The incorporation of an organic soil layer in the Noah-MP land surface model and its evaluation over a boreal aspen forest

Liang Chen1,2, Yanping Li1, Fei Chen3, Alan Barr4, Michael Barlage3, and Bingcheng Wan3

1Global Institute for Water Security, University of Saskatchewan, Saskatoon, SK, Canada
2Key Laboratory of Regional Climate Environment for Temperate East Asia, Institute of Atmospheric Physics, Chinese Academy of Sciences, Beijing, China
3National Center for Atmospheric Research, Boulder, Colorado, USA
4Environment Canada, National Hydrology Research Center, Saskatoon, SK, Canada

Correspondence to: Yanping Li (yanping.li@usask.ca)

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Abstract. A thick top layer of organic matter is a dominant feature in boreal forests and can impact land–atmosphere interactions. In this study, the multi-parameterization version of the Noah land surface model (Noah-MP) was used to investigate the impact of incorporating a forest-floor organic soil layer on the simulated surface energy and water cycle components at the BERMS Old Aspen site (OAS) field station in central Saskatchewan, Canada. Compared to a simulation without an organic soil parameterization (CTL), the Noah-MP simulation with an organic soil (OGN) improved Noah-MP-simulated soil temperature profiles and soil moisture at 40–100 cm, especially the phase and amplitude (Seasonal cycle) of soil temperature below 10 cm. OGN also enhanced the simulation of sensible and latent heat fluxes in spring, especially in wet years, which is mostly related to the timing of spring soil thaw and warming. Simulated top-layer soil moisture is better in OGN than that in CTL. The effects of including an organic soil layer on soil temperature are not uniform throughout the soil depth and are more prominent in summer. For drought years, the OGN simulation substantially modified the partitioning of water between direct soil evaporation and vegetation transpiration. For wet years, the OGN-simulated latent heat fluxes are similar to CTL except for the spring season when OGN produced less evaporation, which was closer to observations. Including organic soil produced more subsurface runoff and resulted in much higher runoff throughout the freezing periods in wet years.

1 Introduction

Land surface processes play an important role in the climate system by controlling land–atmosphere exchanges of momentum, energy, and mass (water, carbon dioxide, and aerosols). Therefore, it is critical to correctly represent these processes in land surface models (LSMs) that are used in weather prediction and climate models (e.g., Dickinson et al., 1986; Sellers et al., 1996; Chen and Dudhia, 2001; Dai et al., 2003; Oleson et al., 2008; Niu et al., 2011). Niu et al. (2011) and Yang et al. (2011) developed the Noah LSM with multi-parameterization options (Noah-MP) and evaluated its simulated seasonal and annual cycles of snow, hydrology, and vegetation in different regions. Noah-MP has been implemented in the community Weather Research and Forecasting (WRF) model (Barlage et al., 2015), which is widely used as a numerical weather prediction and regional climate model for dynamical downscaling in many regions worldwide (Chotamonsak et al., 2012). The performance of Noah-MP was previously evaluated using in situ and satellite data (Niu et al., 2011; Yang et al., 2011; Cai et al., 2014; Pilotto et al., 2015; Chen et al., 2014). Those evaluation results showed significant improvements in modeling runoff, snow, surface heat fluxes, soil moisture, and surface skin temperature compared to the Noah LSM (Chen et al., 1996; Ek et al., 2003). Recently, Chen et al. (2014) compared Noah-MP to Noah and four other LSMs regarding the simulation of snow and surface heat fluxes at a forested site in the Colorado headwaters region, and found a generally good performance...
of Noah-MP. However, it is challenging to parameterize the cascading effects of snow albedo and below-canopy turbulence and radiation transfer in forested regions as pointed out by Clark et al. (2015) and Zheng et al. (2015).

The Canadian boreal region contains one-third of the world’s boreal forest, approximately 6 million km$^2$ (Bryant et al., 1997). The boreal forests have complex interactions with the atmosphere and have significant impacts on regional and global climate (Bonan, 1991; Bonan et al., 1992; Thomas and Rowntree, 1992; Viterbo and Betts, 1999; Ciais et al., 1995). Several field experiments were conducted to better understand and model these interactions, including BOREAS (Boreal Ecosystem Atmosphere Study) and BERMS (Boreal Ecosystem Research and Monitoring Sites). Numerous studies have evaluated LSMs using the BOREAS and BERMS data (Bonan, 1997). Levine and Knox (1997) developed a frozen soil temperature (FroST) model to simulate soil moisture and heat flux and used BOREAS northern and southern study areas to calibrate the model. They found that soil temperature was underestimated and large model biases existed when snow was present. Bonan (1997) examined NCAR LSM1 with flux-tower measurements from the BOREAS, and found that the model reasonably simulated the diurnal cycle of the fluxes. Bartlett et al. (2002) used the BOREAS Old Jack Pine (OJP) site to assess two different versions of CLASS, the Canadian Land Scheme (2.7 and 3.0), and found that both versions underestimated the snow depth and soil temperature values, especially the version CLASS 2.7.

Boreal forest soils often have a relatively thick upper organic horizon. The thickness of the organic horizon directly affects the soil thermal regime and soil hydrological processes. Compared with mineral soil, the thermal and hydraulic properties of the organic soil are significantly different. Dingman (1994) found that the mineral soil porosity ranges from 0.4 to 0.6, while the porosity of organic soil is seldom less than 0.8 (Radford and Brawner, 1977). The hydraulic conductivity of organic soil horizons can be very high due to the high porosity (Boelter, 1968). Less suction is observed for a given volumetric water content in organic soils than in mineral soils, except when it reaches saturation. The thermal properties of the soil are also affected by the underground hydrology. Organic soil horizons also have relatively low thermal conductivity, relatively high heat capacity, and a relatively high fraction of plant-available water. Prior studies illustrated the importance of parameterizing organic soil horizons in LSMs for simulating soil temperature and moisture (e.g., Letts et al., 2000; Beringer et al., 2001; Mölders and Romanovsky, 2006; Nicolinsky et al., 2007; Lawrence and Slater, 2008).

The current Noah-MP model does not include a parameterization for organic soil horizons. It is thus critical to evaluate the effects of incorporating organic matter in surface energy and water budgets in order to enhance the global applicability of the WRF Noah-MP coupled modeling system. Here we conduct a detailed examination of the performance of the Noah-MP model in a Canadian boreal forest site. The main objective of this research is to enhance the modeling of vertical heterogeneity (such as organic matter) in soil structures and to understand its impacts on the simulated seasonal and annual cycle of soil moisture and surface heat fluxes. We recognize that Noah-MP has weaknesses in existing subprocess parameterizations; however the goal of this study is to explore the impact of incorporating organic soil in surface energy and water budgets, rather than comprehensively addressing errors in existing Noah-MP parameterization schemes. In this paper, we present the BERMS observation site in central Saskatchewan (Sect. 2) and our methodology for conducting 12-year Noah-MP simulations with and without the organic soil layer for that boreal forest site (Sect. 3). Section 4 discusses the simulations of the diurnal and annual cycles of the surface energy and hydrological components, in dry and wet periods. Summary and conclusions are given in Sect. 5.

2 BERMS site descriptions

The Old Aspen site (OAS, 53.7° N, 106.2° W, altitude 601 m) is located in mature deciduous broadleaf forest at the southern edge of the Canadian boreal forest in Prince Albert National Park, Saskatchewan, Canada (Fig. 1). The forest canopy consists of a 22 m trembling aspen overstory (Populus tremuloides) with ~10 % balsam poplar (Populus balsamifera) and a 2 m hazelnut understory (Corylus cornuta) with sparse alder (Alnus crispa). The fully leafed values of the leaf area index varied among years from 2.0 to 2.9 for the aspen overstory and 1.5 to 2.8 for the hazelnut understory (Barr et al., 2004). The forest was regenerated after a natural fire in 1919, and in 1998 it had a stand density of ~830 stems ha$^{-1}$. The soil is an Orthic Gray Luvisol (Canadian Soil Classification System) with an 8–10 cm deep forest floor (LFH; litter, fibric, and humic) organic horizon overlying a loam Ae horizon (0–21 cm), a sandy clay loam Bt horizon (21–69 cm), and a sandy clay loam Ck horizon (deeper than 69 cm). 30 % of the fine roots are in the LFH horizon and 60 % are in the upper 20 cm of mineral soil. The water table lies from 1 to 5 m below the ground surface, varying spatially in the hummocky terrain and varying in time in response to variations in precipitation. A small depression near the tower had ponded water at the surface during the wet period from 2005 to 2010. Mean annual air temperature and precipitation at the nearest long-term weather station are 0.4°C and 467 mm, respectively (Waskesiu Lake, 53°55′ N, 106°04′ W, altitude 532 m, 1971–2000 climatic normal).

Air temperature and humidity were measured at 36 m above ground level using a Vaisala model HMP35cf or HMP45cf temperature/humidity sensor (Vaisala Oyj, Helsinki, Finland) in a 12-plate Gill radiation shield (R.M. Young model 41002-2, Traverse City, MI, USA). Wind speed was measured using a propeller anemometer (R.M.
Young model 01503-, Traverse City, MI, USA) located at 38 m above ground level. Atmospheric pressure was measured using a barometer (Setra model SBP270, distributed by Campbell Scientific Inc., Logan, UT, USA). Soil temperature was measured using thermocouples in two profiles at depths of 0–15, 15–30, 30–60, 60–90, and 90–120 cm. Three of the eight probes that were the most free of data gaps were used in this analysis. The TDR probes were located in a low-lying area of the site that was partially flooded after 2004, resulting in high volumetric water content (VWC) values that may not be characteristic of the flux footprint. VWC is also measured at 2.5 and 7.5 cm depth in the forest-floor LFH layer, using two profiles of soil moisture reflect meters (model CS615, Campbell Scientific Inc., Logan, UT, USA), inserted horizontally at a location that did not flood.

Eddy-covariance measurements of the sensible and latent heat flux densities were made at 39 m above the ground from a twin scaffold tower. Details of the eddy-covariance systems are given in Barr et al. (2006). Data gaps were filled using a standard procedure (Amiro et al., 2006).

The net radiation flux density, $R_n$, was calculated from component measurements of incoming and outgoing shortwave and long-wave radiation, made using paired Kipp and Zonen (Delft, the Netherlands) model CM11 pyranometers and paired Eppley Laboratory (Newport, RI, USA) model PIR pyrgeometers. The upward-facing radiometers were mounted atop the scaffold flux tower in ventilated housings to minimize dew and frost on the sensor domes. The net radiometer and the downward-facing radiometers were mounted on a horizontal boom that extended 4 m to the south. The net radiation flux density, $R_n$, was calculated from component measurements of incoming and outgoing shortwave and long-wave radiation, made using paired Kipp and Zonen (Delft, the Netherlands) model CM11 pyranometers and paired Eppley Laboratory (Newport, RI, USA) model PIR pyrgeometers. The upward-facing radiometers were mounted atop the scaffold flux tower in ventilated housings to minimize dew and frost on the sensor domes. The net radiometer and the downward-facing radiometers were mounted on a horizontal boom that extended 4 m to the south. The net radiation flux density, $R_n$, was calculated from component measurements of incoming and outgoing shortwave and long-wave radiation, made using paired Kipp and Zonen (Delft, the Netherlands) model CM11 pyranometers and paired Eppley Laboratory (Newport, RI, USA) model PIR pyrgeometers. The upward-facing radiometers were mounted atop the scaffold flux tower in ventilated housings to minimize dew and frost on the sensor domes. The net radiometer and the downward-facing radiometers were mounted on a horizontal boom that extended 4 m to the south. The net radiation flux density, $R_n$, was calculated from component measurements of incoming and outgoing shortwave and long-wave radiation, made using paired Kipp and Zonen (Delft, the Netherlands) model CM11 pyranometers and paired Eppley Laboratory (Newport, RI, USA) model PIR pyrgeometers. The upward-facing radiometers were mounted atop the scaffold flux tower in ventilated housings to minimize dew and frost on the sensor domes. The net radiometer and the downward-facing radiometers were mounted on a horizontal boom that extended 4 m to the south.

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Soil parameters used in Noah-MP for mineral soil texture classes (sandy clay loam) and organic soil (Hemic Peat).

Table 2. Soil parameters used in Noah-MP for mineral soil texture classes (sandy clay loam) and organic soil (Hemic Peat).

<table>
<thead>
<tr>
<th>Soil type</th>
<th>( \lambda_s ) (w m(^{-1}) K(^{-1}))</th>
<th>( \lambda_{sat} ) (w m(^{-1}) K(^{-1}))</th>
<th>( \lambda_{dry} ) (w m(^{-1}) K(^{-1}))</th>
<th>( c_s ) (J m(^{-3}) K(^{-1}) \times 10(^6))</th>
<th>( \theta_{sat} ) (m s(^{-1}) \times 10(^{-3}))</th>
<th>( \kappa_{sat} ) (mm)</th>
<th>( \psi_{sat} ) (mm)</th>
<th>( b )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mineral</td>
<td>6.04</td>
<td>2.24</td>
<td>0.23</td>
<td>2.0</td>
<td>0.421</td>
<td>0.00445</td>
<td>-135</td>
<td>6.77</td>
</tr>
<tr>
<td>Organic</td>
<td>0.25</td>
<td>0.55</td>
<td>0.05</td>
<td>2.5</td>
<td>0.88</td>
<td>0.002</td>
<td>-10.3</td>
<td>6.1</td>
</tr>
</tbody>
</table>

The soil parameters are as follows: \( \lambda_s \) is the thermal conductivity of soil solids, \( \lambda_{sat} \) is the unfrozen saturated thermal conductivity, \( \lambda_{dry} \) is the dry soil thermal conductivity, \( c_s \) is the soil solid heat capacity, \( \theta_{sat} \) is the saturated volumetric water content (porosity), \( \kappa_{sat} \) is the saturate hydraulic conductivity, \( \psi_{sat} \) is the saturated matric potential, and \( b \) is the Clapp and Hornberger parameter.

Table 1. Noah-MP parameterization options used in this study.

<table>
<thead>
<tr>
<th>Parameterization description</th>
<th>Options</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dynamic vegetation</td>
<td>4: table LAI, shdfac = maximum</td>
</tr>
<tr>
<td>Stomatal resistance</td>
<td>1: BALL-Berry (Ball et al., 1987)</td>
</tr>
<tr>
<td>Soil moisture factor for stomatal resistance</td>
<td>1: original Noah (Chen and Dudhia, 2001)</td>
</tr>
<tr>
<td>Runoff/soil lower boundary</td>
<td>2: TOPMODEL with equilibrium water table (Niu et al., 2005)</td>
</tr>
<tr>
<td>Surface layer drag coefficient calculation</td>
<td>1: Monin–Obukhov (Bretscher, 1982)</td>
</tr>
<tr>
<td>Supercooled liquid water</td>
<td>1: no iteration (Niu and Yang, 2006)</td>
</tr>
<tr>
<td>Soil permeability</td>
<td>1: linear effects, more permeable (Niu and Yang, 2006)</td>
</tr>
<tr>
<td>Radiative transfer</td>
<td>3: two-stream applied to vegetated fraction</td>
</tr>
<tr>
<td>Ground surface albedo</td>
<td>2: CLASS (Verseghy, 1991)</td>
</tr>
<tr>
<td>Precipitation partitioning between snow and rain</td>
<td>1: Jordan (Jordan, 1991)</td>
</tr>
<tr>
<td>Soil temp lower boundary</td>
<td>2: TBOT at ZBOT (8 m) read from a file</td>
</tr>
<tr>
<td>Snow/snow temperature time</td>
<td>1: semi-implicit</td>
</tr>
</tbody>
</table>

As the value increased, the topsoil moisture decreased. Saturated matric potential and the Clapp and Hornberger parameter only influence the frozen period. For saturated matric potential, the topsoil moisture decreased when the parameter value increased, while for the Clapp and Hornberger parameter, the topsoil moisture increased when the parameter value increased. Based on the site measurement, the soil bulk density of the top layer is about 160 kg m\(^{-3}\). As described in Letts et al. (2000), this organic soil can be defined as hemic peat, a medium humified organic soil. Table 2 gives the recommended parameters for hemic peat, with 0.88, 2.0, 0.0102, and 6.1 for porosity, saturated hydraulic conductivity, saturated matric potential, and the Clapp and Hornberger parameter, respectively (Letts et al., 2000). From the sensitivity test mentioned above, it seems that the recommended values from Letts et al. (2000) produced soil moisture and soil temperature close to observations.

3.2 Forcing data

The 30 min meteorological observations, including air temperature, specific humidity, wind speed, pressure, precipitation, downward solar, and long-wave radiation, at 36 m height from OAS were used as atmospheric forcing data to drive Noah-MP in an offline 1-D mode. Figure 2 shows the annual mean temperature (1.5 °C) and total precipitation (406 mm) at this site during the study period (1998–2009). The most significant climatic features during the study pe-
period are a prolonged drought that began in July 2001 and extended throughout 2003, and an extended wet period from 2004 to 2007.

3.3 Evaluation of model performance

Outputs from the Noah-MP simulations were evaluated against observations, using the root mean squared error (RMSE), square of the correlation coefficient ($R^2$), and index of agreement (IOA) (Zhang et al., 2014). The IOA is calculated as

$$
\text{IOA} = 1 - \frac{\sum_{i=1}^{N} (M_i - O_i)^2}{\sum_{i=1}^{N} (|O_i - \bar{O}| + |M_i - \bar{O}|)^2},
$$

where $M_i$ and $O_i$ are simulated and observed values of the same variable, respectively, and $\bar{O}$ is the mean of the observed values. IOA ranges from 0 (no agreement) to 1 (perfect match).

4 Results and discussions

4.1 Noah-MP model spin-up

The LSM spin-up is broadly defined as an adjustment process as the model approaches its equilibrium following the initial anomalies in soil moisture content or after some abnormal environmental forcing (Yang et al., 1995). Without spin-up, the model results may exhibit drift as model states try to approach their equilibrium values. To initialize LSMS properly, the spin-up time required for LSMS to reach the equilibrium stage needs to be examined first (Chen and Mitchell, 1999; Cosgrove et al., 2003). In this study, model runs for the year 1998 were performed repeatedly until all the soil-state variables reached the equilibrium state, defined as when the difference between two consecutive 1-year simulations becomes less than 0.1 % for the annual means (Cai et al., 2014; Yang et al., 1995). Yang et al. (1995) discussed the spin-up processes by comparing results from 22 LSMS for grass and forest sites, and showed a wide range of spin-up timescales (from 1 to 20 years), depending on the model, state variable, and vegetation type. Cosgrove et al. (2003) used four NLDAS-1 LSMS to discuss the spin-up time at six subregions covering North America, and showed that all models reached equilibrium between 1 and 3 years for all six subregions. In this study, we found that it requires 9 years for deep-soil moisture (100–200 cm layer) in Noah-MP to reach its equilibrium, 8 years for latent heat flux and evapotranspiration, but only 3 years for the surface soil moisture (Fig. 3). Cosgrove et al. (2003) and Chen et al. (1999) indicated that it takes a long time to reach equilibrium, especially in the deep soil layers and sparse vegetation, because the evaporation was limited by slow water diffusion timescales between the surface and deep soil layers. When using the groundwater component of Noah-MP, it might take at least 250 years to spin up the water table depth in arid regions (Niu et al., 2007). Cai et al. (2014) found that water table depth requires less than 10 years to spin up in a wet region, but more than 72 years for a dry region. For this boreal forest site where the water table depth is shallower (less than 2.5 m), it takes $\sim 7$ years for water table depth to reach equilibrium. However, the freezing/thawing is a relatively slow process, so we set 10 years for the spin-up time for all the experiments discussed here.

4.2 Seasonal cycle of soil temperature and moisture

We defined the simulation without incorporation of organic soil as the control experiment (CTL), and the simulation with the organic soil incorporated as the organic layer experiment (OGN). We first evaluated the CTL- and OGN-simulated soil temperature and moisture at the OAS site in relation to observations for the period of 1998–2009.

As shown in Fig. 4, the effects of including a 10 cm organic topsoil layer on simulated soil temperature are not uniform both throughout the soil depth and during the year. Figure 4a...
shows that the CTL and OGN simulations produced nearly identical top-layer temperatures which are in agreement with the observations except for a low bias in the winter period, especially during drought years 2002–2003. However, for deep layers (10–100 cm), soil temperature from the OGN is lower (higher) than the CTL simulation during summer (winter), especially for the drought years 2002–2003, leading to a good agreement between OGN and observations for second- and third-layer soil temperature (Fig. 4b, c). Lawrence and Slater (2008) indicated that strong cooling in summer is due to the modulation of early and midsummer soil heat flux, while higher soil temperature in fall and winter is due to less efficient cooling of organic soils. The soil thawing period in spring is significantly affected by the OGN parameterization since the thermal conductivity of the organic horizon is much lower than that of the mineral soil (∼0.4 W m⁻¹ K⁻¹ compared to ∼2.0 W m⁻¹ K⁻¹), which delays the warming of the deep soil layers after snowmelt. In winter, the organic soil layer insulates the soil and results in relatively higher wintertime soil temperatures for OGN compared with CTL. The difference is most pronounced in drought years (2002 and 2003) (Fig. 4). In summer, due to lower saturated thermal conductivity (0.25 W m⁻¹ K⁻¹ for organic compared to ∼6.04 W m⁻¹ K⁻¹ for mineral) in OGN, the downward transfer of heat from the topsoil layer is less and the deep soil temperature in OGN is lower than that in CTL.

In winter, with the presence of soil ice, the thermal heat conductivity in OGN (∼2.20 W m⁻¹ K⁻¹) is lower than that in CTL (6.04 W m⁻¹ K⁻¹); it reduces the upward transfer of heat from deep soils to topsoil and therefore results in higher deep-soil temperature in OGN. These results are consistent with studies that showed a simulated increase in winter soil temperature of up to 5°C in boreal regions when including an organic layer (Koven et al., 2009; Rinke et al., 2008; Lawrence and Slater, 2008). For the topsoil layer, the OGN parameterization increases the liquid soil water content in summer as water fills the larger pore space of organic soil, though the liquid soil water content in winter did not change much, due to the contrasting water retention characteristics of organic and mineral soil (Koven et al., 2009; Rinke et al., 2008; Lawrence and Slater, 2008). Higher porosity in OGN leads to an increase in total soil water content, while the lower topsoil temperature (Fig. 4a) in OGN enhances the ice content. Note that the observed soil water content during wet years may be higher than the site truth because the sensors were located in a low spot that is prone to flooding. This site got flooded in 2004 and the ground water has not dried since then; so the soil was oversaturated during the period of 2004–2008. In the second soil layer, the observed soil water content was incorrect after the site got flooded (2004–2008). With more precipitation during the wet period, the real soil water content should have a relatively high value. Since the OGN increases the soil water content, it should be closer to the true observation. From Fig. 5, it can be seen that the OGN improved the liquid water simulation in non-frozen periods. The soil moisture data are not reliable when the soil is frozen and are therefore not very useful during the winter. In late spring when snow starts melting, both CTL and OGN simulate the same topsoil temperature (Fig. 4). It is clear that the soil liquid water content is mainly controlled by precipitation, soil hydraulic conductivity, and runoff. The high porosity of organic soil in the topsoil layer helps to retain more snowmelt water and hence increases the topsoil layer liquid water content. For the deep soil layers, the soil liquid water content is highly influenced by the soil temperature. Liquid soil water content increases during soil-ice thawing period. The higher deep soil layer liquid water content in OGN is mainly because the soil hydraulic conductivity is higher for organic soil than mineral soil, so liquid water in the first layer can be transported downward quickly into the deeper layers. Although the organic soil layer is only added to the first two layers in this study, it still can affect the deep layer due to the infiltration characteristics of the topsoil.

The water retention characteristics of the organic soil horizon favor both higher water retention and reduced evaporation. The thermal conductivity is lower compared with that of the mineral soil, which then prevents the deeper soil from warming up rapidly after the snowmelt season. The lower thermal conductivity of the top organic soil affects the annual cycle of the ground heat flux. In summer, the top layer is warmer than the deep layers; the ground heat flux then transfers heat downward. Because air temperature is lower than land surface temperature, heat is transferred upward from soil to the land surface; the low thermal conductivity of the organic soil can prevent the soil from cooling. On the other hand, snowfall in winter may form a snow layer that will insulate the soil and make the simulations less sensitive to thermal conductivity. This may be the reason why the OGN-simulated winter soil temperature is higher compared...
to CTL simulations. With the organic soil layer on the top, the reduction of surface layer saturation levels in wintertime (Fig. 5) reduces the heat loss through evaporation. The winter soil temperature then becomes significantly higher compared with the CTL experiment. On the contrary, the higher soil water content in the topsoil layer during summertime (Fig. 5) increases the heat loss through evaporation; the summer soil temperature then becomes significantly lower compared with the CTL experiment.

### 4.3 Seasonal cycles of sensible and latent heat flux

Simulated differences in top-layer soil temperature and liquid soil water content lead to the differences in simulated surface energy fluxes. Figure 6 shows that the CTL run captures the observed monthly mean daytime sensible heat and latent heat flux reasonably well. However, SH is underestimated in spring and overestimated in summer. Accordingly, LH is overestimated in spring and underestimated in summer during most of the time period except for drought years 2002–2003 where LH is slightly overestimated. Generally, the OGN simulations show similar characteristics to the CTL, with improved correlation coefficients between observations and simulations: increasing from 0.88 (CTL) to 0.92 (OGN) for SH and from 0.94 (CTL) to 0.96 (OGN) for LH (Fig. 7). Overall, both CTL and OGN perform well in winter when snow is present and fluxes are small. During the spring snowmelt season, the OGN results are much better than the CTL (Figs. 6 and 7).

The OGN simulations also improved the underestimation of SH in spring in CTL, but they still overestimate summer SH. The reason for high bias in summer SH will be further discussed in Sect. 4.4. SH and especially LH show improvement in OGN compared to CTL, which is related to timing of soil thaw and warming in spring. CTL thaws the soil too early, causing a premature rise in LH in spring (April–May) and an associated underestimation of spring SH. The spring (April–May) fluxes are much improved in the OGN parameterization. However, both OGN and CTL retain a serious positive bias in SH from June to September, especially for wet years. The reduction of surface layer saturation levels in OGN led to lower soil evaporation and associated reductions in the total latent heat flux, and the reduction of LH is accompanied by a rise in SH (Fig. 6).

### 4.4 Impact of organic soil on diurnal cycle of surface energy and hydrology

The quality of nighttime flux-tower data is questionable (Chen et al., 2015), especially for OAS located in a boreal forest. Therefore, we focused our analysis on daytime observation data. In general, the OGN parameterization improved the simulation of daily daytime LH in terms of both RMSE and IOA, and increased IOA for SH (Table 3). Nevertheless, compared with CTL, OGN increased the bias in SH slightly by ~ 3% (Table 3).

For the 12-year simulation period, the study site experienced a prolonged drought that began in July 2001 and extended throughout 2002 and 2003. We choose year 2002 and 2003 to represent typical drought years, and year 2005 and 2006 to represent typical wet years (Fig. 2), to examine the effect of the organic soil under different climate conditions. For drought years 2002–2003, OGN increased daytime SH especially in spring, and slightly decreased SH at nighttime (Fig. 8, a, b, c, and d). LH is well simulated in both OGN and CTL (Fig. 8e, f, g, and h), with slightly increased daytime LH in OGN. OGN overestimates daytime SH compared with observations, while CTL underestimates daytime LH for spring (Fig. 8a). Both OGN and CTL overestimate SH for summer, autumn, and winter (Fig. 8b, c, d).
Table 3. Averaged statistical indices for CTL- and OGN-simulated SH and LH compared with the observations for each year (daytime, 08:00–16:00 local time (LT)) ($R^2$: correlation coefficient square; RMSE: root mean square error; IOA: index of agreement).

<table>
<thead>
<tr>
<th>Year</th>
<th>SH</th>
<th></th>
<th>LH</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>$R^2$</td>
<td>RMSE</td>
<td>IOA</td>
<td>$R^2$</td>
</tr>
<tr>
<td>1998</td>
<td>0.56</td>
<td>80.92</td>
<td>0.83</td>
<td>0.72</td>
</tr>
<tr>
<td>1999</td>
<td>0.64</td>
<td>64.30</td>
<td>0.88</td>
<td>0.74</td>
</tr>
<tr>
<td>2000</td>
<td>0.62</td>
<td>71.20</td>
<td>0.87</td>
<td>0.70</td>
</tr>
<tr>
<td>2001</td>
<td>0.72</td>
<td>63.09</td>
<td>0.90</td>
<td>0.78</td>
</tr>
<tr>
<td>2002</td>
<td>0.75</td>
<td>69.60</td>
<td>0.91</td>
<td>0.69</td>
</tr>
<tr>
<td>2003</td>
<td>0.77</td>
<td>56.52</td>
<td>0.93</td>
<td>0.73</td>
</tr>
<tr>
<td>2004</td>
<td>0.72</td>
<td>61.88</td>
<td>0.91</td>
<td>0.73</td>
</tr>
<tr>
<td>2005</td>
<td>0.69</td>
<td>60.98</td>
<td>0.90</td>
<td>0.76</td>
</tr>
<tr>
<td>2006</td>
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<td>67.70</td>
<td>0.86</td>
<td>0.77</td>
</tr>
<tr>
<td>2007</td>
<td>0.65</td>
<td>65.15</td>
<td>0.89</td>
<td>0.76</td>
</tr>
<tr>
<td>2008</td>
<td>0.71</td>
<td>63.54</td>
<td>0.91</td>
<td>0.76</td>
</tr>
<tr>
<td>2009</td>
<td>0.69</td>
<td>66.52</td>
<td>0.90</td>
<td>0.72</td>
</tr>
</tbody>
</table>

Figure 7. Scatter plots of the daytime monthly averaged (a) sensible heat fluxes and (b) latent heat fluxes (W m$^{-2}$) for CTL vs. the observation above the canopy; the monthly averaged (c) sensible heat fluxes and (d) latent heat fluxes (W m$^{-2}$) for OGN vs. the observation above the canopy. The color represents each month from January (1) to December (12).

For wet years (Fig. 9), OGN produces higher daytime SH than CTL in general. For spring, OGN-simulated SH agrees with the observation better than CTL, but it is similar to or slightly worse than CTL for other seasons. Simulated LH for both OGN and CTL agree with observations well, with an improvement by OGN in spring, because the snowmelt process dominates during spring months. For other seasons, the OGN results are close to CTL.

It is clear from Figs. 4, 8, and 9 that in both CTL and OGN, summer sensible heat fluxes are overestimated for wet and dry years. We hypothesized that such high bias in summer sensible heat flux is partly attributed to energy imbalance in observations. We then calculated the energy balance residual term: $R_{net} - (SH + LH + G)$ for summer months (June, July, and August). In wet years, $G$ in CTL and OGN is close to observed values; modeled latent heat flux is underestimated by $\sim 10$ W m$^{-2}$; modeled sensible heat flux is overestimated by $\sim 30$ W m$^{-2}$; and the residual term is $\sim 17$ W m$^{-2}$. Hence, it is reasonable to argue that the surface energy imbalance ($\sim 17$ W m$^{-2}$) in observations contributes to a large portion of the $\sim 30$ W m$^{-2}$ high bias in sensible heat fluxes. In dry
years, the summer energy imbalance (~15 W m\(^{-2}\)) is nearly equal to the high bias in sensible heat flux (~15 W m\(^{-2}\)).

### 4.5 Impact of an organic soil horizon on annual cycle of surface energy and hydrology

In the previous section, it is clear that the incorporation of the top organic layer helps improve the simulation of the diurnal cycle of the surface energy and hydrologic components in spring season. In the following, we focus on a detailed analysis of the annual cycle of the surface energy and hydrology variables for dry (Fig. 10) vs. wet years (Fig. 11). Between June and September as shown in Fig. 10h, the upper two soil layers were unfrozen. The topsoil is wetter in OGN for both dry and wet years compared with CTL because organic soil can retain more water. As discussed in Sect. 4.2, for the deep soil layers, the liquid water content is influenced by the soil temperature and the movements of the soil liquid water content between soil layers. Since the soil hydraulic conductivity is higher for OGN than mineral soil, the water moves faster into deep soil layers than CTL; therefore the OGN simulates higher soil liquid water content in deep layers. OGN has a major impact on the daily cycle of soil temperature. Consistent with discussions in Sect. 4.2, the soil temperature below 10 cm simulated by OGN is lower in summer and higher in winter than that of the CTL simulation, and the OGN simulation shows less bias than the CTL simulation (Fig. 4). In the OGN simulation, the water moves faster into deep layers than in the CTL simulation, leading to more infiltrated water in the deep soil and hence a higher base flow. Consequently, the total runoff is increased. Due to the high soil porosity of the organic soil, OGN simulation shows higher soil-ice fraction at the topsoil layer during the freezing periods. The higher water capacity and higher soil-ice fraction of the organic soil then reduce liquid water content/soil moisture, leading to less evaporation (i.e., latent heat flux) during spring freezing periods, and a compensating increase of the sensible heat flux.

By adding an organic soil layer, the soil-ice content becomes higher due to higher porosity. For dry years, the impact of the organic soil on surface and subsurface runoff is not significant (Fig. 10e, f). The increase in the summer latent heat flux and sensible heat flux are compensated by a decrease in soil heat flux, leading to a significant decrease in summer soil temperature. In winter, the latent and sensible

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**Figure 8.** Comparison of the seasonal averaged diurnal cycle of the sensible and latent heat fluxes at OAS site for drought years.

**Figure 9.** Comparison of the seasonal averaged diurnal cycle of the sensible and latent heat fluxes at OAS site for wet years.
heat fluxes are not modified by the organic soil, but increased
soil heat flux leads to an increased soil temperature in win-
ter. The most prominent change by including the organic soil
layer is the partition between vegetation transpiration and di-
rect ground evaporation (Fig. 12a and b), where the OGN
simulation slightly increased ground surface evaporation and
vegetation transpiration.

For wet years (Fig. 11), the impact of the organic soil
on surface and subsurface runoff becomes more significant,
especially for subsurface runoff. The organic soil decreases
the surface runoff during the summer season, and increases
the subsurface runoff during the freezing periods, while it
decreases the subsurface runoff during summer season. Be-
cause of the higher surface layer soil-ice content, the in-
crease of subsurface flow should be due to the production
of a wetter soil profile by OGN. The sensible heat flux also
increases significantly in spring, with an associated reduc-
tion in latent heat flux and soil heat flux. The summer soil
temperature also decreases but to a lesser degree than that
in dry years, because the soil heat flux decreases less com-
pared with dry years. Unlike dry years, there is a significant
runoff change in wet years, and the ground evaporation is
also decreased (Fig. 12c and d). OGN produces more soil-
ice content and higher soil porosity, and leads to higher soil
Additionally, due to higher porosity of the organic soil, the OGN simulation was able to retain more soil water content in summer. However, the effects of including an organic soil layer on soil temperature are not uniform throughout the soil depth and year, and those effects are more prominent in summer and in deep soils.

For drought years, the OGN simulation substantially modified the partition between direct soil evaporation and vegetation transpiration. When water is limited in drought years, the OGN simulation slightly increased the direct soil evaporation and produced higher summer total evapotranspiration. Increased latent heat flux and sensible heat flux in summer in OGN are compensated by decreased soil heat flux, leading to reduced soil temperature in summer. For wet years, the OGN-simulated latent heat fluxes are similar to CTL, except for the spring season where OGN produced less evaporation. In addition, the impact of the organic soil on subsurface runoff is substantial with much higher runoff in freezing periods and lower runoff in summer season.

This preliminary study explored the effects of incorporating organic soil parameterization in Noah-MP on the surface energy and water cycles for one flux site in a boreal forest area. Given the important role of boreal forests in the regional climate system through reducing winter albedo and also acting as a carbon sink and water source to the atmosphere, further work is needed to evaluate the Noah-MP with organic soil parameterization at regional scales. We plan to evaluate the performance of the offline Noah-MP model and Noah-MP coupled with WRF for a broad boreal forest region including Alberta and Saskatchewan.

6 Data availability

The code for incorporation of an organic soil layer in the Noah-MP land surface model is available upon request from Liang Chen at the University of Saskatchewan (liang.chen@usask.ca). The FLUXNET data are publicly available from the ORNL DAAC (Distributed Active Archive Center) at ftp://daac.ornl.gov/data/fluxnet/fluxnet_canada/data/SK-OldAspen/ (ORNLA DAAC, 2016).

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