America and southeastern Siberia are minor.

5 Discussion

Todt et al. (2018) found roughly similar magnitudes for daytime overestimations and nighttime underestimations of sub-canopy longwave radiation in CLM4.5, however, this study shows that different durations of day and night over the snow cover season

15 result in a net positive impact of correction on daily averages of sub-canopy longwave radiation. Correction factors change throughout the snowmelt season due to increasing insolation and length of day. Consequently, net impact on daily averages of sub-canopy longwave radiation varies resulting in spatial differences in impact on cold content over the snow cover season and meltout date. Net increase in sub-canopy longwave radiation during the snow cover season is highest for regions of early snowmelt where snow is already comparatively warm, which results in accelerated snowmelt. Lundquist et al. (2013) showed

20 that forests enhance snowmelt compared to open area in regions where winters are warm and mid-winter melt events happen, during which longwave enhancement outweighs shading. Spatial differences in change of meltout date broadly agree with this pattern as the highest acceleration of melt occurs for regions of warmer winters as indicated by snow surface temperatures (Fig. 7c), suggesting that mid-winter melt events could be underestimated by CLM4.5. Conversely, correction of sub-canopy longwave radiation results in slightly delayed snowmelt in southeastern Siberia albeit average cold content over the entire

25 snow cover season being reduced. This delay is due to meltout happening substantially later than the boreal spring equinox and high insolation during the snowmelt period, which result in reduction in daytime sub-canopy longwave radiation outweighing increased sub-canopy longwave radiation during night. Consequently, overestimated diurnal cycles of sub-canopy longwave radiation in CLM4.5 lead to a compressed snowmelt duration across boreal forests in offline simulations.

Comparison of offline simulations of CLM4 with observations have shown CLM4 failing to accurately simulate the timing

30 of both snow ablation and snow accumulation across boreal forests (Thackeray et al., 2014, 2015). Observations show one major peak in reduction of Snow Cover Fraction (SCF) across boreal forests, from April to May, and smaller reductions in SCF in early and late spring. In contrast, CLM4 simulates major reductions in SCF from March to April and April to May but underestimates reductions from February to March and May to June, indicating snowmelt duration is compressed into the

period March to May. Correction of overestimated diurnal cycles of sub-canopy longwave radiation might thus help to expand snowmelt duration across boreal forests by accelerating early snowmelt and delaying late snowmelt. Thackeray et al. (2014, 2015) also showed SCF increasing earlier than observed across boreal forests in CLM4 and, although this study does not focus on the snow accumulation period, processes governing the influence of correction factors are the same as for the snow abla-

5 tion period. As most snowfall occurs past the boreal autumn equinox, when daily average sub-canopy longwave radiation is increased due to correction factors, correction could delay the accumulation of snow across boreal forests. Therefore, overestimated diurnal cycles in sub-canopy longwave radiation also potentially contribute to this deficiency in the simulation of snow cover timing.

Changing seasonality in a warming climate may have implications for snowmelt and longwave enhancement. Future warm-

10 ing will lead to earlier snowmelt, when less energy from insolation is available for melt, which will likely result in lower melt rates (Musselman et al., 2017). A shortened snow season indicates more asymmetrical lengths of day and night during snowmelt and consequently, overestimated diurnal cycles of sub-canopy longwave radiation in CLM4.5 could result in even higher underestimations in daily averages. Moreover, underestimated sub-canopy longwave radiation suggests that CLM4.5 underestimates melt rates in general. In turn, future projections are complex, as corrected and thus increased sub-canopy long-

15 wave radiation might cancel out reduced energy from insolation due to earlier snowmelt. Nonetheless, the contribution of longwave enhancement to snowmelt is likely to increase in the future, further necessitating accurate simulation of sub-canopy longwave radiation.

Implementation of correction factors resulted in realistic average diurnal ranges of sub-canopy longwave radiation and longwave enhancement, but more substantial underestimation than overestimation of longwave enhancement seen in Fig. 5

20 suggests that the impact of shortcomings in CLM4.5 on snow cover and snowmelt might still be underestimated by this study. Gouttevin et al. (2015) and Todt et al. (2018) have shown the implementation of biomass heat storage to result in a net positive impact on sub-canopy longwave radiation as well as a slight reduction of diurnal cycles. This suggests that heat storage by biomass could further reduce nighttime underestimation in CLM4.5 and improve the simulation of sub-canopy longwave radiation and longwave enhancement.

25 6 Conclusions

This study assessed the impact of deficiencies in simulated longwave enhancement by forest canopies on snow cover in CLM4.5. Sub-canopy longwave radiation simulated by CLM4.5's single-layer vegetation was corrected based on the damping effect of multiple canopy layers. Correction factors were derived from forest stand-scale simulations and subsequently implemented for evergreen needleleaf trees in land-only simulations of CLM4.5. Correction reduces overestimated diurnal cycles of

30 sub-canopy longwave radiation by decreasing daytime overestimations and nighttime underestimations. This results in a net increase of sub-canopy longwave radiation over the entire snow cover season, due to longer nights than days. Consequently, correction results in increasing average snow temperatures, indicating that CLM4.5 underestimates snow temperatures due to overestimated diurnal cycles of sub-canopy longwave radiation. Simulations exhibit a spatially uniform underestimation of

snow temperatures by CLM4.5 across evergreen boreal forests. However, impact on meltout timing displays spatial differences that depend on insolation and duration of snow on the ground. The effect of overestimated diurnal cycles on daily average sub-canopy longwave radiation changes throughout the snowmelt season as insolation and length of day increase. Conse-

5 quently, CLM4.5 delays snowmelt more in regions of warmer snow cover and earlier meltout, which results in a shortened snowmelt season across boreal forests. However, spatial variability in impact on snow cover is limited in land-only simulations of CLM4.5 due to a lack of variability in meteorological conditions.

Code and data availability. Code is available on GitHub at https://github.com/mtodt/2018_OfflineSimulations in order to derive correction factors, implement correction factors in CLM4.5, post-process simulations, and create figures shown in this study. Meteorological observations for forest stands are available as follows: (1) on GitHub at https://github.com/mtodt/2018_ToyModel for Alptal and See-

10 hornwald; (2) from the British Atmospheric Data Centre at http://catalogue.ceda.ac.uk/uuid/9c8c86ed78ae4836a336d45cbb6a757c for Sodankylä and http://catalogue.ceda.ac.uk/uuid/6947880b98d32e249a8638ebe768efd2 for Abisko; and (3) from the Arctic Data Center at https://arcticdata.io/catalog/view/doi:10.18739/A2BG2H890 for Cherskiy. Forest stand-scale simulations were performed by Todt et al. (2018) and code is available on GitHub at https://github.com/mtodt/2018_ToyModel.

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

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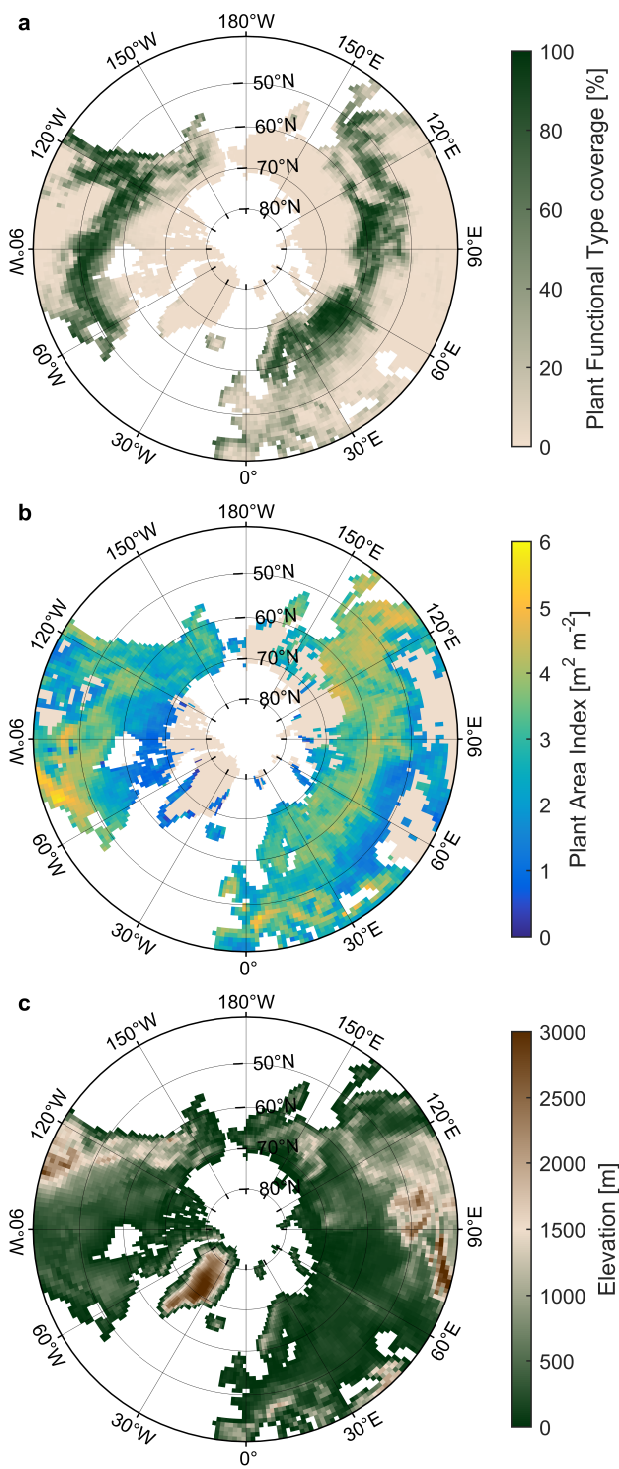
Figure 1. Coverage of vegetated landunit within grid cell by combination of Needleleaf Evergreen Boreal Trees (NEBTs) and Needleleaf Evergreen Temperate Trees (NETTs) (a), Plant Area Index (PAI) for combination of NEBTS and NETTs weighted by PFT fractions (b), and grid-cell average elevation (c) based on CLM4.5's $0.9° \times 1.25°$ surface dataset.
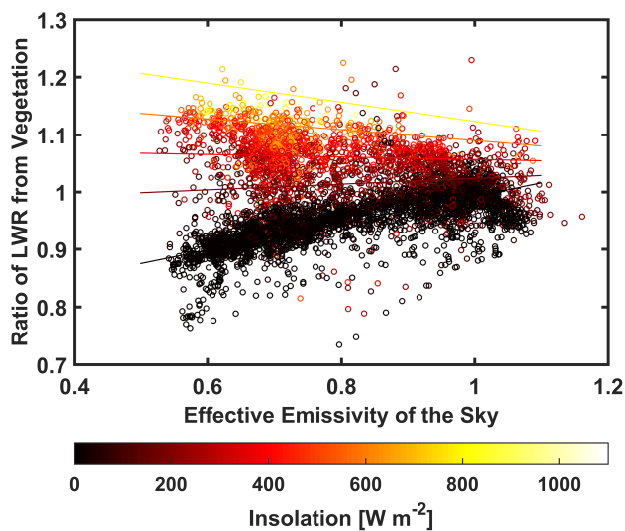
Figure 2. Ratio of longwave radiation emitted from vegetation simulated by CLM4.5 and estimated from forest stand observations as a function of effective emissivity of the sky (abscissa) and insolation (colour) for Alptal (season 2005), Seehornwald (season 2009), Sodankylä, and Cherskiy. Lines represent solutions of Eq. (6) for multiple values of insolation: 0, 200, 400, 600, and 800 Wm^{-2}.
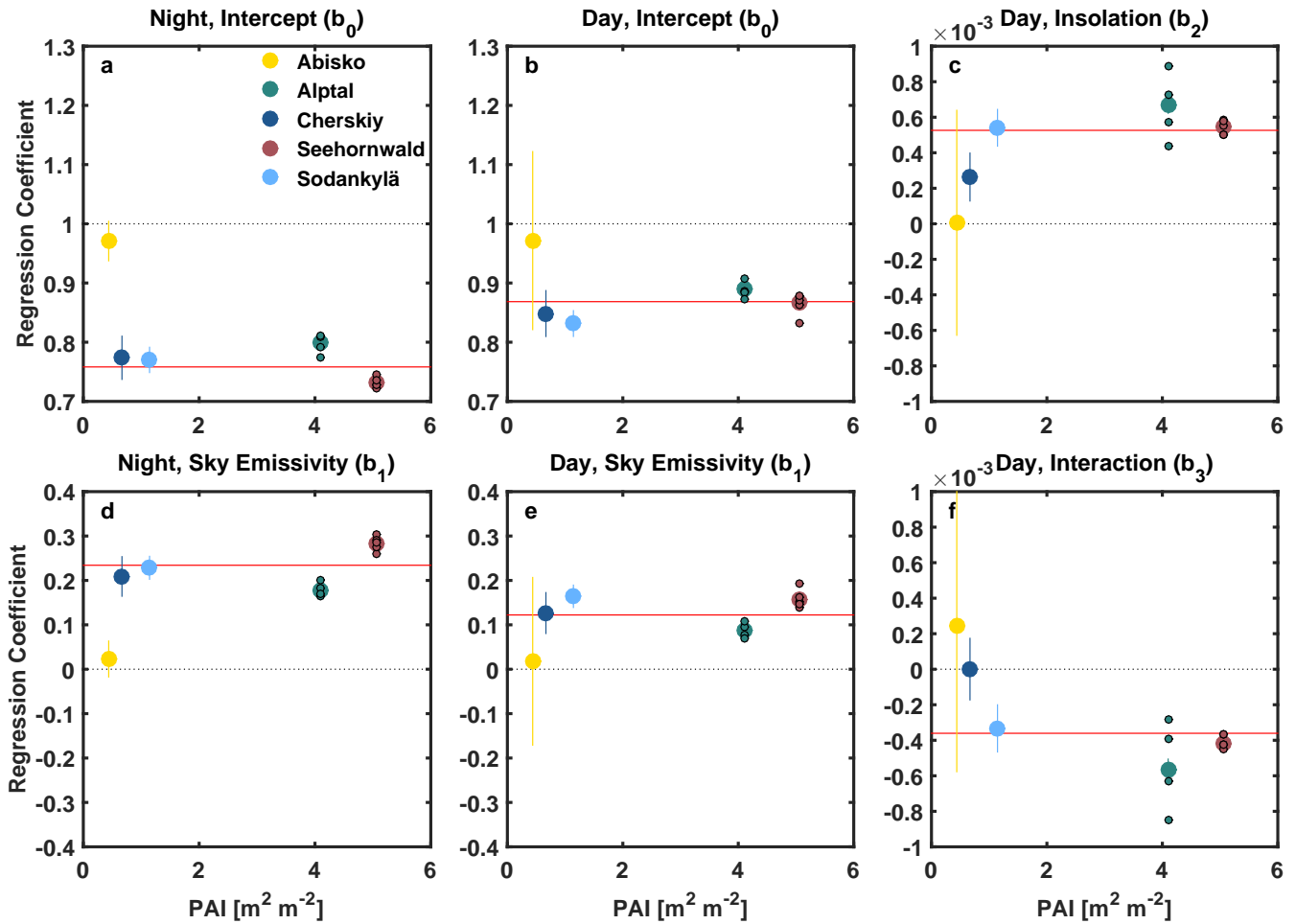
Figure 3. Regression coefficients (Eq. (6)) for forest stands at Abisko (yellow), Alptal (green), Cherskiy (dark blue), Seehornwald (maroon), and Sodankylä (light blue) with small circles indicating individual seasons for Alptal and Seehornwald and solid lines indicating 95%-confidence intervals. Red lines display regression coefficients calculated from a combination of Alptal season 2005, Cherskiy, Seehornwald season 2009, and Sodankylä. Intercept b_0 and regression coefficient for ε_{sky} b_1 are differentiated for night (a and d, respectively) and day (b and e, respectively). Regression coefficient for insolation b_2 and regression coefficient for interaction of ε_{sky} and insolation b_3 are shown for day only (c and f, respectively). Regression coefficients involving insolation have the unit $\mathrm{W}^{-1}\mathrm{m}^2$.
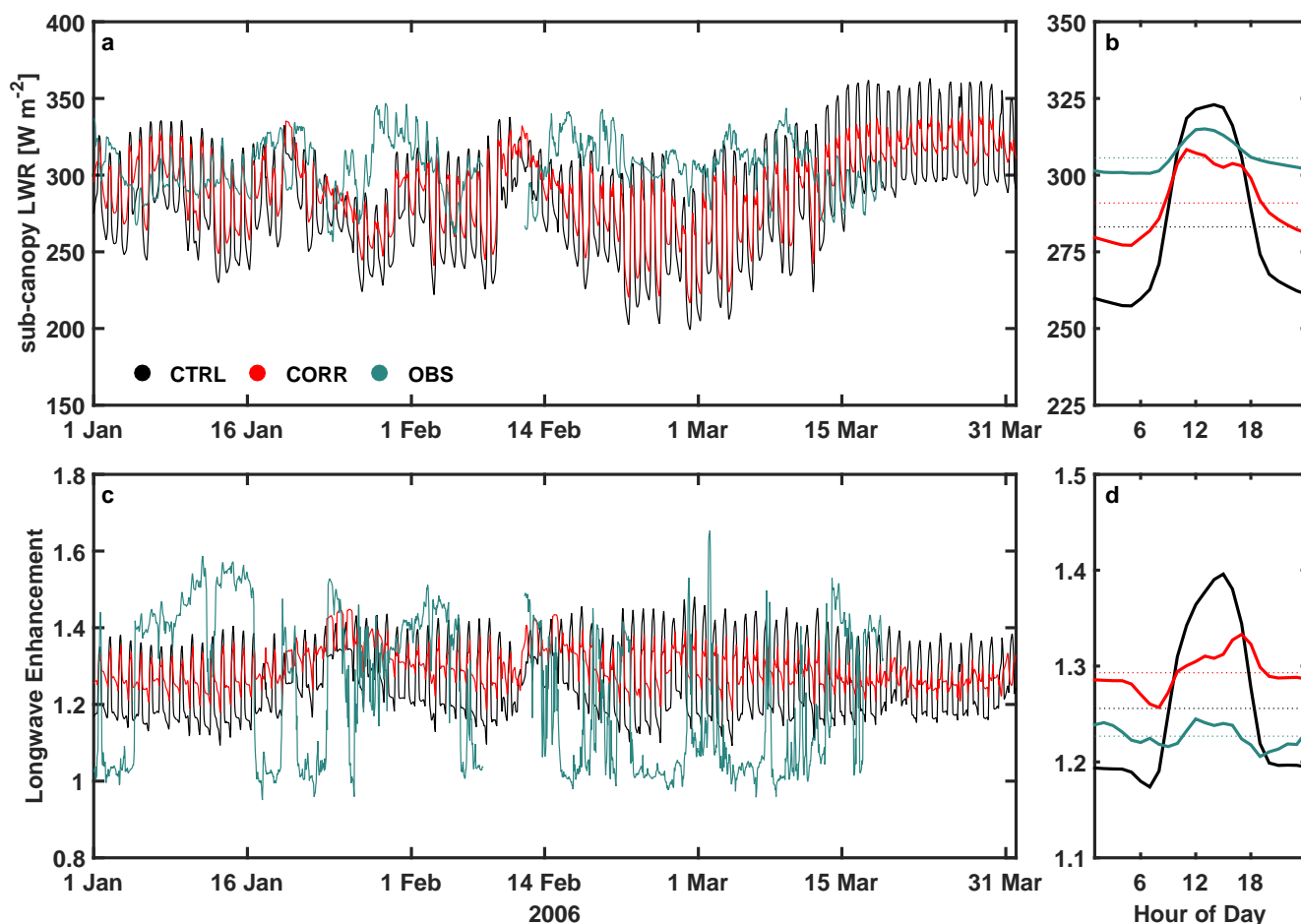
Figure 4. Hourly time series (a, c), diurnal cycles (solid in b, d), and JFM averages (dotted in b, d) of sub-canopy longwave radiation (a, b) and longwave enhancement (c, d) for the snowmelt season in 2006 at Alptal, Switzerland. Measurements at the forest stand (green) are shown for comparison with offline simulations CTRL (black) and CORR (red) for boreal evergreen needleelaf trees in the corresponding gridcell of Alptal, Switzerland. Gaps in measurements are due to quality checks and excluded from calculation of diurnal cycle and JFM average.
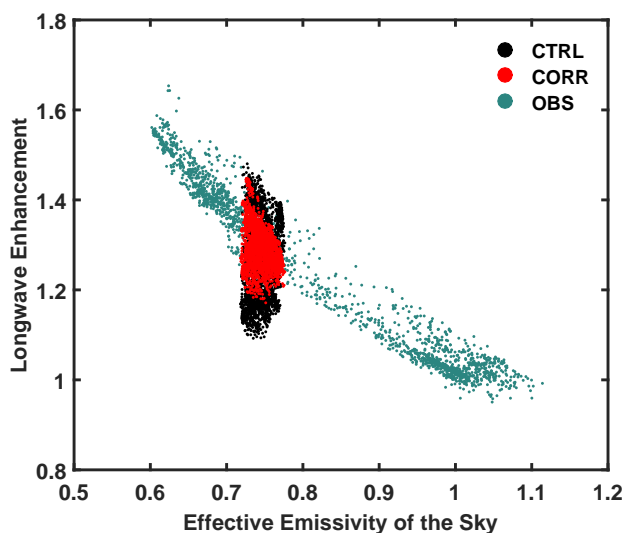
Figure 5. Longwave enhancement measured (green) at the forest stand of Alptal, Switzerland and simulated in CTRL (black) and CORR (red) for boreal evergreen needleleaf trees in the corresponding gridcell of Alptal, Switzerland as a function of effective emissivity of the sky. Each data point represents an hourly average seen in Fig. 4c.
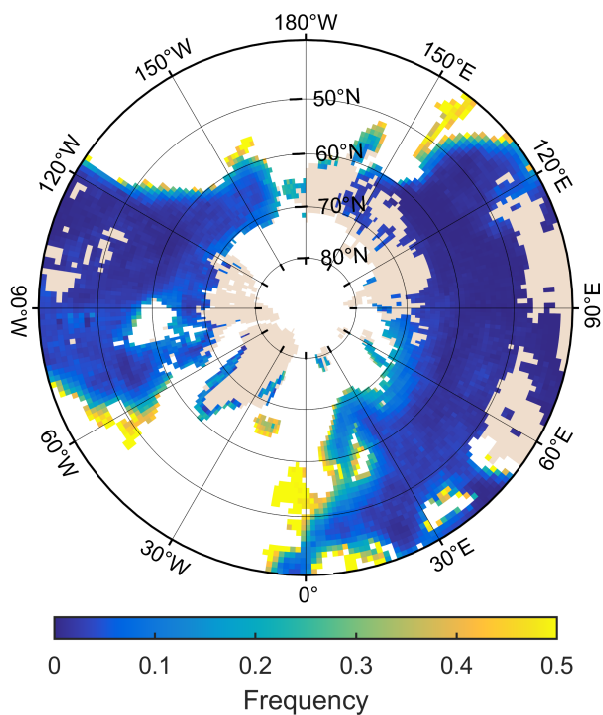
Figure 6. Frequency of days for 2004 - 2007 during which implementation of correction factors results in higher nighttime than daytime sub-canopy longwave radiation despite higher daytime than nighttime atmospheric longwave radiation.
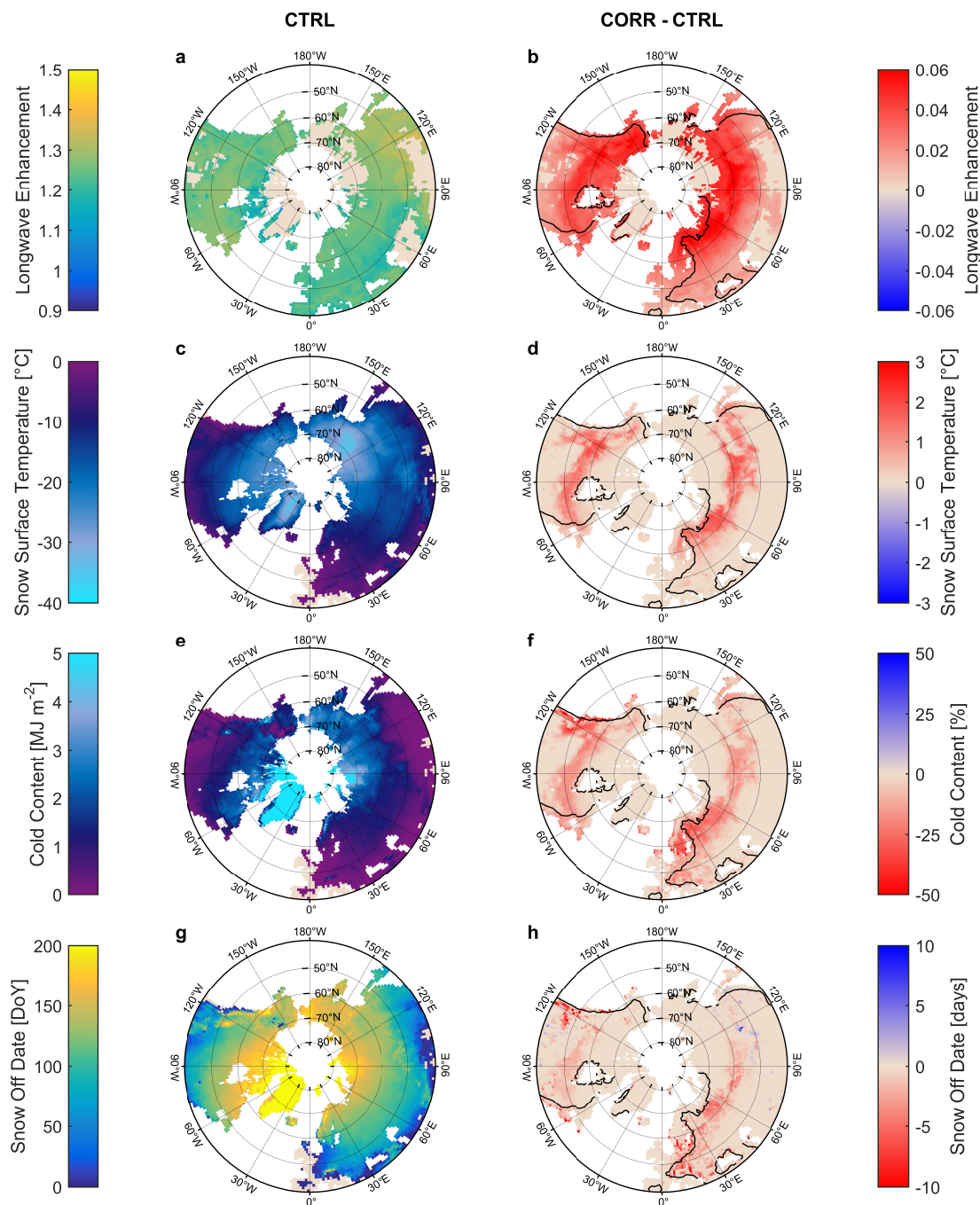
Figure 7. Averages in CTRL (a, c, e, g) and differences between CORR and CTRL (b, d, f, h) for longwave enhancement beneath evergreen needleleaf trees (a, b), snow surface temperature (c, d), cold content (e, f), and snow off date (g, h). Longwave enhancement is averaged over December to May while snow surface temperature and cold content are averaged over entire snow cover seasons. Differences CORR - CTRL are calculated as averages of differences between each individual snow cover season. For panels c-h, a mask is applied to filter out grid cells that are not perennially snow-covered. Black lines demarcate continental areas with less than 10% of overcorrected days.
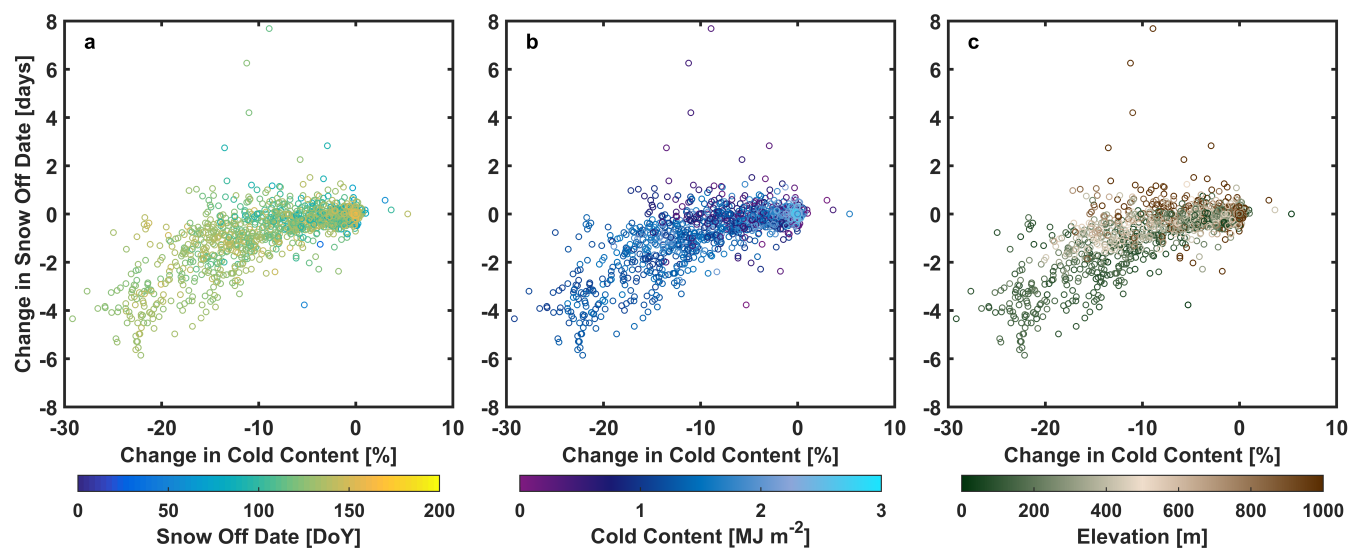
Figure 8. Change in cold content and snow off date from CTRL to CORR as a function of (a) snow off date and (b) cold content in CTRL as well as elevation (c) for grid cells within the area 40°E to 140°E and 42°N to 70°N.