The importance of a surface organic layer in simulating permafrost thermal and carbon dynamics

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Abstract

Permafrost-affected soils contain twice as much carbon as currently exists in the atmosphere. Studies show that warming of the perennially frozen ground could initiate significant release of the frozen soil carbon into the atmosphere. To reduce the uncertainty associated with the modeling of the permafrost carbon feedback it is important to start with the observed soil carbon distribution and to better address permafrost thermal and carbon dynamics. We initialized frozen carbon using the recent Northern Circumpolar Soil Carbon Dataset. To better address permafrost thermal and carbon dynamics we implemented a dynamic surface organic layer with vertical carbon redistribution. In addition, we introduced dynamic root growth controlled by active layer thickness, which improved soil carbon exchange between frozen and thawed pools. Our results indicate that a dynamic surface organic layer improved permafrost thermal dynamics and simulated thaw depth. These improvements allowed us achieve better agreement with the estimated carbon stocks in permafrost-affected soils using historical climate forcing.

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1. Introduction

Warming of the global climate will lead to widespread permafrost thaw and degradation with impacts on ecosystems, infrastructure, and emissions to amplify climate warming (Oberman, 2008; Callaghan et al., 2011; Shuur et al., 2015). Permafrost-affected soils in the high northern latitudes contain 1300±200 Gt of carbon, where about 800 Gt C is preserved frozen in permafrost (Tarnocai et al., 2009; Hugelius et al., 2014). As permafrost thaws, organic matter frozen within permafrost will thaw and decay, which will initiate the permafrost carbon feedback (PCF), releasing an estimated 120±85 Gt of carbon emissions by 2100 (Schaefer et al., 2014). The wide range of estimates of carbon emissions from thawing permafrost depend in large part on the ability of models to simulate present permafrost area extent (Brown et al., 1997). For example, the simulated permafrost in some models is significantly more sensitive to thaw, with corresponding larger estimates of carbon emissions (Koven et al., 2013). Narrowing the uncertainty in estimated carbon emissions requires improvements in how Land Surface Models (LSMs) represent permafrost thermal and carbon dynamics.

The active layer in permafrost regions is the surficial layer overlying the permafrost, which undergoes seasonal freeze-thaw cycles. Active layer thickness (ALT) is the maximum depth of thaw at the end of summer. LSMs used to estimate emissions from thawing permafrost typically assume that the frozen carbon is located in the upper permafrost above 3 meters depth and below the maximum active layer thickness (ALT) (Koven et al., 2011; Schaefer et al., 2011; MacDougall et al., 2012). Thus, the simulated ALT determines the volume of permafrost in the top 3 meters of soil, and thus the initial amount of frozen carbon. Consequently, any biases in the simulated ALT strongly
influence the initial amount of frozen carbon, even if different models initialize the frozen carbon in the same way. Also, the same thermal biases that lead to deeper simulated active layers also lead to warmer soil temperatures, making the simulated permafrost more vulnerable to thaw and resulting in higher emissions estimates (Koven et al., 2013).

The surface organic layer (SOL) is the surface soil layer of nearly pure organic matter that exerts a huge influence on the thermodynamics of the active layer. The organic layer thickness (OLT) usually varies between 5-30 cm, depending on a balance between the litter accumulation rate relative to the organic matter decomposition rate (Yi et al., 2009; Johnstone et al., 2010). Recent model intercomparison study shows that LSMs need more realistic surface processes such as upper organic layer and better representations of subsoil thermal dynamics (Ekici et al., 2014a). The low thermal conductivity of the SOL makes it an effective insulator decreasing the heat exchange between permafrost and the atmosphere (Rinke et al., 2008). The effect of the SOL has been well presented in several modeling studies. For example, Lawrence and Slater (2008) showed that soil organic matter affects the permafrost thermal state in the Community Land Model (CLM), and Jafarov et al., (2012) discussed the effect of the SOL in the regional modeling study for Alaska, United States. Recently, Chadburn et al., (2015a,b) incorporated the SOL in the Joint UK Land Environment Simulator (JULES) model to illustrate its influence on ALT and ground temperatures both at a site specific study in Siberia, Russia, and globally. In essence, the soil temperatures and ALT decrease as the OLT increases. Consequently, how (or if) LSMs represent the SOL in the simulated soil thermodynamics will simultaneously determine the initial amount of frozen permafrost carbon and the vulnerability of the simulated permafrost to thaw.
Here we describe a fully dynamic SOL to demonstrate the importance of coupling soil biogeochemistry and thermodynamics to improve the simulated permafrost temperature and ALT. We improved the Simple Biosphere/Carnegie-Ames-Stanford Approach (SiBCASA) model (Schaefer et al., 2011) by adding a dynamic SOL and limiting plant growth in frozen soils and demonstrate that these changes improve permafrost thermal and carbon dynamics in comparison. Then we used the modified model to evaluate current permafrost carbon stock (Hugelius et al., 2014) under the steady state climate in the early 20th century.

2. Methods

We used the SiBCASA model (Schaefer et al., 2008) to evaluate current soil carbon stocks in permafrost-affected soils. SiBCASA has fully integrated water, energy, and carbon cycles and computes surface energy and carbon fluxes at 10 minute time steps. SiBCASA predicts the moisture content, temperature, and carbon content of the canopy, canopy air space, and soil (Sellers et al., 1996a; Vidale and Stockli, 2005). To calculate plant photosynthesis, the model uses a modified Ball-Berry stomatal conductance model (Ball, 1998; Collatz et al., 1991) coupled to a C3 enzyme kinetic model (Farquhar et al., 1980) and a C4 photosynthesis model (Collatz et al., 1992). It predicts soil organic matter, surface litter, and live biomass (leaves, roots, and wood) in a system of 13 prognostic carbon pools as a function of soil depth (Schaefer et al., 2008). The model biogeochemistry does not account for disturbances, such as fire, and does not include a nitrogen cycle. SiBCASA separately calculates respiration losses due to microbial decay (heterotrophic respiration) and plant growth (autotrophic respiration).
SiBCASA uses a fully coupled soil temperature and hydrology model with explicit treatment of frozen soil water originally from the Community Climate System Model, Version 2.0 (Bonan, 1996; Oleson et al., 2004). To improve simulated soil temperatures and permafrost dynamics, Schaefer et al. (2009) increased the total soil depth to 15 m and added the effects of soil organic matter on soil physical properties. Simulated snow density and depth, and thus thermal conductivity, significantly influence simulated permafrost dynamics, so Schaefer et al. (2009) added the effects of depth hoar and wind compaction on simulated snow density and depth. Recent model developments include improved numerical scheme for frozen soil biogeochemistry (Schaefer and Jafarov, 2015).

We spun SiBCASA up to steady-state initial conditions using an input weather dataset from the Climatic Research Unit National Center for Environmental Predictions (CRUNCEP)\(^1\) (Wei et al, 2014) for the entire permafrost domain in the northern hemisphere (Brown et al., 1997). CRUNCEP is modeled weather data at 0.5x0.5 degree latitude and longitude resolution optimally consistent with a broad array of observations. The CRUNCEP dataset used in this study spans 110 years, from 1901 to 2010. We selected the first 30 years from the CRUNCEP dataset (1901 to 1931) and randomly distributed them over 900 years. To run our simulations we used JANUS High Performance Computing (HPC) Center at University of Colorado at Boulder. The 900-yr time span was chosen in order to make optimal use of the computational time, which allowed us to finish one spinup simulation on JANUS HPC without interruptions.

\(^1\)ftp://nacp.ornl.gov/synthesis/2009/frescati/temp/land_use_change/original/readme.htm

Deleted: The 900-yr time span was chosen in order to make optimal use of the computational time, which allowed us to finish one spinup simulation within 24 hours, the maximum allocated time at the JANUS HPC node to run the model without interruptions.
2.1. Frozen carbon initialization

The Permafrost Carbon Network published a revised Northern Circumpolar Soil Carbon Dataset version 2 (NCSCDv2) (Hugelius et al., 2013). The NCSCDv2 includes soil carbon density maps in permafrost-affected soils available at several spatial resolutions ranging from 0.012° to 1°. The dataset comprise spatially extrapolated soil carbon data from more than 1700 soil core samples. This dataset has three main layers each 1 meter in depth, distributed between ground surface and 3 meter depth.

We placed the frozen carbon within the top three meters of simulated permafrost, ignoring deltaic and loess deposits that are known to extend well beyond 3 meters of depth (Hugelius et al., 2014). The bottom of the permafrost carbon layer is fixed at 3 meters, while the top varies spatially with changes in ALT during the spinup run. Defining the permafrost table as the maximum ALT, we essentially assume that the soil above the permafrost table has thawed frequently enough over thousands of years to decay away all the old carbon.

We initialized frozen carbon between the permafrost table and 3 meters depth using two scenarios: 1) spatially uniform distribution of the frozen carbon throughout the permafrost domain (Schaefer et al., 2011), and 2) observed distribution of the frozen carbon according to the NCSCDv2. It is important to know the “stable” depth of the active layer before initializing frozen carbon. We run the model for several years in order to calculate ALT, and then initialized frozen carbon below the maximum calculated ALT.

The frozen carbon was initialized only once during the first equilibrium run cycle. For the next equilibrium run we used the previously calculated permafrost carbon. We defined
an equilibrium point when changes in overall permafrost carbon were negligible or almost zero.

The total initial frozen carbon in each soil layer between the permafrost table and 3 meters is

\[ C_{fr}^i = \rho_c \Delta z_i, \]  

(1)

where \( C_{fr}^i \) is the total permafrost carbon within the \( i \)th soil layer, \( \rho_c \) is the permafrost carbon density, and \( \Delta z_i \) is the thickness of the \( i \)th soil layer in the model. For the uniform permafrost carbon distribution, \( \rho_c = 21 \) kg C m\(^{-3}\) assumed to be spatially and vertically uniform (Schaefer et al., 2011). For the observed distribution from the NCSCDv2, \( \rho_c \) varies both with location and depth (Hugelius et al., 2013).

The carbon in each layer is divided into three pools as follows:

\[
\begin{align*}
C_{slow}^i &= 0.8 C_{fr}^i \\
C_{met}^i &= 0.2 f_{root2met} C_{fr}^i \\
C_{str}^i &= 0.2 f_{root2str} C_{fr}^i,
\end{align*}
\]

(2)

where \( f_{root2met} \) and \( f_{root2str} \) are the simulated fractions of root pool losses to the soil metabolic and structural pools respectively (Schaefer et al., 2008). The nominal turnover time is 5 years for the slow pool, 76 days for the structural pool, and 20 days for the metabolic pool. Schaefer et al. (2011) has a 5% loss to the metabolic pool and a 15% loss to the structural pool based on observed values in Dutta et al. (2006). The simulated fractions are actually 5.6% to the metabolic pool and 14.4% to the structural pool. We found it encouraging that the numbers calculated with the SiBCASA metabolic fractions resulted in numbers that are close to the observed values in Dutta et al. (2006).
2.2. Dynamic SOL

We modified SiBCASA to include a dynamic SOL by incorporating the vertical redistribution of organic material associated with soil accumulation. SiBCASA already accounted for the effects of organic matter on soil properties like porosity, hydraulic conductivity, and thermal conductivity. The physical properties include soil porosity, hydraulic conductivity, heat capacity, thermal conductivity, and matric potential. The model calculates the organic fraction used in the weighted mean as the ratio of simulated carbon density to the density of pure organic matter. SiBCASA does not account for the compression of organic matter. Since the prognostic soil carbon pools vary with depth and time, the organic fraction and the physical properties all vary with time and depth. We only summarized these calculations here since the calculations are covered in detail in Schaefer et al. (2009).

Each model layer has a complete set of prognostic soil carbon pools. The previous version of the model distributed fine and coarse root growth vertically within the soil column based on observed root distributions. As the roots die, carbon is transferred to the soil carbon pools for that layer. Thus, the maximum rooting depth determined the maximum depth of ‘current’ or ‘active’ carbon in the model. Of course, if the maximum rooting depth fell below the permafrost table, the model would accumulate permafrost carbon. The current version of the model initializes the permafrost carbon by assigning carbon to the soil carbon pools below the maximum thaw depth. These frozen pools remained inactive until the layer thaws. As live, above-ground biomass in the model dies, carbon is transferred into the first layer as litter. Without the vertical redistribution we describe here to create a surface organic layer, the top layer of the model tended to
accumulate carbon in excess of that expected for pure organic matter.

To allow vertical movement and build up a SOL, we placed a maximum limit on the amount of organic material that each soil layer can hold. When the simulated carbon content exceeds this threshold, the excess carbon is transferred to the layer below. This is a simplified version of the Koven et al. (2009) carbon diffusion model, which accounts for all sedimentation and cryoturbation processes, while we wanted to limit our model only to the buildup of a SOL.

We calculate the maximum allowed carbon content per soil layer, $C_{\text{max}}$, as

$$C_{\text{max}} = \rho_{\text{max}} \Delta z \frac{1000}{MW_C},$$

(3)

where $\rho_{\text{max}}$ is the maximum density of pure organic matter or peat, $\Delta z$ is the soil layer thickness (m), $MW_C$ is the molecular weight of carbon (12 g mol$^{-1}$), and the factor of $10^3$ converts from grams to kilograms. Based on observations of bulk densities of peat, we assume $\rho_{\text{max}}$ is 140 kg m$^{-3}$ (Price et al., 2005). The $MW_C$ term converts the expression into mol C m$^{-2}$, the SiBCASA internal units for carbon. The simulated organic soil fraction per soil layer, $f_{\text{org}}$, is defined as

$$f_{\text{org}} = \frac{C}{C_{\text{max}}},$$

(4)

where $C$ is the carbon content per soil layer (mol m$^{-2}$). To convert to carbon we assume $f_c$ is 0.5, which means that half of the organic matter by mass is carbon. The original formulation allowed $f_{\text{org}}$ to exceed 1.0 such that the excess organic material was essentially ‘compressed’ into the top soil layer, resulting in a 2-cm simulated SOL. We place an upper limit of 0.95 on $f_{\text{org}}$ and transfer the excess carbon to the layer below. The OLT is defined as the bottom of the lowest soil layer where $f_{\text{org}}$ is 0.95.
2.3. Root growth and soil thermal factor

Fine roots supply nutrients and water for photosynthesis, so essentially the leaves and roots together define the photosynthetic capacity of the plant. Plants have optimized carbon allocation to grow only enough fine roots to properly supply the leaves with the correct amount of water and nutrients to support photosynthesis. So, as plants grow new leaves, they also grow new fine roots to supply them with nutrients and water. Linking root growth to leaf growth is a convenient and simple way to represent this coupling in SiBCASA.

Here we take this coupling one step further and recognize that frozen soil reduces the plant photosynthetic capacity and regulates root and leaf growth. Plants cannot photosynthesize in frozen soil. Frozen soil in the root zone reduces the photosynthetic capacity of the plant by limiting the water available for photosynthesis. Roots cannot grow while soil is frozen and if roots can’t grow, leaves can’t grow. The changes we implement link soil thermodynamics to root growth, leaf growth, and plant photosynthesis.

In the original formulation (without dynamic SOL), plant photosynthesis, leaf growth, and fine root growth were controlled primarily by canopy air space temperature: when the canopy air temperature exceeded 0°C, leaves and roots started to grow. SiBCASA assumes fine root growth decreases exponentially with depth based on observed vertical root distributions (Schaefer et al., 2008) with 90% of fine root growth occurring in the top 1 meter of soil. Consequently, the vertical distribution of new root growth between the soil layers is prescribed using exponential curve fits to observed vertical root
distributions. Before the changes we describe here, the maximum rooting depth sometimes exceeded the thaw depth in permafrost soils, resulting in root growth directly in permafrost, which resulted in false permafrost carbon accumulation since growing roots cannot penetrate frozen soil. The roots that grew into the permafrost never thawed and soon died, but never decayed, resulting in an unrealistic buildup of carbon in the upper layers of permafrost. This, in turn, set up a feedback where the unrealistic increase in organic matter in the simulated permafrost changed the thermodynamic properties and decreased the ALT, resulting in additional carbon buildup. To solve this problem, we kept the original exponential vertical rooting profile, but set maximum rooting depth equal to the thaw depth. This allowed the maximum rooting depth to vary with time and effectively restricted all root growth to within the thawed portion of the active layer.

Soil thaw always lags behind warming of the canopy. Photosynthesis is limited by water availability as well as canopy temperature and starts later in spring after the surface soil layers thaw out. Before the changes described below, leafout and new root growth occurred as many as 60 days before the start of photosynthesis. In reality, leafout, root growth, and the start of photosynthesis should occur at the same time.

We synchronized leafout, root growth, and photosynthesis by restricting root growth to occur only in thawed soil layers. In SiBCASA, leaf growth is linked to fine root growth (Schaefer et al., 2008), so this also delays spring leafout until the soil begins to thaw. We first calculated the fraction of thawed roots:

\[ R_{th} = \sum_{k=1}^{n_{\text{root}}} R_f (1 - F_{\text{ice}}), \]

where \( R_{th} \) is the fraction of total roots that are thawed, \( n_{\text{root}} \) is the deepest soil layer with roots, \( R_f \) is the reference root fraction per soil layer based on observed root distributions.
(Jackson et al., 1996), and $F_{\text{ice}}$ is the ice fraction per soil layer. $F_{\text{ice}}$ is calculated from the liquid water and ice content of each soil layer, both of which are prognostic variables, accounting for latent heat effects. $F_{\text{ice}}$ varies from zero for a completely thawed soil layer to one for a completely frozen soil layer. This assumes that water in each layer is evenly distributed such that $F_{\text{ice}}$ equals the frozen fraction.

Restricting root growth within the thawed portion of the active layer results from the fact that roots cannot penetrate frozen soil. Root growth still decreases exponentially with depth, but we used an effective rooting depth equal to the thaw depth or the theoretical maximum rooting depth from the exponential soil distributions (whichever is less). We calculated an effective root fraction $R_{\text{eff}}$, to control the vertical distribution of new growth carbon within the soil column:

$$R_{\text{eff}} = R_f (1 - F_{\text{ice}})/R_{\text{th}}.$$  \hfill (7)

Dividing by $R_{\text{th}}$ ensures that $R_{\text{eff}}$ sums to one within the soil column, which essentially ensures that all new root growth is distributed only within the thawed portion of the soil column. Here, we replaced $R_f$ with $R_{\text{eff}}$ in the vertical distribution of coarse woody roots in the wood pool and the fine roots in the root pool and associated calculations of autotrophic respiration.

To synchronize growth primary productivity (GPP) with leafout, we treated the reference vertical root distribution, $R_f$, as the potential root growth defining the maximum potential GPP. Since the sum of $R_f$ is always one, when $R_{\text{th}} < 1$, GPP must be less than its full potential. We defined a GPP scaling factor, $S_{\text{soilfrz}}$, as

$$S_{\text{soilfrz}} = \begin{cases} R_{\text{th}}, & R_{\text{th}} \geq 0.01 \\ 0, & R_{\text{th}} < 0.01 \end{cases}.$$  \hfill (7)
This assumes that at least 1% of the roots must be thawed for GPP to occur, corresponding to about ~1 cm of thawed soil. $S_{\text{soilfrz}}$ is applied along with the soil moisture and canopy temperature scaling factors to constrain photosynthesis while soil is frozen (Schaefer et al., 2008). To constrain wood growth, we applied $S_{\text{soilfrz}}$ to the temperature scaling factors to the decay rate constant that controls wood growth

$$k_{\text{eff}} = S_{\text{nsc}}S_{T}S_{M}S_{\text{frost}}S_{\text{soilfrz}}k_{\text{wood}},$$

(9)

where $k_{\text{eff}}$ is the effective growth rate, $S_{\text{nsc}}$ is the non-structural carbohydrate scaling factor, $S_{T}$ is the canopy temperature scaling factor, $S_{\text{frost}}$ is the frost inhibition function, and $k_{\text{wood}}$ is the reference wood growth rate. $S_{T}$, $S_{\text{frost}}$, and $S_{\text{soilfrz}}$ are the same scaling factors that control GPP under the assumption that the factors that control photosynthesis also control wood growth (Schaefer et al., 2008). This is also consistent with what we normally see in discontinuous permafrost zones: trees cannot grow in shallow permafrost. Indeed, one can often detect the presence of permafrost in the discontinuous zone simply by noting the lack of trees.

To constrain leaf growth, we added $S_{\text{soilfrz}}$ to the frozen leaf scaling factor

$$S_{\text{leaffrz}} = S_{\text{soilfrz}}(1 + \exp(1.3(273 - T_{\text{can}}))),$$

(10)

where $S_{\text{leaffrz}}$ is the frozen leaf scaling factor and $T_{\text{can}}$ is canopy temperature.

3. Results

The dynamic SOL decreased the simulated ALT on average 50% across the domain and allowed the model to simulate permafrost in discontinuous zones where it could not before (Figure 1). The area of near surface permafrost simulated with the current version
of the model equals to 13.5 mil km$^2$ which is almost 38% greater than without the
dynamic SOL (Schaefer et al., 2011). This area is closer to the observation from the
International Permafrost Association which is about 16.2 mil km$^2$ (Brown et al., 1997).
Simulated ALT less than 2 m covers about 92% of the area in the new simulations
(Figure 1B) in comparison to 66% of the area in the Schaefer et al. (2011) simulations
(Figure 1A). The previous version of SiBCASA could not simulate permafrost in many
parts of the discontinuous zone with relatively warm climate. Adding the dynamic SOL
essentially decreased the thermal conductivity of the surface soil to allow SiBCASA to
simulate permafrost where the mean annual air temperatures (MAAT) are close to 0 °C.

To illustrate the improvement of the simulated ALT with respect to the observed
data, we compared simulated ALT with measured values from Circumpolar Active Layer
Monitoring (CALM) stations. The CALM network is a part of the Global Terrestrial
Network for Permafrost (GTN-P) (Burgess et al., 2000). The monitoring network
measures ALT either using a mechanical probe or a vertical array of temperature sensors
(Brown et al., 2000; Shiklomanov et al., 2010). After matching up the CALM coordinates
with the coordinates of previously simulated ALT (Schaefer et al., 2011), we excluded
sites with no measurements or ALT greater than 3m depth, ending up with 76 CALM
stations. Figure 2 shows simulated vs. observed ALT for the 76 CALM sites. The current
simulations have a higher resolution than Schaefer et al. (2011) simulations, which
allowed us to reach a higher order of heterogeneity between measured and simulated
ALTs. The Pearson’s correlation coefficient, R, is negative and not significant for the
Schaefer et al. (2011) simulations (Figure 2A), but is positive and statistically significant.
for the current simulations assuming p< 0.05 (Figure 2B). The dynamic SOL greatly improves the simulated ALT, but SiBCASA still tends to overestimate ALT.

Figure 3 illustrates the effect of the frozen soil restrictions on phenology and GPP at a single point in central Siberia. Before applying a frozen soil restriction, SiBCASA maintained fine roots even in winter, resulting in root growth all year with a strong peak in spring corresponding to simulated leafout (Figure 3A). Simulated GPP was restricted by liquid water availability and was closely tied to thawing of the active layer, resulting in a lag as high as 60 days between leafout and start of GPP in spring. Restricting growth and GPP to when the soil is thawed essentially synchronizes all phenological events to occur at the same time (Figure 3B).

In the previous version without a dynamic SOL the ALT was generally deep in forest biomes, but in the new version there is a thick SOL (due to high GPP), which leads to a shallower ALT. Without restricting root growth within only thawed part of the soil the shallower ALT feedback leads to a significant amount of root growth in the permafrost itself, which puts carbon directly into the permafrost stores. This is unrealistic since growing roots cannot penetrate frozen soil, so the frozen soil restrictions on GPP and root growth together eliminate this problem.

Restricting growth and GPP to when the soil is thawed delayed the onset of plant photosynthesis in spring in permafrost-affected regions. Introduction of the thawed root fraction in the model reduced GPP primarily in early spring. To illustrate the difference between unconstrained and restricted root growth (Figure 3), we ran the model for ten years for both cases. The difference between unconstrained and restricted root growth cases (Figure 4) indicates an overall ~9% reduction in GPP for the entire permafrost
domain, nearly all of which occurred in spring.

To illustrate soil carbon distribution with depth we selected three representative areas: a continuous permafrost area corresponding to tundra type biome above the Arctic circle, an area in the boundary of continuous and discontinuous permafrost corresponding to the boreal forest biome, and an area near the south border of the discontinuous permafrost corresponding to poorly vegetated-rocky areas. We calculated mean and standard deviation of the carbon density distribution with depth for 200 grid points around each of the three selected locations. Simulated typical carbon densities from selected locations are shown on Figure 5. All profiles shown on Figure 5 show a similar pattern: a 20-30 cm SOL with reduced carbon content at the bottom of the active layer. In contrast, the observed vertical carbon profiles show fairly uniform carbon density with depth throughout the active layer and into the permafrost Harden et al., (2012). SiBCASA lacks the cryotubation processes such as cryotic mixing that would redistribute carbon within the active layer. As a result, the carbon at the bottom of the active layer decayed and respired away during spinup.

The decrease in ALT resulting from a dynamic SOL increases the volume of permafrost in the top 3 meters of soil, greatly increasing the initial amount of frozen permafrost carbon in the simulations. Schaefer et al. (2011) without the dynamic SOL assumed a uniform permafrost carbon density of \( 21 \text{ kg} \cdot \text{C} \cdot \text{m}^{-3} \), resulting in a total of 313 Gt of permafrost carbon at the start of their transient run (Figure 6A). To compare the overall permafrost carbon storage, we equilibrated the current version of the model assuming similar uniform distribution (Figure 6B). Assuming the same uniform carbon density, the current version with the dynamic SOL results in a total of \(~680\) Gt C.
compared to 313 GtC in Schaefer et al. (2011). The dynamic SOL effectively doubled the volume of permafrost in the top three meters of soil and the amount of simulated frozen carbon.

Prescribing permafrost carbon according to the NCSDC dataset allowed us to better match with the observed pattern in the soil carbon. However, it does not mean that after the spinup simulated permafrost carbon stocks exactly matched the NCSDC data. During spinup, ALT varies with time, introducing carbon movement from frozen to thawed pools. In discontinuous zones, if the model simulated permafrost, it tended to produce a deeper ALT and thus less permafrost carbon than the NCSCD. The major difference between uniform frozen carbon initialization (Fig 7A) and initialization according to the NCSCD (Fig 7B) is that SiBCASA simulated permafrost in more places. However, the NCSCD map (Fig 7B) shows that not all frozen soil contains a uniform amount of frozen carbon. Therefore simulating ‘correct’ ALT is important and should improve the overall permafrost carbon storage.

Initializing SiBCASA with the observed spatial distribution of permafrost carbon from the NCSCDv2 resulted in ~560 GtC of carbon stored in permafrost after spinup. SiBCASA underestimated the SOC in the Eastern Canada and Western Siberia, and overestimated SOC in Central Siberia (Figure 7A and B). Failure to simulate soil carbon in South-East Canada and South-West Siberia (Figure 7C) could be attributed to deep active layer thickness. The overestimation of the SOC in Central Siberia is a result of coupling between GPP and ALT. The overall amount of soil frozen carbon is less than that calculated assuming uniform frozen carbon distribution. It is important to note that the SOL, ALT, and the permafrost thickness are the same for both cases (Figure 7A and B).
This is due to the fact that in both cases soil carbon is added in the permafrost layer below the active layer. Consequently, the amount of soil carbon in the active layer stays does not change between simulations and has the same thermal and carbon dynamics, and thus ALT. The smaller permafrost carbon stock simulated for the non-uniform case is mainly due to the fact that we did not initialize frozen carbon in regions where according to the NCSCDv2 it is not present, such as the Brooks Range in Alaska.

4. Discussion

The dynamic SOL insulates ALT from air temperature, allowing SiBCASA to simulate permafrost in many discontinuous permafrost regions where it could not before. This result complements similar findings by Lawrence and Slater (2008), Yi et al., (2009), Ekici et al., (2014b), and Chadburn et al., (2015a,b), when changes in thermal properties associated with the presence of soil organic matter cooled the ground. In southeastern Canada and southwestern Siberia, SiBCASA simulates ALT up to 3 meter, and therefore almost no frozen carbon. For example, observed mean annual ground temperatures within southeast Canada region ranges from below to above 0 °C (Smith, and Burgess, 2000), which suggests that the actual permafrost distribution and associated ALT in these regions would be highly heterogeneous. Models like SiBCASA cannot capture such sub-grid heterogeneity, resulting in a deeper, uniform ALT across the grid cell.

To address the effect of different environmental factors we correlated ALT with near surface air temperature (NSAT), down-welling long-wave radiation (DLWR), snow depth (SND), and soil wetness fraction (SWF). The NSAT has a significant effect on the ALT (Camill 2005, Gallaghan et al., 2011). To show this influence, we averaged NSAT
in early fall, for two months September and October over 10 years (Figure 8A). The areas with deep ALT (Figure 1B) fall into the regions where NSATs are greater than one degree centigrade and greater than 5 °C in the south-east Canada. Figure 9A shows the correlation between NSAT and ALT, which indicates clear relation between NSAT and ALT.

The DLWR averaged over 10 years showed higher radiation along the south boundaries of the domain, in particular southeast Canada and southwest Siberia (Figure 8B). The effect of the DLWR on the ALT is more scattered as opposed to NSAT (Figure 9B). However, in general it showed behaviour similar to NSAT.

SND is another important factor contributing to the permafrost thermal state. Zhang (2005) indicates that SND less than 50cm have the greatest impact on soil temperatures. Figure 8C shows maximum simulated snow depth calculated over the last 10 years of the steady state run. The snow effect on the soil thermal state is less obvious and highly dependent on different physical processes, such as wind, snow metamorphism, and depth hoar formation (Sturm et al., 1997). Ekici et al., (2014) confirm nonlinear behavior of snow after modeling soil temperatures for four sites using six CLM model. Similarly, Jafarov et al., (2014) shows that snow thermal properties not always regulated by the SND. Figure 9C also indicate no correlation with ALT.

Figure 8D shows an averaged soil wetness map, which indicates high SWF in both regions where model simulates deep ALT. This suggests that SOL does not provide enough protection for permafrost in regions with wet soils and mild air temperatures, which complements similar funding by Lawrence and Slater (2008). The calculated
Partial correlation between SWF and ALT indicate a clear relationship between them. Figure 9D confirms this statement showing that wetter soils associated with higher ALT.

Before implementing the dynamic SOL, the maximum rooting depth only occasionally fell below the permafrost table. However, after implementing the dynamic SOL, the simulated ALT decreased and new root growth was placed directly into the permafrost with no chance to decay. This phenomenon occurred primarily in the mixed deciduous evergreen forest in south-central Siberia and resulted in a long-term carbon sink into the permafrost carbon pool. It resulted from the fact that the maximum rooting depth determined by the fixed, exponential root distribution incorrectly extended into the permafrost. In permafrost-affected soils, seasonal root growth is largely regulated by the soil thermal conditions (Tryon and Chapin 1983, Van Cleve et al., 1983). Therefore in the LSMs it is important to restrict root growth to thawed soil layers only. Moreover, previous studies showed that the date of snowmelt usually determines the start date of the growing season and the start of active layer thawing (Grøndahl et al. 2007; Wipf and Rixen 2010). Restricting GPP and all growth using the scaling factors described above synchronizes the simulated start of the growing season.

The ability of the ecosystem and climate models to reproduce current frozen soil carbon distribution is important and could reduce uncertainty associated with modeling of the permafrost carbon feedback. Simulated permafrost vulnerability is tightly coupled with the accurate modeling of the present permafrost distribution, which depends on soil thermal properties. We calculate soil thermal properties based on prognostic soil carbon and soil texture from the Harmonized World Soil Database (HWSD) (FAO et al., 2009). Observations indicate that soils in the southeast Canada have high soil carbon as a result.
of a large number of peat lands (Hugelius et al., 2014). Peat has low thermal conductivity and could preserve permafrost even at NSATs about zero degrees centigrade (Jafarov et al., 2012) even if the surrounding areas do not have permafrost. However, the HWSD input data does not have enough soil carbon in the southeast Canada and southwest Russia, as a result, we could not simulate permafrost in those regions.

Including dynamic SOL in the model allows us to study the interaction of plant dynamics and soil thermodynamics. In addition it allows us to study other processes in the future, such as fire impacts on soil thermodynamics and recovery from fire, both of which are strongly influenced by the changes in the SOL (Jafarov et al., 2013). For example, Yuan et al. (2012) evaluated the role of wildfire in soil thermal dynamics and ecosystem carbon in Yukon River Basin of Alaska using Terrestrial Ecosystem Model (Yi et al., 2010), showing wildfires and climate change could substantially alter soil carbon storage. The current version of the model does not include the effects of fire, which means that topsoil carbon stays in the system and provides resilience to permafrost. However, in reality, upper SOL could be removed by fire, which would alter soil thermal properties and perturb permafrost carbon stability.

5. Conclusion

Presence of the SOL improves the permafrost thermal dynamics by reducing heat exchange between near surface atmosphere and subsurface. Similarly, to Koven et al. (2009) we show that inclusion of the surface carbon layer dynamics into the model leads to an improved agreement with the estimated carbon stocks in permafrost-affected soils. However, to better simulate known permafrost distribution in the discontinuous climate, such as southeastern Canada and southwestern Siberia (Fig 1A). We surmise that the CRUNCEP data may not place the 0°C MAAT isotherm at the correct latitudes to match with the southern edge of discontinuous permafrost. The carbon rich soils in these regions may have already begun to thaw, which means that these regions may be currently respiring to the atmosphere and will continue to respire during this century.
permafrost zone it is important to know the exact thickness of the upper soil organic layer. Our setup does not include thick organic layer along the southeastern boundaries of the permafrost domain in Canada as well as southwestern part of Russia, which did not allow the model to simulate permafrost in those regions.

The simplified scheme of the soil carbon dynamics improves permafrost resilience, but does not fully reproduce observed carbon distribution with depth (Harden et al., 2012). The dynamic SOL and rooting depth strengthens the feedback between GPP and ALT (Koven et al., 2009). Higher GPP produces greater litter fall, which increases the input soil carbon at the surface and results in a thicker SOL. The dynamic SOL changes the properties of the near surface soil, resulting in a shallower ALT and cooler soil temperatures. The dynamic rooting depth accounts for a shallower ALT and modulates GPP accordingly. The cooler soil temperatures slow microbial decay and increase the carbon accumulation rate, which in turn increases the SOL and reduces ALT further. Eventually, this feedback results in the development of a peat bog. The changes we describe here indicate that SiBCASA can simulate the dynamics of peat bog development, but the model does not yet include a dynamic vegetation model to account for conversions between biome types, such as boreal forest to peat bog.

6. Acknowledgements

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Software tools used in this study include m_map MATLAB package and shadedErrorBar.m MATLAB script.

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Figure 1. Maximum thaw depth averaged over last five years after spinup from A) Schaefer et al., (2011) and B) this study, in meters.

Figure 2. Comparison of the mean active layer thickness (ALT) from 76 Circumpolar Active Layer Monitoring stations with the averaged ALT from last five years after spinup from A) Schaefer et al., (2011) and B) this study. r is a Pearson’s correlation coefficient and p is a significance value, p<0.05 stands for the 95% of confidence.
Figure 3: A) and B) root growth without and with the frozen soil constraint on growth.

Figure 4. The difference between GPP without and with freezing constraint averaged over ten years.
Figure 5. An averaged soil carbon distribution from 200 grid cells A) for the tundra region in continuous permafrost zone, B) for the boreal forest on the boundary between continuous and discontinuous zones, and C) for the low carbon soil at the south border of the discontinuous permafrost zone. The solid blue curve indicates the mean the white blue shading indicate the spread in the soil carbon density.

Figure 6. The frozen carbon maps obtained assuming uniform frozen carbon distribution at the initial time step, and averaged over five years at the end of the steady state run: A) from Schaefer et al., (2011), and B) from the current run, correspondingly.
Figure 7. The soil carbon maps averaged over top 3 meters: A) from SiBCASA at the end of the steady state run, with constant permafrost density, B) from SiBCASA at the end of the steady state run, with non-constant permafrost density, and C) from the NCSCDv2, correspondingly.
Figure 8. A) The near air temperature for averaged over first two month of the fall season. B) The down-welling long-wave radiation, averaged yearly over 10 years. C) The maximum snow depth obtained over 10 years for the steady state run, and D) the soil wetness fraction (dimensionless fraction of 1), representing overall near-surface soil wetness, averaged yearly over 10 years.
Figure 9. The correlation between ALT and: A) near air temperature for averaged over first two month of the fall season, and B) the down-welling long-wave radiation, averaged yearly over 10 years. C) the maximum snow depth over 10 years for the steady state run, and D) the soil wetness fraction, averaged yearly over 10 years.
Reviewers comment and responses

We would like to thank all reviewers for their insightful comments, which helped us to improve the manuscript. Our revisions reflect all reviewers’ suggestions and comments. For detail, please refer to the responses as follows: reviewer comments in black fonts, responses are in blue fonts.

Reviewer #1

Abstract - This should summarise all the results, so add a sentence about the impact of the root developments as well as the dynamic SOL. Also it would be good to add that initialising with observed SOC gives a better SOC distribution in the simulation than the previous method, just to make this clear.

We implemented suggested rewrites.

Introduction -

Page 3139 Line 5: Please define what you mean by active layer thickness.

We added the following sentences (line 36):

The active layer in permafrost regions is the surficial layer overlying the permafrost, which undergoes seasonal freeze-thaw cycles. Active layer Thickness is the maximum depth of thaw at the end of summer.

Lines 7-8: In fact, Burke et al (2013) does not use a constant carbon density but uses the observations from NCSCD.

We thank the reviewer for pointing it out. We removed Burke et al., (2013) from the list.

Page 3140, Line 5: “Here we describe a fully dynamic SOL to demonstrate the importance of coupling soil biogeochemistry and thermodynamics”: In the analysis later you demonstrate the importance of having a soil organic layer of realistic thickness but you do not demonstrate the importance of making it dynamic. I would not suggest that you change this statement but rather that you should do some more analysis/discussion (see later on).

We expanded the discussion section and stressed the importance of dynamic SOL.

Methods -

Is the SiBCASA model used in any coupled GCM/earth system model? This would be good to mention if so.

SiBCASA is not currently coupled to a GCM. SiBCASA’s predecessor, SiB, has been coupled to the Colorado State University GCM, although it has since been replaced with a different land surface parameterization. We reference the coupling [Sellers et al., 1996], but decided not to delve into this background in the methods section.
Section 2.1 (Frozen carbon initialization)

It is not clear how you initialise the permafrost carbon - is it at the beginning of the 900 year equilibrium run? If so how do you determine the maximum active layer thickness? Or is it at the end of the 900 year equilibrium run, so your equilibrium run is performed with no permafrost carbon present? Or do you do two equilibrium runs: one without permafrost carbon and a second one when you have initialised it based on the active layer in the previous run? Please describe this procedure more clearly.

We added the following clarification to the Methods section.

'It is important to know the “stable” depth of active layer before initializing frozen carbon. We run the model for several years in order to calculate ALT, and then initialized frozen carbon below the maximum calculated ALT. The frozen carbon was initialized only once during the first equilibrium run cycle. For the next equilibrium run we used the previously calculated permafrost carbon. We defined an equilibrium point when changes in overall permafrost carbon were negligible or almost zero.'

Page 3142, line 18: These equations appear unexpectedly with nothing leading up to them to say what they are. I suggest to add a sentence here along the lines of ‘The carbon in each layer is divided into three pools as follows:’ and then give a definition of the pools. We added a sentence as suggested. (L.128)

Section 2.2 (Dynamic SOL)

“SiBCASA already accounted for the effects of organic matter on soil properties like porosity, ...”. Please give more details of this! It is important for understanding the work. For example, did the properties vary depending how ‘compressed’ the organic matter was? Do they vary with depth assuming the organic matter is more compressed at depth? What properties are used and how are they combined with mineral soil properties?

We added the following text (L.149): SiBCASA calculates the soil physical properties as a weighted average of those of organic matter, mineral soil, ice and water (Schaefer et al., 2009). The physical properties include soil porosity, hydraulic conductivity, heat capacity, thermal conductivity, and matric potential. The model calculates the organic fraction used in the weighted mean as the ratio of simulated carbon density to the density of pure organic matter. SiBCASA does not account for the compression of organic matter. Since the prognostic soil carbon pools vary with depth and time, the organic fraction and the physical properties all vary with time and depth. We only summarized these calculations here since the calculations are covered in detail in Schaefer et al. (2009).

It is unclear how the carbon was dealt with in the previous version. The implication is that all carbon entered (and stayed in) the top soil layer and there was none in any deeper layers? Is that true? Please clarify in the text. How did that allow the model to include permafrost carbon (with constant density), as you have mentioned and compared with (Schaefer et al. 2011)?

Each model layer has a complete set of prognostic soil carbon pools. The previous
version of the model distributed fine and coarse root growth vertically within the soil column based on observed root distributions. As the roots die, carbon is transferred to the soil carbon pools for that layer. Thus, the maximum rooting depth determined the maximum depth of ‘new’ or ‘active’ carbon in the model. Of course, if the maximum rooting depth fell below the permafrost table, the model would accumulate permafrost carbon. The current version of the model initializes the permafrost carbon by assigning carbon to the soil carbon pools below the maximum thaw depth. These frozen pools remained inactive until the layer thaws. As live, above-ground biomass in the model dies, carbon is transferred into the first layer as litter. Without the vertical redistribution we describe here to create a surface organic layer, the top layer of the model tended to accumulate carbon in excess of that expected for pure organic matter.

In equation (4) you multiply \( C_{\text{max}} \) by \( f_c \). This seems strange given that there was already a factor of \( f_c \) in the definition of \( C_{\text{max}} \) in equation (3). Is there a mistake here? When you say ‘\( C_{\text{max}} \) is 140 kgm\(^{-3}\)’, perhaps you mean ‘\( \rho_{\text{max}} \)’ here? Please check this.

There is a mistake: \( f_c \) should not appear in equation 4. We corrected the equations accordingly.

What is the purpose of defining \( \text{OLT}_{\text{max}} \)? (Equation 5)

We used \( \text{OLT}_{\text{max}} \) early in the development phase as a diagnostic tool to analyze our results. However, we now see that it does not provide much additional insight and decided to remove it from the paper.

Section 2.3 (Root growth and soil thermal factor)

I am slightly confused as to how your root growth works. Is the root profile prescribed as exponential? If so, what difference does it make if the roots are only growing in the thawed layers? Does this affect the input of carbon to the different soil layers (as well as the autotrophic respiration)? That part was not clear in the text.

We added the following text (L.209): The vertical distribution of new root growth between the soil layers is prescribed using exponential curve fits to observed vertical root distributions. Before the changes we describe here, the maximum rooting depth sometimes exceeded the thaw depth in permafrost soils, resulting in root growth directly in permafrost, which is unrealistic since growing roots cannot penetrate frozen soil. Since the permafrost never thawed, these simulated roots soon died, but never decayed, resulting in an unrealistic buildup of carbon in the upper layers of permafrost. This in turn set up a feedback where the unrealistic increase in organic matter in the simulated permafrost changed the thermodynamic properties and decreased the ALT, resulting in additional carbon buildup. To solve this problem, we kept the original exponential vertical rooting profile, but set maximum rooting depth equal to the thaw depth. This allowed the maximum rooting depth to vary with time and effectively restricted all root growth to within the active layer.
Results -
Page 3147, line 13-14 “so the effect is not as pronounced” - what effect are you talking about here? Not clear.
We clarified it with the following sentence (L.293) “… which allows us to reach a higher heterogeneity between measured and simulated ALTs”

Line 20-21: Slightly confusing to say “a strong peak” when referring to the first plot, since the peak is stronger in the second plot. Probably better to just say “a peak”.
We implemented suggested change.

Line 26: This paragraph is not very clear. I’m not sure what you mean by ‘coupling’. It would be better explained along the lines of… ‘In the version without SOL the ALT was generally deep in forest biomes, but in the new version there is a thick SOL (due to high GPP), which leads to a much shallower ALT. This means there is now a significant amount of root growth in the permafrost itself, which puts carbon directly into the permafrost stores. This is unrealistic, and the frozen soil restrictions on GPP and root growth together eliminate this problem.’ (maybe you could write it better but that is the general idea.)
We changed the confusing paragraph to the following (L.305): In the previous version without a dynamic SOL the ALT was generally deep in forest biomes, but in the new version there is a thick SOL (due to high GPP), which leads to a much shallower ALT. Without restricting root growth within only thawed part of the soil the shallower ALT feedback leads to a significant amount of root growth in the permafrost itself, which puts carbon directly into the permafrost stores. This is unrealistic since growing roots cannot penetrate frozen soil, and the frozen soil restrictions on GPP and root growth together eliminate this problem.

It is misleading to suggest that ‘coupling between GPP and ALT’ does not reflect reality, when in fact there is a real coupling between these quantities. Here it is a negative feedback: Increased GPP $\rightarrow$ increased litter $\rightarrow$ increased SOL $\rightarrow$ reduced thawing $\rightarrow$ reduced GPP. There is also, however, a positive feedback on soil organic carbon as more soil carbon $\rightarrow$ increased SOL $\rightarrow$ reduced temperature $\rightarrow$ reduced respiration $\rightarrow$ more soil carbon. These feedbacks are potentially important and it is great that your model now includes them, but there is no analysis of these feedback processes, which is a shame.
Please cite Koven et al (2009) (which appears in your reference list but not in the manuscript), and include some analysis to demonstrate the coupling in your model. This would significantly improve the worth of the research.
We removed confusing paragraph and added following paragraph to the conclusion:
‘The dynamic SOL and rooting depth strengthens the feedback between GPP and ALT (Koven et al., 2009). Higher GPP produces greater litter fall, which increases the input soil carbon at the surface and results in a thicker SOL. The dynamic SOL changes the properties of the near surface soil, resulting in a shallower ALT and cooler soil temperatures. The dynamic rooting depth accounts for a shallower ALT and modulates
GPP accordingly. The cooler soil temperatures slow microbial decay and increase the carbon accumulation rate, which in turn increases the SOL and reduces ALT further. Eventually, this feedback results in the development of a peat bog. The changes we describe here indicate that SiBCASA can simulate the dynamics of peat bog development, but the model does not yet include a dynamic vegetation model to account for conversions between biome types, such as boreal forest to peat bog.

Page 3148, line 14 onwards. You are discussing the total amount of soil carbon in the simulations. Here it would be good to compare this with the observed total from NCSCD (as well as spatial distribution in Fig. 6).

Prescribing permafrost carbon according to the NCSDC dataset allowed us to better match with the observed pattern in the soil carbon. However, it is does not mean that after the spinup simulated permafrost carbon stocks exactly matched the NCSDC data. During spinup thaw depth varies with time, introducing carbon movement from frozen to thawed pools. In discontinuous zones, if the model simulated permafrost, it tended to produce a deeper active layer depth and thus less permafrost carbon than the NCSDC. The major difference between uniform frozen carbon initialization (Fig 6A, old; Fig 7A, new) and initialization according to the NCSCD (Fig 7B new) is that permafrost exists in more places. However, the NCSCD map (Fig 7C) shows that not all frozen soil contains prescribed by default uniform amount of frozen carbon. Therefore simulating the ‘correct’ ALT is important and should improve the overall permafrost carbon storage.

Page 3149, line 1 (and end of prev. page): “overestimation of the SOC in Central Siberia occurs due to high SOC at the initial time step” - what do you mean by initial timestep? How could that happen? I think if you explained the spinup procedure more clearly in the methods this would be easier to understand.

We rephrase the sentence for better clarity: The overestimation of SOC in Central Siberia is a result of coupling between GPP and ALT.

Page 3149, line 3: How can the ALT be identical when the soil carbon amounts are different? I thought that the physical soil properties depended on the amount of carbon? Again perhaps if you had explained in more detail the changes to soil properties in section 2.2 it would be clearer how this could be the case. (See my comments for ‘Methods’ section, above.)

We included suggested improvements to the method section and added the following text (L.360) for better clarity: This is due to the fact that in both cases soil carbon is added in the permafrost layer below the active layer. Consequently, the amount of soil carbon in the active layer stays does not change between simulations and has the same thermal and carbon dynamics, and thus ALT.

Discussion

In the second paragraph of the discussion you are talking about Figure 7, which shows the spatial patterns of various physical forcings (air temperature, downward longwave
radiation, snow depth, and soil wetness factor). Many claims are made about which factors are contributing to the permafrost distribution and in what way. This analysis is not rigorous and must be improved before the work can be published. For example, “maximum snow depth in South-East Canada is almost half that of West Siberia, which suggests that snow in SE Canada, most likely, is not a major contributor to warm ground temperatures” - well what about the fact that there is a ‘critical’ snow depth below which the soil is much more sensitive to changes in snow depth (see eg. Ekici et al. 2014a): shallower snow does not necessarily imply that it has much less effect. An actual relationship between two variables can only be asserted based on analysis such as a regression or spatial correlations. If you would like to say which factors have the most influence, either comparing the spatial correlations of the forcing variables with ALT, or perhaps performing a multiple regression of all the variables against ALT, would give you much more definite claims. I suggest you replace this paragraph by a more rigorous analysis of the influencing factors.

We added Figure 9 (shown below) that shows the correlations between ALT and every parameter shown in Figure 7. Also, we expanded the discussion for each of the influencing factors. We added and extra discussion on effect of snow on ALT.
Figure 9. The correlation between ALT and: A) near air temperature for averaged over first two month of the fall season, and B) the down-welling long-wave radiation, averaged yearly over 10 years. C) the maximum snow depth over 10 years for the steady state run, and D) the soil wetness fraction, averaged yearly over 10 years.

Similarly, the final paragraph mentions that the CRUNCEP data may not have the zero degree isotherm in the right place. What makes you think that? Why might it not be a problem with the longwave radiation, for example? Could you perhaps find some in-situ air temperature measurements to support that claim?

We agree that paragraph sounded more like speculations. We substituted the paragraph with the following (L.420): Simulated permafrost vulnerability is tightly coupled with the accurate modeling of the present permafrost distribution, which depends on soil thermal properties. We calculate soil thermal properties based on prognostic soil carbon and soil texture from the Harmonized World Soil database (FAO et al., 2009). Observations indicate that soils in the southeast Canada have high soil carbon as a result of a large number of peat lands (Hugelius et al., 2014). Peat has low thermal conductivity and could preserve permafrost even at NSATs about zero degrees centigrade (Jafarov et al., 2012) even if the surrounding areas do not have permafrost. However, the HWSD input data does not have enough soil carbon in the southeast Canada and southwest Russia, as a result, we could not simulate permafrost in those regions.

Conclusions -
Again you claim that the dynamics of the SOL are crucial. In fact, as far as I can tell it is the presence of the (right kind of thickness) SOL that is important. For example, this could be achieved with a static method based on soil carbon observations (such as in Chadburn et al. (2015)). You should show in the paper why your method is better or at least discuss the findings of Koven et al. (2009), which showed the impact of the dynamic coupling between soil carbon and soil properties.

We agree and rewrote the conclusion section.

Lines 9-10. I’m not sure what this sentence means. Particularly, how did the change to plant root growth improve the ALT? And how did it improve the soil carbon? Was this because the carbon was no longer input to the permafrost? I think this needs to be clarified here or better explained in the analysis.

We rewrote the conclusion section.

Line 14-15: “The initialised soil carbon respired during spinup due to abundance of permafrost within the top 3m.” This does not make sense physically. Please check.

We agree and rewrote the conclusion.

Figures -
Figure 7, please define ‘water stress factor’. Also, do you really mean ‘non-dimensionless’? I guess it should just be ‘dimensionless’?

We substituted the water stress factor with more meaningful description soil wetness.
factor. We corrected the typo.

A final suggestion I have which is currently missing from the paper: Some observed data to support the changes in GPP that have resulted from your changes to the model (eg. those shown in Figure 4). Does simulated GPP improve, and if not, why not - or how bad is it?

The major goal was not compare the simulated GPP with measured GPP, but rather to indicate the effect of soil thermal conditions on GPP. Our goal was to illustrate the reduction in GPP when the soil thermal physics is coupled with plant photosynthesis. Preliminary comparisons indicate the SiBCASA simulated GPP may be too high globally, but the cause of this bias is the remotely sensed phenology used as input and not the soil biogeochemistry and thermodynamics. Including such a comparison here would ‘open a can of worms’ and introduce ideas that are completely unrelated to SOL dynamics. This would greatly dilute the emphasis of the paper of SOL dynamics, so we decided not to include it here. We are planning a rigorous comparison under the Multi-scale Synthesis and Terrestrial Model Intercomparison project (MsTMIP, nacp.ornl.gov).

Reviewer #2

p. 3138, l. 13: Better than what? Need to be more precise here about what you are treating as your reference case.

We rewrote that part of the abstract. Now it reads: These improvements allowed us achieve better agreement with the estimated carbon stocks in permafrost-affected soils using historical climate forcing.

p. 3138, l. 19: Hugelius et al., 2014 state the range of C is 1100-1500Pg, with a best estimate of 1300Pg. That paper is an update of the Tarnocai database, so the newer estimate should be used.

We correct our estimates accordingly.

p. 3139, l. 7-12. Not sure the phrase "typically assumed a spatially uniform permafrost carbon density" is an accurate characterization of all of those papers. And also not sure if it is necessary for your argument; the point is that frozen volume matters, irrespective of what the soil carbon content is or how it is calculated.

We corrected the sentence as follows: Consequently, any biases in the simulated ALT could influence the initial amount of frozen carbon, even if different models initialize the frozen carbon in the same way. (L.42)

p. 3139, l. 18. More accurate to say "the same thermal biases that lead to deeper modeled active layers also lead to warmer soil temperatures."

We implemented suggested changes. (L.44)
p. 3141, l. 11. This isn’t the correct reference for the CRUNCEP dataset. Better to put a link to the Viovy website as is typically done.

We have added the recommended reference. We included the original reference as well because we used the CRUNCEP dataset as corrected by the Multi-scale Synthesis and Terrestrial Model Intercomparison project (MsTMIP, nacp.ornl.gov).

p. 3141, l. 20. Is this relevant? Can’t the model read inputs from restart files anyway?

That is right, the model reads its inputs from the restart file. The sentence explains why we choose 900yr for the spinup in opposed to, for example, a 1000yr due to computing constraints. We shortened the sentence accordingly.

p. 3141, l. 22. The citation should be moved to the end of the sentence and the name of the RCN is either the Permafrost Carbon Research Coordination Network, or the Permafrost Carbon Network.

We implement suggested improvements.

p. 3142, l. 19. I think you are missing a sentence here to transition from how you set the total C stocks to how you set the partitioning among pools. Also what are nominal turnover times of these pools at some reference temperature?

We included the following transition sentence: ‘The carbon in each layer is divided into three pools as follows:...’ The nominal turnover times for these pools are 5 years for the slow pool, 76 days for the structural pool, and 20 days for the metabolic pool. We defined the different pools in the text. (L.136)

p. 3143, l. 10. I think the right reference for this is Koven et al., (2009); also note that that model does not include sedimentation processes.

We corrected the corresponding citation. (L.173)

p. 3143, l. 21-23. How do these assumed C densities compare with observations, such as the vertical profiles shown in Harden et al., 2012? It would seem well, given that observed C densities top out at about 60-80 Kg C / m3 for all three permafrost soil types, so maybe useful to mention that as a check on the parameters used here.

We inserted a paragraph and a Figure 5 (see below) showing the simulated vertical carbon distribution. We included the following paragraph into the results section: To illustrate soil carbon distribution with depth we selected three representative areas: a continuous permafrost area corresponding to tundra type biome above the Arctic circle, an area in the boundary of continuous and discontinuous permafrost corresponding to the boreal forest biome, and an area near the south border of the discontinuous permafrost corresponding to poorly vegetated-rocky areas. We calculated mean and standard deviation of the carbon density distribution with depth for 200 grid points around each of the three selected locations. Simulated typical carbon densities from selected locations are shown on Figure 5. All profiles shown on Figure 5 show a similar
pattern: a 20-30 cm SOL with reduced carbon content at the bottom of the active layer. In contrast, the observed vertical carbon profiles show fairly uniform carbon density with depth throughout the active layer and into the permafrost Harden et al., (2012). SiBCASA lacks the cryoturbation processes such as cryotic mixing that would redistribute carbon within the active layer. As a result, the carbon at the bottom of the active layer decayed and respired away during spinup. (L.318)

**Figure 5.** An averaged soil carbon distribution from 200 grid cells A) for the tundra region in continuous permafrost zone, B) for the boreal forest on the boundary between continuous and discontinuous zones, and C) for the low carbon soil at the south border of the discontinuous permafrost zone. The solid blue curve indicates the mean the white blue shading indicate the spread in the soil carbon density.

p. 3144, l. 5. I’m not sure I understand the purpose of OLTmax. In the peatland case, OLT can be several meters.
We used OLTmax early in the development phase as a diagnostic tool to analyze our results. However, we now see that it does not provide much additional insight and decided to remove it from the paper.

p. 3144, l. 10-19. Useful to discuss that functional roles the prognostic roots play in the model behavior. Do they control productivity directly, or are they just used to track C stocks into the ground? Also, what is the basis for linking leaf growth to root growth, as in line 22?

Fine roots supply nutrients and water for photosynthesis, so essentially the leaves and roots together define the photosynthetic capacity of the plant. Plants have optimized carbon allocation to grow only enough fine roots to properly supply the leaves with water and nutrients. So, as plants grow new leaves, they also grow new fine roots to supply them with nutrients and water. Linking root growth to leaf growth is a convenient and simple way to represent this coupling in SiBCASA.

Here we take this coupling one step further and recognize that frozen soil reduces the
plant photosynthetic capacity and regulates root and leaf growth. Plants cannot photosynthesize in frozen soil. Frozen soil in the root zone reduces the photosynthetic capacity of the plant by limiting the water available for photosynthesis. Roots cannot grow in frozen soil and if roots can’t grow, leaves can’t grow. The changes we made link soil thermodynamics to root growth, leaf growth, and plant photosynthesis.

p. 3145, l. 3. Need to explain how $F_{\text{ice}}$ works. Is this just a step function of one below freezing and zero above, or is it more complex?

We added text explaining that $F_{\text{ice}}$ is calculated from the liquid water and ice content of each soil layer, both of which are prognostic variables, accounting for latent heat effects. $F_{\text{ice}}$ varies from zero for a completely thawed soil layer to one for a completely frozen soil layer. (L.233)

p. 3145, l. 6-10. This looks self-contradictory. Either you limit roots to unfrozen layers, or you use an exponential profile, which necessarily continues to have roots (perhaps small, but nonzero) in the frozen layers. Need to explain this better.

We included a better explanation. See earlier comments.

p. 3145, l. 10. If $F_{\text{ice}}$ is not a step function, then there will still be some root growth in partially-frozen soils?

Yes, there will be root growth in a partially frozen soil layer. We added text stating this in the manuscript.

p. 3146, l. 18. Some discussion of how thermal dynamics were calculated in the old version is needed, since that is being used as a reference case. Was soil organic matter included in the thermal calculations at all, or if so, how did it differ from the current version? More generally, is the comparison against the old model the right comparison? Maybe it would be more informative to use just the new version, but turn off various processes to understand their relevance?

We modified the methods and discussion section to reflect the fact that the original version of SiBCASA included the effects of organic matter on soil thermodynamic properties (Schaefer et al., 2009), but not a dynamic SOL and rooting depth. We always strive to build a model flexible enough to turn different processes on and off. We are not software engineers, however, and in this case we found that our code to switch the dynamic SOL and rooting depth processes on and off did not work properly and produced odd results at isolated pixels. We did learn that the dynamic SOL and rooting depth were coupled such that you had to run them both together to get good results. So rather than spend several weeks debugging code that would not affect our results, we decided simply to compare the new and old versions.

I don’t see any information about vertical C profiles in the results, which would seem like a crucial analysis to assess the approach presented here. What does a typical profile
look like? How do overall C profiles compare against datasets such as Harden et al., 2012?

We inserted the paragraph with the corresponding figure on the vertical distribution of soil carbon (see Figure 5).

Figure 4: This should compare predicted GPP to a reference dataset such as that of Beer et al. to assess whether the GPP changes improve the model relative to observations? The major goal was not compare the simulated GPP with measured GPP but rather to indicate the effect of soil thermal conditions on photosynthesis and its feedback to the overall GPP. Our goal was to illustrate the reduction in GPP when the soil thermal physics is coupled with plant photosynthesis. Preliminary comparisons of indicate the SiBCASA simulated GPP may be too high globally, but the cause of this bias is the remotely sensed phenology used as input and not the soil biogeochemistry and thermodynamics. Including such a comparison here would ‘open a can of worms’ and introduce ideas that are completely unrelated to SOL dynamics. This would greatly dilute the emphasis of the paper of SOL dynamics, so we decided not to include it here. We are planning a rigorous comparison under the MsTMIP project.

Figure 5: Also show the final version, as in fig. 6 here, to note the effect of using variable C density.

The old Figure 6 is now Figure 7 includes the result of the initialization with the constant density.

Figure 6, and also discussion points in the text about comparison of low C bias relative to observations in SE Canada and SW Siberia (p. 3148, l. 28, p. 3149, l. 12-27). The reason why these soils have such high C is that they are vast peatland complexes. I think you should clarify the ways in which the model does and does not include peat-like behavior. I.e. the accumulation of organic rich surface layers does seem like peat-like behavior. But only if there are feedbacks between soil saturation, C accumulation, and soil physical properties, as in peats. So the question to pose is: should the model capture the vast peatland complexes in SW Siberia and SE Canada, or not? Getting this right would require both having the right processes in the model, as well as having the right distribution of saturated soils.

We appreciate this comment and include a better explanation of the frozen carbon input and reasoning of why we are not getting permafrost in those regions. We agree that we do not explicitly represent peat-lands in those areas, which might cause the non-existence of permafrost in those locations.

Figure 6. Given that you start with the NCSCD data in the permafrost layers, this isn’t strictly a valid comparison, as there is a clear circularity in comparing input data against reference data. So it would be more appropriate to restrict the comparison to only the active layer C stocks to avoid this. You could sample the NCSCD only to the thaw depth predicted by SiBCASA, and then compare to only the active layer C stocks in SibCAS,
to make such a comparison.

We agree up to a point. The current version of the manuscript includes a better explanation of the frozen carbon input. In our case, it is not fully circular simply because we inserted only frozen carbon instead of whole soil carbon profiles from NCSCD. Moreover, during model equilibration soil carbon within the active layer equilibrates according to the input from the vegetation. The frozen carbon also equilibrates itself based on fluctuations on the thaw depth. Therefore equilibrated carbon differs from NCSCD map. Initially, we did not expect any match between these datasets. The main goal of this comparison was to show that now we can better preserve frozen carbon and therefore better match NCSCD.

p. 3150, l. 29 - p. 3151, l. 6. This is speculation. There are many reasons why a model may overestimate or underestimate permafrost area, e.g. soil processes, snow processes, albedo, etc. So it is not correct to infer that actual permafrost area is being lost just because a model does not simulate permafrost in a given area.

We agree. We change the paragraph with following:

Simulated permafrost vulnerability is tightly coupled with the accurate modeling of the present permafrost distribution, which depends on soil thermal properties. We calculate soil thermal properties based on prognostic soil carbon and soil texture from the Harmonized World Soil Database (HWSD) (FAO et al., 2009). Observations indicate that soils in the southeast Canada have high soil carbon as a result of a large number of peat lands (Hugelius et al., 2014). Peat has low thermal conductivity and could preserve permafrost even at NSATs about zero degrees centigrade (Jafarov et al., 2012) even if the surrounding areas do not have permafrost. However, the HWSD input data does not have enough soil carbon in the southeast Canada and southwest Russia, as a result, we could not simulate permafrost in those regions. (L.419)

Reviewer #3 (S.Yi)

Surface organic layer (SOL) plays an important role in soil thermal dynamics and especially permafrost dynamics. There are several modeling studies, which have already implemented the effects of SOL, in land surface models and ecosystem models. Jafarov and Schaefer tried again to implement SOL in SiBCASA. This work is worth for publish after the following issues on dynamic SOL are addressed.

1. How dynamic SOL is implemented?
The description of SOL dynamics might be too simple. For example, in Pg. 3144 Ln.
We added an extended description of the SOL almost in every section of the manuscript.

2 "the excess organic material was essentially "compressed" into the top soil layer, resulting in a 2 cm simulate SOL".
Does SiBCASA have dynamic soil structure? When top soil layer has excess soil carbon,
2. What will SiBCASA do if disturbances happen?
The authors claimed that SiBCASA performed better than previous version in regions with discontinuous permafrost. These regions have boreal forests and usually have wildfire. Yi et al. (2010) also implemented the processes of buildup of SOL and removal by wildfire in Terrestrial Ecosystem Model; and Yuan et al. (2012) evaluated the role of wildfire in soil thermal dynamics and ecosystem carbon in Yukon River Basin of Alaska. Although the authors mentioned that SiBCASA does not consider disturbance in this version. It is important to provide a prospective for the further development and application of SiBCASA in relating to disturbance since wildfire is common in boreal forest regions. Tundra regions are having more wildfires.

Currently, the Discussion part is too short. I suggest the authors provide more discussion on 1) the differences among different methods of dynamic SOL implementation; 2) the shortcoming of assuming soil carbon transferring downward; and 3) disturbances.

We expanded the discussion section as suggested. The current version of the model does not include fire disturbances. To clarify this we inserted the following paragraph in the discussion section.

Including dynamic organic layer in the model allows us to study the interaction of plant dynamics and soil thermodynamics. In addition it allows us to study other processes in the future, such as fire impacts on soil thermodynamics and recovery from fire, both of which are strongly influenced by the changes in the organic layer (Jafarov et al., 2013). For example, Yuan et al. (2012) evaluated the role of wildfire in soil thermal dynamics and ecosystem carbon in Yukon River Basin of Alaska using Terrestrial Ecosystem Model (Yi et al., 2010), showing wildfires and climate change could substantially alter soil carbon storage. The current version of the model does not include the effect of fire, which means that topsoil carbon stays in the system and provides resilience to permafrost. However, in reality upper SOL could be removed by fire, which would alter soil thermal properties and perturb permafrost carbon stability.