Soil frost-induced soil moisture precipitation feedback over high northern latitudes

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Abstract

Permafrost or perennially frozen ground is an important part of the terrestrial cryosphere; roughly one quarter of Earth’s land surface is underlain by permafrost. The impact of the currently observed warming, which is projected to persist during the coming decades due to anthropogenic CO2 input, certainly has effects for the vast permafrost areas of the high northern latitudes. The quantification of these effects, however, is scientifically still an open question. This is partly due to the complexity of the system, where several feedbacks are interacting between land and atmosphere, sometimes counterbalancing each other. Moreover, until recently, many global circulation models (GCMs) and Earth system models (ESMs) lacked the sufficient representation of cold region physical soil processes in their land surface schemes, especially of the effects of freezing and thawing of soil water for both energy and water cycles. Therefore, it will be analysed in the present study how these processes impact large-scale hydrology and climate over northern hemisphere high latitude land areas. For this analysis, the atmosphere-land part of MPI-ESM, ECHAM6-JSBACh, is driven by prescribed observed SST and sea ice in an AMIP2-type setup with and without newly implemented cold region soil processes. Results show a large improvement in the simulated discharge. On one hand this is related to an improved snowmelt peak of runoff due to frozen soil in spring. On the other hand a subsequent reduction of soil moisture leads to a positive land atmosphere feedback to precipitation over the high latitudes, which reduces the model’s wet biases in precipitation and evapotranspiration during the summer. This is noteworthy as soil moisture – atmosphere feedbacks have previously not been in the research focus over the high latitudes. These results point out the importance of high latitude physical processes at the land surface for the regional climate.

Keywords: Soil moisture – precipitation feedback, soil water freezing, permafrost regions,
global climate modelling, high latitudes

1 Introduction

Roughly one quarter of the northern hemisphere terrestrial land surface is underlain by permafrost (Brown et al., 1997; French, 1990), which is defined as ground that is at or below zero degrees Celsius for more than two consecutive years. Permafrost soils build a globally relevant carbon reservoir as they store large amounts of deep-frozen organic material with high carbon contents. In recent years, estimates for the amount of carbon stored in soils have attracted more and more attention, and here especially the consideration of the vast permafrost regions increased numbers of these estimates drastically (Tarnocai et al., 2009; Zimov et al., 2006; Schuur et al., 2008; McGuire et al., 2009). It is believed to store between 1400 and 1800 Pg of C in the upper few meters of the soil (Schuur et al., 2008), which would be twice the amount of the atmosphere’s content. The high northern latitudes are one of the critical regions of anthropogenic climate change, where the observed warming is clearly above average due to the so-called Arctic Amplification (Solomon et al., 2007; ACIA, 2005). Climate model simulations project this trend to continue (Serreze and Barry, 2011). The combination of the high C stocks in sub-arctic and arctic soils with the pronounced warming in the affected regions could thus lead to a positive feedback through the release of formerly trapped, “deep-frozen’ C into the atmosphere, when near-surface permafrost thaws. For the thawed soils and their biogeochemistry, it is decisive whether dry or wet conditions predominate: Aerobic decomposition is relatively fast and leads to the release of CO2, while anaerobic decomposition is much slower and leads to the release of CH4 as the main product of the combustion of organic soil material. CH4 is a much more potent greenhouse gas, but has a shorter lifetime of about 10 years after which it becomes CO2. Therefore, not only the soil’s temperature, but also its moisture status are important for the assessment of the
biogeochemical response to climatic conditions, and thus should be represented in climate or
Earth System models in a realistic and process-based manner. Thus, the adequate
representation of permafrost hydrology is a necessary and challenging task in climate
modelling.

Hagemann et al. (2013a) described relevant hydrological processes that occur in permafrost
areas and that should preferably be represented in models simulating interactions of
permafrost hydrology with vegetation, climate and the carbon cycle. The current state of the
representation of processes in general circulation models (GCMs) or Earth system models
(ESMs) can be obtained by systematic model intercomparison through the various climate
model intercomparison projects (CMIPs; Meehl et al., 2000) that have a long history within
the climate modelling community. Results from CMIPs provide a good overview on the
respective state of ESM model accuracy and performance. Koven et al. (2012) analysed the
performance of ESMs from the most recent CMIP5 exercise over permafrost areas. They
found that the CMIP5 models have a wide range of behaviours under the current climate, with
many failing to agree with fundamental aspects of the observed soil thermal regime at high
latitudes. This large variety of results originates from a substantial range in the level of
complexity and advancement of permafrost-related processes implemented in the CMIP5
models (see, e.g., Hagemann et al., 2013a), whereat most of these models do not include
permafrost specific processes, not even the most basic process of freezing and melting of soil
water. Due to missing processes and related deficiencies of their land surface schemes,
climate models often show substantial biases in hydrological variables over high northern
latitudes (Luo et al., 2003; Swenson et al., 2012). Moreover, the land surface
parameterizations used in GCMs usually do not adequately resolve the soil conditions (Walsh
et al., 2005), which often rely on either point measurements or on information derived from
satellite data. Therefore, large efforts are ongoing to extend ESMs in this respect, in order to
improve simulated soil moisture profiles and associated ice contents, river discharge, surface and sub-surface runoff. The ESM improvement over permafrost areas was, e.g., one of the research objectives of the European Union Project PAGE21 (www.page21.org).

The most basic process in permafrost areas is the seasonal melting and freezing of soil water in the presence of continuously frozen ground below a certain depth. The response of the soil to freezing leads to specific variations in the annual cycle of soil hydrology. Frozen ground and snow cover also influence rainfall-runoff partitioning, the timing and magnitude of spring runoff, and the amount of soil moisture that subsequently is available for evapotranspiration in spring and summer (Koren et al., 1999). Soil moisture controls the partitioning of the available energy into latent and sensible heat flux and conditions the amount of surface runoff. By controlling evapotranspiration, it is linking the energy, water and carbon fluxes (Koster et al., 2004; Dirmeyer et al., 2006; Seneviratne and Stöckli, 2008). Seneviratne et al. (2006) stated that a northward shift of climatic regimes in Europe due to climate change will result in a new transitional climate zone between dry and wet climates with strong land–atmosphere coupling in central and eastern Europe. They specifically highlight the importance of soil-moisture–temperature feedbacks (in addition to soil-moisture–precipitation feedbacks) for future climate changes over this region. A comprehensive review on soil moisture feedbacks is given by Seneviratne et al. (2010).

Largely, soil moisture feedbacks to the atmosphere are confined to regions where the evapotranspiration is moisture-limited. These are regions where the soil moisture is in the transitional regime between the permanent wilting point (soil moisture content below which the plants can not extract water from the soil by transpiration as the suction forces of the soil are larger than the transpiration forces of the plants) and the critical soil moisture $W_{crit}$ above which plants transpire at the potential rate (see, e.g., Fig. 5 in Seneviratne et al., 2010). In this
respect, the high-latitudes are usually excluded those regions as they are considered to be predominantly energy-limited (Teuling et al., 2009), and where the coupling between soil moisture and the atmosphere does not play a role (Koster et al., 2004, 2006).

Note that in previous studies where an ESM’s land surface model (LSM) was equipped with cold region soil processes, effects of resulting model improvements usually have not been directly considered in a coupled atmosphere-land context. Either simulated changes were only considered in the LSM standalone mode (e.g. Ekici et al., 2014, 2015; Lawrence and Slater, 2005; Gouttevin et al., 2012; Slater et al., 1998), or changes between different LSM version were not limited to cold region processes alone (Cox et al., 1999). Thus, any soil moisture feedbacks to the atmosphere related to cold region soil processes have been neglected so far.

In the present study, we show that the implementation of cold region soil processes into the ESM of the Max Planck Institute for Meteorology, MPI-ESM, has a pronounced impact on the simulated terrestrial climate over the northern high latitudes, and that this is mainly related to a positive soil moisture-precipitation feedback. Section 2 introduces the used ESM version and the setup of the associated simulations, Section 3 discusses the main results over several high latitude river catchments, followed by a summary and conclusions in Section 4.

2 Model, data and methods

2.1 Model description

In this study, the atmosphere and land components of the Earth System Model (ESM) of the Max Planck Institute for Meteorology (MPI-M), MPI-ESM 1.1, are utilized that consist of the atmospheric GCM ECHAM6.3 (Stevens et al., 2013) and its land surface scheme JSBACH 3.0 (Raddatz et al., 2007, Brovkin et al., 2009). Both models have undergone several further
developments since the version (ECHAM6.1/JSBACH 2.0) used for the Coupled Model Intercomparison Project 5 (CMIP5; Taylor et al., 2012). Several bug fixes in the ECHAM physical parameterizations led to energy conservation in the total parameterized physics and a re-calibration of the cloud processes resulted in a medium range climate sensitivity of about 3 K. JSBACH 3.0 comprises several bug fixes, a new soil carbon model (Goll et al., 2015) and a five layer soil hydrology scheme (Hagemann and Stacke, 2015) replaced the previous bucket scheme. In addition, a permafrost-ready version of JSBACH is considered (JSBACH-PF) in which physical processes relevant at high latitude land regions have been implemented by Ekici et al. (2014). Most importantly, these processes comprise the freezing and melting of soil moisture. Consequently, the latent heat of fusion dampens the amplitude of soil temperature, infiltration is decreased when the uppermost soil layer is frozen, soil moisture is bound in solid phase when frozen, and, hence, cannot be transported vertically or horizontally. Dynamic soil thermal properties now depend on soil texture as well as on soil water and ice contents. Dynamic soil hydraulic properties that depend on soil texture and soil water content are decreased when soil moisture is frozen. Moreover a snow scheme has been implemented in which snow can now develop in up to five layers while the current scheme only represents up to two layers. The latter also thermally lets the snow grow inside the soil (i.e. soil temperatures are mixed with snow temperatures), while the new scheme accumulates the snow on top of the soil using snow thermal properties. Further, a homogeneous organic top layer is added with a constant depth and specific thermal and hydraulic properties.

### 2.2 Experimental setup

Two ECHAM6.3/JSBACH simulations were conducted at T63 horizontal resolution (about 200 km) with 47 vertical layers in the atmosphere. They were forced by observed sea surface temperature (SST) and sea ice from the AMIP2 (Atmospheric Model Intercomparison Project
2) dataset for 1970-2009 (Taylor et al., 2000). 1970-1988 are regarded as spin-up phase so that only the period 1989-2009 is considered for the analyses. The two simulations are:

- **ECH6-REF**: Simulation with the standard version of JSBACH 3.0 with a fixed vegetation distribution and using a separate upper layer reservoir for bare soil evaporation as described in Hagemann and Stacke (2015). Note that the latter is switched off by default in JSBACH 3.0 to achieve a better performance of simulated primary productivity, which is not of interest in the present study.

- **ECH6-PF**: As ECH6-REF, but using JSBACH-PF.

Note that both simulations used initial values of soil moisture, soil temperature and snowpack that were obtained from an offline-simulation (land only) using JSBACH (as in ECH6-REF) forced with WFDEI data (Weedon et al., 2014).

### 2.3 Calculation of internal model climate variability

The internal climate variability of ECHAM6/JSBACH with respect to 20-year mean values has been estimated from results of three 20-year, 5-member ensembles, in which the ensembles used different land-atmosphere coupling setups (deVrese et al., 2016). Within each ensemble, the model setup is identical but the simulations were started using slightly differing initial conditions. Following the approach of Hagemann et al. (2009), we first calculated the standard deviation of 20-year means for each ensemble, and then the spread for each model grid box is defined as the maximum of the three ensemble standard deviations. This spread is then used as an estimate of the model’s internal climate variability. Thus, if simulated differences between ECH6-PF and ECH6-REF are larger than this spread, they are considered as robust and directly related to the introduction of cold region soil processes into JSBACH.
2.4 Observational data

We use climatological observed river discharges from the station network of the Global Runoff Data Centre (Dümenil Gates et al., 2000). Near surface air (2m) temperature and precipitation are taken from the global WATCH dataset of hydrological forcing data (WFD; Weedon et al., 2014). The WFDEI combine the daily statistics of the Interim re-analysis of the European Centre for Medium-Range Weather Forecasts (ERA-Interim; Dee et al., 2011) with the monthly mean observed characteristics of temperature from the Climate Research Unit dataset TS2.1 (CRU; Mitchell and Jones, 2005) and precipitation from the Global Precipitation Climatology Centre full dataset version 4 (GPCC; Fuchs et al., 2007). For the latter, a gauge-undercatch correction following Adam and Lettenmaier (2003) was used, which takes into account the systematic underestimation of precipitation measurements that have an error of up to 10-50% (see, e.g. Rudolf and Rubel, 2005).

For an estimate of observed evapotranspiration (ET), we are using data from the LandFlux-EVAL dataset. This new product was generated to compile multi-year global merged benchmark synthesis products based on the analyses of existing land evapotranspiration datasets (monthly time scale, time periods 1989-1995 and 1989-2005). The calculation and analyses of the products are described in Mueller et al. (2013). In our study we are using the diagnostic products available for the period 1989-2005 that are based on various observations, i.e. from remote sensing, diagnostic estimates (atmospheric water-balance estimates) and ground observations (flux measurements). Here, we considered the mean, minimum and maximum of the respective diagnostic ensemble.

Surface solar irradiance (SSI; 2000-2010) is taken from the Clouds and Earth Radiation Energy System (CERES; Kato et al., 2013) that provides surface solar radiation fluxes at global scale derived from measurements onboard of the EOS Terra and Aqua satellites (Loeb...
et al., 2012). We used surface albedo data from MODIS (MCD43C3, ver5; 2000-2011; Cescatti et al., 2012), CERES (2000-2010) and the GlobAlbedo project (1998-2011; Muller et al., 2012) of the European Space Agency (ESA). With regard to the accumulated snowpack, we compared model data to snow water equivalent data from the ESA GlobSnow project (Takala et al., 2011), NASA’s Modern-Era Retrospective Analysis for Research and Applications (MERRA; 1979-2013; Rienecker et al., 2011) and the snow data climatology (SDC) of Foster and Davy (1988).

2.5 Permafrost extent

Observational datasets of permafrost extent usually give three or four classes of spatial permafrost occurrence, where the respective percentage of permafrost covered area is > 90 % (‘continuous’), between 90 and 50 % (‘discontinuous’), < 50 % (‘sporadic’), and, in some references, < 10 % (‘isolated’). This is the case in the data of Brown et al. (1997) shown here in Fig. 1a. In most climate models, such a diversification of permafrost classes is not possible. In those models as well as in JSBACH, soil temperatures are computed for one point at the centre of a grid cell, thereby representing the whole area of that cell. Consequently, no ‘non-continuous’ permafrost can be computed by JSBACH. Thus, the comparison of simulated with observed permafrost extents focuses on the continuous class in the observations.

In order to diagnose permafrost extent from JSBACH output, its fifth layer soil temperature has been extracted and checked whether it has been lower than 0 °C for more than two years in a row. This criterion was applied to a 30 year time series of monthly means (1979-2009), and during every proceeding month, the sum of ‘permafrost months’ have been set into relationship to the total number of months in the time series analysed so far. This enables us to have temporal variation, and avoid ‘loosing’ permafrost areas where it simply did not occur during the last two years of the analysed time series.
3 Results

Initially, the simulated permafrost extents are compared with the data of Brown et al. (1997) in Fig. 1. The implementation of permafrost relevant soil processes into JSBACH leads to an improved permafrost representation in terms of continuous permafrost extent, as the too large extent in western Siberia as well as in Alaska decreases in ECH-PF. Reasons for this improvement are presumably the changed snow scheme and the separation of snow and soil temperatures on the one hand, and the new formulation of the soil thermal properties on the other hand. Combined with the organic top layer, they change the conditions for heat transfer into and within the ground, which leads to more realistic deep soil temperatures in the above mentioned regions.

Then, both simulations are evaluated over the northern high latitudes analogously to the evaluation of surface water and energy fluxes of the CMIP5 version of MPI-ESM by Hagemann et al. (2013b). The main differences in precipitation and 2m temperature between both simulations occur in the boreal summer. In ECH6-PF, precipitation is generally reduced compared to ECH-REF over the northern high latitudes (Fig. 2). On the one hand, this leads to a general reduction of the wet bias compared to WFDEI data over the more continental areas north of about 60°N, especially over Canada and Russia. On the other hand, it enhances the dry bias over the adjacent mid-latitudes. Note that this summer dry bias of MPI-ESM 1.1 over mid-latitudes is more pronounced and wide-spread than in the CMIP5 version of MPI-ESM (cf. Fig. 4, middle row, in Hagemann et al., 2013b), which is likely associated with bug-fixes or the re-calibration of cloud processes in ECHAM6.3 (cf. Sect. 2.1). The same is also the case for northern hemisphere summer warm biases in ECH6-REF (Fig. 3). These warm biases are enhanced in ECH6-PF. This enhancement is partly related to the fact that the reduced precipitation is accompanied by a reduced cloud cover, and, hence an increased...
incoming solar radiation at the land surface (Fig. 4). Compared to CERES data, the low bias in SSI over the high latitudes is largely removed while the overestimation over the mid-latitudes is slightly increased. The reason for the warmer air temperatures can partly be found in a decreased evapotranspiration (ET) when permafrost relevant physical soil processes are switched on. A detailed analysis of their effects was carried out to elucidate the specific influence of these processes and is shown for two large example catchments (Fig. 5). 1) The Arctic catchment is represented by the six largest rivers flowing into the Arctic Ocean: Kolyma, Lena, Mackenzie, Northern Dvina, Ob and Yenisei. The associated catchments comprise a large fraction of permafrost covered areas (cf. Fig. 1). 2) The Baltic Sea catchment includes only a low amount of permafrost covered areas but soil moisture freezing still plays a role over large parts of the catchment during the winter.

Arctic River catchments

ECH6-PF simulates the discharge of the six largest Arctic rivers more reliably than ECH6-REF, especially with regard to timing and size of the snow melt induced discharge peak in spring (Fig. 6a). This is largely related to the fact that in ECH6-PF, a major part of the snow melt turns into surface runoff as it cannot infiltrate into the ground when this is still frozen in the beginning of spring. This is opposite to ECH6-REF where larger parts of the snow melt are infiltrating into the soil due to the missing freezing processes such that the observed discharge peak is largely underestimated.

Also with regard to precipitation, ECH6-PF shows a large improvement in the simulated summer precipitation as the large wet bias of ECH6-REF is strongly reduced and, hence, much closer to WFDEI data (Fig. 6c). This reduction in summer precipitation is accompanied by a reduction in summer evapotranspiration (Fig. 7a) that is now much closer to the mean of diagnostic estimates from the LandFlux dataset, while it is likely overestimated in ECH6-REF.
as the simulated evapotranspiration is close to the upper limit of the LandFlux diagnostic estimates. This ET reduction in ECH6-PF is directly related to a completely changed seasonal cycle of liquid relative soil moisture (actual soil moisture divided by the maximum soil water holding capacity) in the root zone (Fig. 7c). In ECH6-REF, the soil is very wet throughout the whole year with somewhat lower values in summer that are related to the summer ET. In ECH6-PF, the soil is rather dry in winter as larger parts of the soil moisture are frozen (Fig. 8), and, hence, not accessible for ET. With infiltration of snowmelt in the spring when the soil water of the upper layer has melted, the soil moisture is increasing and reaches its maximum in summer. The total amount of liquid soil moisture in ECH6-PF is much lower than in ECH6-REF. On the one hand large parts of the soil are frozen in winter and adjacent months (Fig. 8), and on the other hand this is related to the much lower infiltration in spring, so that less moisture is available throughout the whole year. In the autumn and winter, the total amount of soil water is somewhat increasing (Fig. 7c) as due to freezing, it is locally bound and can neither flow off laterally nor evaporate. If compared to the model’s internal climate variability (Fig. 9) we note that the differences between ECH6-PF and ECH6-REF are robust for ET and precipitation from April-October and April-August, respectively.

The decreased ET during warm months, however, brings about less evaporative cooling of the land surface, and near surface air temperature increases with the use of the PF scheme. This results in a further increase of the warm bias in 2m air temperature in comparison to WFDEI data (Fig. 10a). Parts of the summer warm bias is caused by an overestimated incoming surface solar irradiance (SSI). In ECH6-REF, the simulated SSI is close to CERES data (Fig. 10c), but in ECH6-PF the reduced ET leads to a reduced upward moisture flux into the atmosphere that in turn seems to reduce cloud cover, and, hence SSI is increased.

The surface albedo is rather similar in both experiments (Fig. 11a) but shows some distinct
biases if compared to various observational datasets. During the winter JSBACH seems to overestimate the mainly snow-related albedo, indicating that it may have difficulties to adequately represent snow-masking effect of boreal forests (Note that a version of MODIS albedo data was used where low quality data over the very high northern latitudes were filtered out in the boreal winter due to too low available radiation (A. Löw, pers. comm., 2016). Due to these missing data over mainly snow covered areas, MODIS albedo averaged over the six largest Arctic rivers is biased low in the winter). During the summer, there is a larger uncertainty in the observations. While the simulated albedo is close to MODIS and CERES data, it is lower than GlobAlbedo data. As a too low albedo would lead to a warm bias, this might indicate a better reliability of the GlobAlbedo data for this region in summer. Note that a sensitivity test where surface albedo was increased by 0.05 north of 60°N led to a reduction of the warm bias by about 1-2 K (not shown). As already indicated by the surface albedo, the simulated snow cover does not significantly differ between the experiments, either (Fig. 11c). It is lower than various observational estimates, which should impose a low albedo bias in winter. As this bias is in the opposite direction, it can be concluded that the low snow pack is compensating part of the snow masking problem mentioned above.

**Baltic Sea catchment**

A similar effect of the frozen ground is found over the Baltic Sea catchment, although this is less strong than for the Arctic rivers. The frozen ground leads to an enhanced snow melt runoff in spring (Fig. 6b) and a less strong replenishment of the ground by water during the winter as it is the case for ECH6-REF (Fig. 7d). Consequently the average level of liquid soil moisture is lower in ECH6-PF compared to ECH6-REF. This leads to more infiltration of water and less drainage, and hence, less runoff in the summer, which in turns leads to an improved simulation of discharge (Fig. 6b). The impact on the atmosphere is much less
pronounced than for the Arctic rivers. On one hand there is less frozen ground in the Baltic
Sea catchment (Fig. 8), on the other hand the average soil moisture content is larger than for
the Arctic rivers (Fig. 7d). In ECH6-REF, the soil moisture is generally above $W_{crit}$ (c.f. Sect. 1)
over the Baltic Sea catchment so that ET is largely energy limited and mostly occurring at
its potential rate. Even though the ECH6-PF soil moisture is lower, it is generally still close to
$W_{crit}$ so that ET is only slightly reduced, especially in the second half of the year (Fig. 7b).
Precipitation is also somewhat reduced (Fig. 6d) but this seems to be mostly related to the
internal climate variability except for September and October when a somewhat stronger and
robust reduction in ET leads to a robust precipitation decrease (Fig. 9).

4 Discussion and conclusions

The results described in the previous section show that the introduction of cold region
processes into MPI-ESM led to a positive soil moisture-precipitation feedback over large
parts of northern mid- and high latitudes during the boreal summer. The chain of processes
leading to this feedback is sketched in Fig. 12. The frozen soil during the cold season (late
autumn to early spring) leads to less infiltration of rainfall and snowmelt during this season,
and, hence, to more surface runoff especially during the snowmelt period. On one hand this
leads to a large improvement in simulated discharge, mainly due to the improved snowmelt
peak. This improved discharge due to the representation of frozen ground has been also
reported for other models (Beer et al., 2006, 2007; Ekici et al., 2014; Gouttevin et al., 2012).
On the other hand, this leads to a decrease of soil moisture. During the boreal summer, this
actually causes more infiltration and less runoff, and, hence, less discharge. The latter strongly
improves the simulated discharge in the Baltic Sea catchment from summer to early winter.
The decreased soil moisture leads to a reduced ET in regions where the soil moisture is in the
transitional regime. Here, there is less recycling of moisture into the atmosphere, and the
lower atmospheric moisture causes a reduction of precipitation that in turn leads to a further
reduction of soil moisture. This positive soil moisture-precipitation feedback improves the
simulated hydrological cycle, especially over the Arctic rivers where the wet biases in
summer precipitation and ET are reduced. Less ET, and, hence, less evaporative cooling cause
an increase in summer 2m air temperatures. This, in combination with more incoming surface
solar radiation due to fewer clouds, increases and extends the existing summer warm bias of
MPI-ESM north of about 50°N.

Such a positive soil moisture-precipitation feedback has not been pointed out for the northern
high latitudes so far, which previously have generally been considered as energy-limited
regimes where land-atmosphere coupling due to soil moisture does not play a role (e.g.
Teuling et al., 2009). But this principal feedback loop has been found for drier regions where
the soil moisture is generally in the transitional regime and land-atmosphere coupling plays a
role. Koster et al. (2004) considered the strength of coupling between soil moisture and
precipitation in an ensemble of atmospheric GCMs. The resulting map is very similar to the
map regarding the strength of coupling between soil moisture and temperature in the same
GCMs (Koster et al., 2006). This suggests that in these models, the same process controls
both couplings, namely the ET sensitivity to soil moisture that leads to a positive feedback
(Seneviratne et al., 2010). But in those studies (Koster et al., 2004; Teuling et al., 2009),
usually annual mean diagnostics were considered. Our study has shown that seasonally, i.e.
during the boreal summer, soil moisture conditions may prevail that allow for land-
atmosphere coupling and a positive soil moisture-precipitation feedback over the northern
high and mid-latitudes.

Even though our results are obtained with a modelling study, their physical consistency
suggests that cold region soil processes, especially melting and freezing of soil moisture, may
lead to a positive soil moisture precipitation feedback during the summer in reality, too. A prerequisite for the occurrence of a soil moisture precipitation feedback is that soil moisture is in the transitional regime. Thus, the strength of the feedback depends on the wetness of the soil and, hence, is likely model dependent. Models with wetter/drier soils over the considered regions may simulate a weaker/stronger feedback.

Several modelling studies pointed out that there are not only positive feedback loops between soil moisture and precipitation but also negative ones that, under specific conditions, such as convective instability and/or cloud formation, may be stronger over dry soils (e.g. Hohenegger et al., 2009; Froidevaux et al., 2014). However, to date, the latter results appear mostly confined to single-column, cloud-resolving, and some high-resolution regional climate simulations (Seneviratne et al., 2010) and may also depend on the choice of the convective parameterisations (e.g. Giorgi et al., 1996). Guillod et al. (2015) noted that precipitation events tend to be located over drier patches, but they generally need to be surrounded by wet conditions; positive temporal soil moisture-precipitation relationships are thus driven by large-scale soil moisture. Thus, negative feedbacks seem to have more an impact on high resolution and thus on the local scale (Ho-Hagemann et al., 2015), where the effects of land surface heterogeneity for the inferred feedbacks also need to be taken into account (Chen and Avissar, 1994; Pielke et al., 1998; Taylor et al., 2013). Consequently most GCMs may not be able to represent negative feedbacks between soil moisture and precipitation via ET. As in the present study, we considered the effect of large-scale soil moisture changes due to soil freezing processes, the identification of potential negative feedbacks on the local scale is not an issue.

In MPI-ESM, an unwelcome effect of implementing cold region soil processes is the increase of the existing warm bias over the high latitudes during summer. In order to estimate the
contribution of biases in SSI and surface albedo to this warm bias, we calculated an upper limit for the temperature change that may be imposed by a radiation difference in the related energy flux into the ground \([\text{SSI} \times (1 - \text{albedo})]\). For this estimation we assume that the surface temperature is adjusting in a way that this radiation difference is compensated by thermal radiation following the Stefan Boltzmann law. Here, any change in the turbulent surface heat fluxes is neglected so that the resulting temperature change is an upper limit for the temperature bias that might be explained by a radiation bias.

Considering the mean summer biases over the six largest Arctic rivers (Table 1) indicates that a part of the warm bias may be attributed to the overestimation in SSI. For ECH6-PF (ECH6-REF), the SSI bias may cause a warm bias of up to 2.9 K (0.9 K). The surface albedo may contribute another 0.7 K (0.8 K) to the warm bias if compared to GlobAlbedo data but this is a rather vague estimation due to the large uncertainty on surface albedo observations (see Fig. 11). Nevertheless biases in both of these variables cannot explain the full bias of 5 K (2.1 K) in 2m temperature. Further contributions to this warm bias might be related to too much advection of warm air or a too weak vertical mixing of heat within the boundary layer. A deeper investigation of this is beyond the scope of the present study and should be dealt with in future model improvements.

We have shown that biophysical land surface processes such as melting and freezing can have a significant impact on the regional climate over the high latitudes and permafrost areas. Flato et al. (2013) reported that CMIP5 GCMs tend to overestimate precipitation over northern high latitudes except for Europe and western Siberia. As many of these GCMs are still missing basic cold region processes (see Sect. 1), a missing soil moisture precipitation feedback in those GCMs might contribute to this wet bias. Beyond the biophysical coupling between land and atmosphere, the coupling to biogeochemistry, i.e. vegetation and carbon cycle including
methane and frozen carbon, is important to quantify feedbacks related to wetlands and permafrost over those areas. The representation of their complex dynamics within ESMs is a challenging task, but it is nevertheless necessary to investigate on-going and future climate changes over the high-latitude regions. Thus, the adequate implementation of physical soil processes into an ESM is only the first necessary step to yield an adequate representation of climate feedbacks over the high latitudes. This also includes the incorporation of wetland dynamics, which will be the next step in the JSBACH development with regard to high latitudes, thereby following an approach of Stacke and Hagemann (2012).

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Figure captions

Fig. 1 Distribution of permafrost areas in the Arctic according to a) Brown et al. (1997), b) ECH6-REF, and c) ECH6-PF.

Fig. 2 Boreal summer (JJA) precipitation differences [%] relative to WFDEI data for a) ECH6-REF, and b) ECH6-PF.

Fig. 3 Boreal summer (JJA) 2m temperature differences [K] to WFDEI data for a) ECH6-REF, and b) ECH6-PF.

Fig. 4 Boreal summer (JJA) surface solar incoming radiation differences [W/m²] to CERES data for a) ECH6-REF, and b) ECH6-PF.

Fig. 5 Catchments of the Baltic Sea and of the six largest Arctic rivers (from left to right: Mackenzie, Baltic Sea, Northern Dvina, Ob, Yenisei, Lena, Kolyma).

Fig. 6 Mean monthly climatology (1989-2009) of discharge (upper panels) and precipitation (lower panels) over the 6 largest Arctic river catchments (left column) and the Baltic Sea catchment (land only, right column). Observations comprise climatological observed discharge and WFDEI precipitation, respectively.

Fig. 7 Mean monthly climatology (1989-2009) of evapotranspiration (upper panels) and relative root zone soil moisture (lower panels) over the 6 largest Arctic river catchments (left column) and the Baltic Sea catchment (land only, right column). Evapotranspiration data comprise the mean, minimum and maximum diagnostic estimates from the LandFlux Eval (LF) dataset. The dashed blue line denotes the total root zone water content (liquid + frozen) for ECH6-PF.

Fig. 8 Mean fraction of frozen root zone soil moisture (1989-2009) over the 6 largest Arctic river catchments (solid curve) and the Baltic Sea catchment (land only, dashed curve).

Fig. 9 Mean monthly climatological differences (1989-2009) of between ECH6-PF and
ECH6-REF for precipitation ($\Delta P$) and evapotranspiration ($\Delta ET$) over the 6 largest Arctic rivers (upper panel) and the Baltic Sea catchment (lower panel). The dashed lines indicate the corresponding spreads obtained from MPI-ESM simulations of deVrese et al. (2016).

Fig. 10 Mean monthly climatology (1989-2009) of 2m temperature differences to WFDEI data (upper panels) and surface solar irradiance (SSI; lower panels) over the 6 largest Arctic river catchments (left column) and the Baltic Sea catchment (land only, right column). SSI observations comprise CERES data for 2000-2010.

Fig. 11 Mean monthly climatology (1989-2009) of surface albedo (upper panels) and snow pack snow water equivalent (SWE; lower panels) over the 6 largest Arctic river catchments (left column) and the Baltic Sea catchment (land only, right column). Albedo observations data from MODIS (2000-2011), CERES (2000-2010) and GlobAlbedo (1998-2011), SWE observations comprise data from GlobSnow (1989-2009), MERRA (1979-2013), and SDC climatology.

Fig. 12 Chain of processes involved in the soil moisture precipitation feedback over high latitudes. Red arrows indicate directions supporting this feedback, blue arrows indicate compensating opposite effects.
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Table 1. Summer (JJA) biases over the six largest Arctic rivers for 2m temperature ($T_{2m}$, to WFDEI), radiative flux ($R$) into the surface due to biases in SSI (to CERES), albedo ($\alpha$, to GlobAlbedo) and their combined effect (comb.) as well as the estimated related impact on surface temperature ($T_s$) and the contribution of the SSI bias to this impact.

<table>
<thead>
<tr>
<th>Experiment</th>
<th>$\Delta T_{2m}$</th>
<th>$\Delta R_{SSI}$</th>
<th>$\Delta R_{\alpha}$</th>
<th>$\Delta R_{comb.}$</th>
<th>$\Delta T_s$ comb.</th>
<th>SSI cont.</th>
</tr>
</thead>
<tbody>
<tr>
<td>ECH6-REF</td>
<td>2.1 K</td>
<td>5.0 W/m²</td>
<td>4.1 W/m²</td>
<td>9.0 W/m²</td>
<td>1.7 K</td>
<td>55%</td>
</tr>
<tr>
<td>ECH6-PF</td>
<td>5.0 K</td>
<td>15.8 W/m²</td>
<td>4.3 W/m²</td>
<td>19.8 W/m²</td>
<td>3.6 K</td>
<td>78%</td>
</tr>
</tbody>
</table>
