We thank the editor and reviewers very much for the dedicated time and efforts they put to improve/sharpen this manuscript with constructive comments. We made our point-by-point response in blue fonts as below. The referee comments are in black fonts.

RC1: Referee comments from #1 anonymous referee.

RC2: Referee comments from #2 anonymous referee.

Anonymous Referee #1

Referee comment on "STEMMUS-UEB v1.0.0: Integrated Modelling of Snowpack and Soil Mass and Energy Transfer with Three Levels of Soil Physical Process Complexities" by Lianyu Yu et al., Geosci. Model Dev. Discuss., https://doi.org/10.5194/gmd-2020-416-RC1, 2021

This manuscript aims to incorporate the snowpack effect into a STEMMUS-FT modeling framework, with various complexities of mass and energy transfer physics, then investigate the effect of snowpack on soil moisture and heat transfer. In general, the manuscript is well written and interesting to me. I recommend a major revision for this manuscript before its acceptance for publication.

We thank the reviewer very much for the time and effort and also for the insightful comments. Please see our specific response in blue fonts below.

General comments:

[1] RC1: There are too many long sentences which make them hard to follow.

Response: We have made changes, separate the long sentences into short ones, trying to make it easy to read and follow.

[2] RC1: Can the simulated time series of daily average albedo and LE (latent heat flux) be longer?

Response: We thank the reviewer for the insightful comments. Yes, the temporal spectrum of daily average albedo and LE (latent heat flux) can be longer. For instance, we have run about 3-year simulation (Mar. 2016 – Aug. 2018) of surface energy fluxes, including radiative components, sensible heat fluxes and LE, against the in-situ observations for this site. It well demonstrated the performance of STEMMUS-FT (Yu et al., 2020). The UEB model also shows its widely and successful application spanning a variety of hydrological conditions (Table S3). All these give us confidence that the integrated STEMMUS-UEB model can be applicable to this experimental site. On the other hand, we have to admit that as the harsh environment, the dataset is difficult to achieve and to have it fully corroborated. The accuracy of precipitation measurement (both the amount and its liquid/solid fractions), which is important to have the snowpack dynamics right, is of uncertain and needs effort to have more dataset to constrain it. We add some text in Section 4.1 Limitations (Line 342).

For this work, we focused on developing the integrated STEMMUS-UEB model (in Sect. 2 and Supplement). Furthermore, upon the confirmation of STEMMUS-UEB model performance, we are trying to emphasize/demonstrate its capability for understanding the effect of snowpack on the subsurface soil water and heat transfer processes. With the aid of both in-situ measurements and numerical experiments, we can see that: i) the presence of snowpack can be identified by the abrupt dynamics of daily average albedo. STEMMUS-UEB model well captured the large abrupt of daily average albedo with the precipitation. ii) models considering the snowpack process generally presented better simulation performance than models without snowpack. Then we further illustrated the capability of STEMMUS-UEB, in terms of understanding how snow water infiltrates downwards and interacts with subsurface soil water and heat regimes.

[3] RC1: The freezing and melting processes are a cyclic process, it will be more reasonable to describe the two processes together (section3.4.1 & 3.4.2).

Response: We agree that freezing and melting processes are inherently bounded together. We made changes about section 3.4.1 & 3.4.2 according to reviewer's comments. First, we presented diurnal dynamics of the observed and model simulated latent heat flux (LE) during the rapid freezing and thawing periods with precipitation events in Figure 6 and 7, respectively.

Then the relative contribution of individual flux components to the total mass transfer were presented in Figure 8 and 9, respectively, for the freezing and thawing periods.

The freezing and thawing processes were described together in Section 3.4.

[4] RC1: The language could be polished in various places in order to facilitate understanding.

Response: Thanks a lot for the helpful comments. We had the manuscript English edited.

Specific comments:

[5] RC1: The overview of the coupled STEMMUS-FT and UEB model framework and model structure in figure 1, the text is too small to read.

Response: We made modifications of the text in figure 1 and also made it as Landscape orientation instead of Portrait orientation.

[6] RC1: Figure 6 is too long, which can be divided into three figures or rearranged. Time series of different variables overlapped and changes in different variables are not visible. Such as qLh and qLT. Same as other figures.

Response: Figures were rearranged to make it easy to read. Figure 6 and Figure 7 were further presented as Figure 6, Figure 7, Figure 8, and Figure 9. The state variables were presented together as a cyclic freezing-thawing process. The comparison of observed and model simulated latent heat fluxes (LE) for the freezing and thawing periods were given sequentially in Figure 6 and Figure 7. Figure 8 and Figure 9 present the model simulated latent heat flux and surface soil thermal and isothermal liquid water and vapor fluxes for the freezing and thawing periods (Section 3.4).

We use different type of lines (dotted line for q_{LT}, dashed lines for q_{Vh}, and q_{Va}, and solid lines for q_{VT}, q_{Lh}, q_{La}, with different colors) to make the different flux component visible in figures.

[7] RC1: In Figure 7 (e, f, h, i), the sharp changes should be explained on Day 103. The figure and legend are overlapped.

Response: We moved the legend a little bit to avoid its overlapping with the figure content. For the sharp changes of isothermal liquid and vapor fluxes (q_{Lh}, q_{Vh}), it is due to the large increase of surface soil moisture after the precipitation events (see Supplement Figure S2 d & f). This resulted in the large gradient of matric potential thus the sharp changes of isothermal liquid and vapor fluxes. We add the explanation in Section 3.4 (Line 321).

Anonymous Referee #2

Referee comment on "STEMMUS-UEB v1.0.0: Integrated Modelling of Snowpack and Soil Mass and Energy Transfer with Three Levels of Soil Physical Process Complexities" by Lianyu Yu et al., Geosci. Model Dev. Discuss., https://doi.org/10.5194/gmd-2020-416-RC2, 2021

[1] RC2: This study presents a new integrated modelling of snowpack and soil water/energy transfers, called STEMMUS-UEB, presenting three levels of soil transfer complexities. The model is evaluated on one site equipped with soil temperature, moisture and energy fluxes sensors. The performances of the 3 options of the model are discussed. This is an interesting paper but quite difficult to follow and some questions need to be addressed before further consideration for publication.

Response: We thank very much the reviewer for the insightful comments. Please see our specific response as below.

[2] RC2: A general issue is that the test site seems to be poorly influenced by snow. I am therefore wondering if it is really appropriate for the model evaluation.

Response: Great thanks for this critical comment. As partly explained in **[2] RC1**, this work is to describe the integrated soil-snow model STEMMUS-UEB and further confirm that STEMMUS-UEB model can identify and understand the snowpack effect, even for regions with the intermittent snowfall events.

The selected site is covered by a seasonal frozen ground with mostly episodic snowfall events. Compared to the sites with heavy snow events and thick snow layer, this site is indeed less influenced in some sense, e.g., snowmelt runoff. Nevertheless, the snow accumulation and snowmelt infiltration and its effects on the subsurface soil can still be identified, which indicates how sensible the STEMMUS-UEB can represent such intermittent snowfall events.

In addition, such conditions (seasonal frozen ground with episodic snowfall events) commonly exist in the high mountain cold regions around the globe, while its implications for subsurface water and heat dynamics (therefore, the subsequent impact on land-atmosphere interactions) are rarely studied. Nevertheless, the STEMMUS-UEB can be applied for the sites with thick snow cover or perennial snowpack, since the coupled model is constructed to account for the physically-based processes.

From the perspective of experimental measurements, it is indeed helpful to enrich the snowpack-relevant data (e.g., high resolution observation of the spatiotemporal field of wind speed, precipitation, and snowpack variations) and make them more constraint and less of uncertainty. We add some text in Section 4.1 Limitations (Line 342).

[3] RC2: The model description in the main paper lacks on the description of the thawing/freezing processes: how is the fraction of liquid/solid water calculated and what about the soil hydraulic conductivity? how the rainfall/snowfall partition is done?

Response: We added the description of thawing/freezing process in Section 2.1 (Line 98 and Line 124).

a. how is the fraction of liquid/solid water calculated and what about the soil hydraulic conductivity?

The frozen soil physics considered in STEMMUS-FT include three parts: i) the ice blocking effect on soil hydraulic conductivities (see Supplement Sect. 2.2.2); ii) the inclusion of ice effect in the calculation of soil thermal capacity/conductivity (see Supplement Sect. 2.2.8); iii) the exchange of latent heat flux during phase change periods. With the aid of Clausius Clapeyron relation, which characterizes the phase transition between liquid and solid phase in the thermal equilibrium system. The soil water characteristic curve (e.g., van Genuchten, 1980) is then extended to consider the freezing temperature dependence, i.e., soil freezing characteristic curve (Hansson et al., 2004; Dall'Amico et al. 2011). The fraction of soil liquid/solid water at a given temperature was then calculated prognostically with the soil freezing characteristic curve. Soil hydraulic parameters were further used in the Mualem (1976) model to compute the soil hydraulic conductivity. The ice effect is considered by reducing the soil saturated hydraulic conductivity as the function of ice content (Yu et al., 2018).

b. how the rainfall/snowfall partition is done?

Two precipitation types, i.e., rainfall and snowfall, are discriminated by the dependence on air temperature.

[4] RC2: Figure 2 is too small and difficult to read

Response: We rescaled the y-axis (from [0,1] to [0.1,0.7]) and enlarged Figure 2 to make it easy to read.

[5] RC2: Figures 6 and 7 are also difficult to understand: the precipitation events are rainfall or snowfall? what is the amount of SWE during that periods? It is surprising to see that the model without snow modeling performs generally better in the simulation of the latent heat flux compared to the snow model. It would be necessary to elaborate a bit more on that result.

Response: Figure 6 and 7 were modified correspondingly. Both the precipitation amount and the simulated snowfall (SWE) component were presented in the updated figures. The rainfall is the precipitation minus snowfall component.

For the discrepancies in terms of latent heat flux for the selected periods, it is possibly due to the inaccurate precipitation measurements and interpretation, in terms of either the amount, time of precipitation, or the partition of precipitation into solid and liquid forms. Moreover, the simple air temperature-precipitation type relation maybe not suitable for this region. As argued by Ding et al. (2014), air temperature is not the best indicator of precipitation types. Other factors, i.e., relative humidity, elevation, and wet-bulb temperature, are also very relevant and should be taken into account. The uncertainties in discriminating the precipitation types can be the possible reason here. The episodic snowfall events are challenging to be well captured and simulated by the current snowpack models. The snowpack accumulation, melting process and water and energy partitioning of snowpack into snow sublimation and the snowmelt are with uncertainties as well.

We add some more text to explain and discuss such limitations in Section 4.1 (Line 349).

Nevertheless, this work focused on identifying the snowpack impacts on the underlying subsurface water and heat dynamics. And, the difference between the model simulations with and without the snow module can be attributed to the different surface water and energy regimes. Models without snow regards the precipitation as the rainfall, i.e., liquid form of water, adds on the topsoil surface immediately. Most of the incoming water directly contributes to the infiltration process.

While for the model with snow module, it considers snowpack related processes, accumulation, sublimation, melting process and then the water infiltration process. Compared with the model without snow module, the increase of surface soil moisture is usually delayed and less significant.

The difference in surface water status results in different gradients of matric potential. Models without snow have larger gradient of matric potential for the top surface soil layer. Then more amount of isothermal liquid and vapor fluxes (q_{Lh}, q_{Vh}) were generated contributing to the total latent heat flux.

Only with consideration of two-phase flow (ACD, ACD-air), the difference between models with and without snow module can be identified during the daytime after winter precipitation events. Generally, from the foregoing, considering the vapor flow/airflow retarded the total surface water transfer (Figure 6 a vs. b/c).

Identifying the difference and understanding what lies behind these differences between models with and without snowpack can be only made by two phase flow model (ACD, ACD-air). These are the highlighting points and benefits of the developed integrated soil-snow model STEMMUS-UEB.

[6] RC2: In the title, I suggest to replace "mass" by "water" to be more precise.

Response: We replace "mass" with "water" in the title.

[7] RC2: The English need to be revised

Response: We carefully revised the English as suggested.

[8] RC2: The abstract need to be rewritten to better highlight the main findings of the work

Response: We rewritten the abstract to highlight the main take home messages of this work.

The main findings of the work are briefly summarized here as:

i) we developed an integrated soil-snow-atmosphere model, STEMMUS-UEB, which takes advantage of the easily transferable and physically based description of the snowpack process by UEB snowmelt model and the detailed interpretation of the soil physical process by STEMMUS-FT model.

ii) the proposed model can well capture the dynamics of daily average albedo, latent heat flux, and the snowpack effect.

iii) three mechanisms, i.e., surface ice sublimation, snow sublimation and increased soil moisture, contribute to the enhanced latent heat flux after winter precipitation events.

iv) Physically realistic analysis of the snowpack effects (e.g., LE enhancement) can only be reproduced by the advanced coupled STEMMUS-UEB (ACD, ACD-air). The basic coupled version of STEMMUS-UEB (BCD) models, however, cannot provide a realistic description of the soil water and heat transfer.

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STEMMUS-UEB v1.0.0: Integrated Modelling of Snowpack and Soil ~~Mass~~ Water and Energy Transfer with Three Complexity Levels of Soil Physical Process ~~Complexities~~

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Abstract

~~Snowpack~~A snowpack~~, as the indispensable component in cold regions,~~ has a profound effect on the hydrology and surface energy conditions of an area through its ~~modification~~ effects on~~of the~~ surface albedo, roughness, and its insulating property. ~~Although t~~The modelling of ~~the~~ a snowpack, soil water dynamics, and the coupling of the snowpack and underlying soil layer has been widely reported~~,~~. However, the ~~analysis of~~ coupled liquid-vapor-air flow mechanisms considering the snowpack effect ~~was~~ have not been~~yet~~ investigated in details. In this study, we incorporated the snowpack effect (Utah Energy Balance model, UEB) into a common modeling framework (Simultaneous Transfer of Energy, Mass, and Momentum in Unsaturated Soils with Freeze-Thaw, STEMMUS-FT~~),~~), i.e., STEMMUS-UEB. It considers soil water and energy transfer physics with three complexity levels ~~various complexities of mass and energy transfer physics~~ (from the basic coupled, to advanced coupled water and heat transfer, and further to the explicit consideration of airflow, termed BCD, ACD, and ACD-air, respectively). We then utilized ~~the in~~ in-situ observations and numerical experiments to investigate the effect of snowpack on soil moisture and heat transfer with the above-mentioned model complexities. Results indicated that the proposed model with snowpack can reproduce the abrupt increase of surface albedo after precipitation events ~~can be only reproduced by models considering~~while this was not the case for the model without snowpack. The BCD model tended to overestimate the land surface latent heat flux (LE). Such overestimations were largely reduced by ACD and ACD-air models. Compared with the simulations considering snowpack, there is less ~~surface latent heat flux~~LE from no-snow simulations due to the neglect of snow sublimation. ~~With coupled models, t~~The ~~enhanced~~ enhancement of LE was found ~~latent heat flux~~ after winter precipitation events, which is~~can be~~ sourced from the surface ice sublimation, snow sublimation, and increased surface soil moisture~~,~~. The relative role of the mentioned three sources depends on the timing and magnitude of precipitation and the pre-precipitation soil hydrothermal regimes. ~~while t~~The simple BCD model cannot provide ~~the~~ a realistic

partition of ~~surface latent heat flux~~mass transfer flux. The ACD model, with its physical consideration of vapor flow, thermal effect on water flow, and snowpack, can identify the relative contributions of different components (e.g., thermal or isothermal liquid and vapor flow) to the total mass transfer fluxes. With the ACD-air model, the relative contribution of each component (mainly the isothermal liquid and vapor flows) to the mass transfer was significantly altered during the soil thawing period. It was found that the snowpack affects not only the soil surface moisture conditions (surface ice and soil water content in the liquid phase) and energy-related states (albedo, LE) but also the transfer patterns of subsurface soil liquid and vapor flow.

1. Introduction

In cold regions, the snowpack has a profound effect on hydrology and surface energy through its ~~modification~~ change of ~~the~~ surface albedo, roughness and insulating property (Boone and Etchevers, 2001; Zhang, 2005). Different than rainfall, ~~precipitation water of~~the melted snowfall enters the soil with a significant lag in time~~.~~ ~~However~~, and a large and sudden outflow or runoff may be produced because of the snowmelt effect. The heat insulating ~~effect~~ property of snow cover also provides a buffer layer to reduce the magnitude of the underlying subsurface temperature variations and thus markedly affect the thickness of the active layer in cold regions. The effect of snow cover on the subsurface soils has been studied and reviewed (e.g., Zhang, 2005; Hrbáček et al., 2016). For instance, snow cover can act as an insulator between atmosphere and soil with its low thermal conductivity (Zhang, 2005; Hrbáček et al., 2016). The snowmelt functions as the energy sink ~~with~~ by the absorption of heat due to phase change (Zhang, 2005). Yi et al. (2015) investigated the seasonal snow cover effect on the soil freezing/thawing process and its related carbon implications. Such studies mainly focus on the thermal effect of snowpack on the frozen soils~~,~~. ~~however~~However, the effect of snowpack on the soil water and vapor transfer process is rarely reported (Hagedorn et al., 2007; Iwata et al., 2010; Domine et al., 2019).

Great amounts of modeling efforts have been made to better reproduce the snowpack characteristic and its effects. Initially, snowpack dynamics ~~was~~ were expressed as a simple ~~empirical~~ function of temperature. Nevertheless, these empirical relations have limited applications in complex climate conditions (Pimentel et al., 2015). Many physically-based models for the mass and energy balance in the snowpack have been developed for their coupling with hydrological models or atmospheric models. Boone and Etchevers (2001) divided these snow models into three main categories: i) simple force-restore schemes with the snow modeled as the composite snow-soil layer (Pitman et al., 1991; Douville et al., 1995; Yang et al., 1997) or a single explicit snow layer (Verseghy, 1991; Tarboton and Luce, 1996; Slater et al., 1998; Sud and Mocko, 1999; Dutra et al., 2010); ii) detailed internal-snow-process schemes with multiple snow layers of fine vertical resolution (Jordan, 1991; Lehning et al., 1999; Vionnet et al., 2012; Leroux and Pomeroy, 2017); iii) intermediate-complexity schemes with physics from the detailed schemes but with a limited amount of layers, which are intended for coupling with atmospheric models (e.g., Sun et al., 1999; Boone and Etchevers, 2001). The intercomparison results of the abovementioned snow models at an alpine site indicated that all three types of schemes are capable of representing the basic features of the snow cover over the 2-year period but behaved differently on shorter timescales. Furthermore, Snow Model Intercomparison Project (SnowMIP) at two mountainous alpine sites revealed that the albedo parameterization was the major factor influencing the simulation of net shortwave radiation~~,~~. ~~which wa~~Though this parameterization is independent of model complexity (Etchevers et al., 2004) ~~but~~ it directly affects the ~~directly~~ snow simulation~~s~~. SnowMIP2 evaluated thirty-three snowpack models across a wide range of hydrometeorological and forest canopy conditions~~,~~. It identified the shortcomings of different snow models and highlighted the necessity of studying the separate contribution of individual components to the mass and energy balance of snowpack (Rutter et al., 2009).

With the majority of research focuses on the intercomparison of the snowpack models with various physical ~~complexity~~complexities, little attention has been paid to the treatment of the underlying soil physical processes (see the brief overview of ~~the~~ current soil-snow modelling efforts in Table 1).

In this paper, one of the widely used snowpack models (Utah energy balance snowpack model, UEB, Tarboton and Luce, 1996) was incorporated into a common soil modeling framework (Simultaneous Transfer of Energy, Mass and Momentum in Unsaturated Soils with Freeze-Thaw, STEMMUS-FT, Zeng et al., 2011a, b; Zeng and Su, 2013; Yu et al., 2018~~),~~). The new model is named STEMMUS-UEB and is configured with various levels of model complexity~~,~~ in terms of mass and energy transport physics. We utilized in situ observations and numerical experiments with STEMMUS-UEB to investigate the effect of snowpack on the underlying soil mass and energy transfer with different complexities of soil models. The description of the coupled soil-snow modeling framework STEMMUS-UEB and the model setup for this study are presented in Section 2. Section 3 verified the proposed model and identified the effect of snowpack on soil liquid/vapor fluxes. The uncertainties and limitations of this study~~,~~ and the ~~benefits we can obtain from~~applicability of the proposed model ~~(i.e., the effect of snowpack on the coupled mass and energy transport in the soil) is~~are discussed in Section 4.

2 Description of Coupled Soil-Snow Modelling Framework and Model Setup

2.1 Soil mass and heat transfer model

The detailed physically based two-phase flow soil model (STEMMUS) was first developed to investigate the underlying physics of soil water, vapor, and dry air transfer mechanisms and their interaction with the atmosphere (Zeng et al., 2011a, b; Zeng and Su, 2013). It is ~~realized~~ achieved by simultaneously solving the balance equations of soil mass, energy, and dry air in a fully coupled way. The mediation effect of vegetation on such interaction was latterly incorporated via the root water uptake sub-module (Yu et al., 2016) and furthermore by coupling with the detailed soil and vegetation biogeochemical process (Wang et al., ~~2020~~2021; Yu et al., 2020a). ~~Implementing the freeze-thaw process (hereafter STEMMUS-FT, for applications in cold regions), it~~ It facilitates our understanding of the hydrothermal dynamics of respective components in the frozen soil medium (i.e., soil liquid water, water vapor, dry air, and ice) by implementing the freeze-thaw process (hereafter STEMMUS-FT, for applications in cold regions, ~~(~~Yu et al., 2018; Yu et al., 2020c).

The frozen soil physics considered in STEMMUS-FT include three parts: i) the ice blocking effect on soil hydraulic conductivities (see Supplement Sect. 2.2.2); ii) the inclusion of ice effect in the calculation of soil thermal capacity/conductivity (see Supplement Sect. 2.2.8); iii) the exchange of latent heat flux during phase change periods. With the aid of Clausius Clapeyron relation, which characterizes the phase transition between liquid and solid phase in the thermal equilibrium system. The soil water characteristic curve (e.g., van Genuchten, 1980) is then extended to consider the freezing temperature dependence, i.e., soil freezing characteristic curve (Hansson et al., 2004; Dall’Amico et al. 2011). The fraction of soil liquid/solid water at a given temperature was then calculated prognostically with the soil freezing characteristic curve. Soil hydraulic parameters were further used in the Mualem (1976) model to compute the soil hydraulic conductivity. The ice effect is considered by reducing the soil saturated hydraulic conductivity as the function of ice content (Yu et al., 2018).

In response to ~~minimizing~~ minimize the potential model-comparison uncertainties ~~rising~~ from various model structures (Clark et al., 2015) and to figure out which process matters, three levels of complexity of mass and heat transfer physics are made available in the current STEMMUS-FT modelling framework (Yu et al., 2020c). First, the 1-D Richards equation and heat conduction were deployed in STEMMUS-FT to describe the isothermal water flow and heat flow (termed BCD). ~~In t~~The BCD model~~,~~ considers the interaction of soil water and heat transfer ~~is only~~ implicitly via the parameterization of heat capacity, thermal conductivity, and the water phase change effect. ~~For the advanced coupled water and heat transfer (ACD model), t~~The water flow is fully affected by soil temperature regimes in the advanced coupled water and heat transfer model (termed ACD model). The movement of water vapor, as the linkage between soil water and heat flow, is explicitly characterized. STEMMUS-FT further enables the simulation of temporal dynamics of three water phases (liquid, vapor, and ice), together with the soil dry air component (termed ACD-air model). The governing equations of liquid water flow, vapor flow, air~~-~~flow, and heat flow were listed in Appendix A.1 (see the more detailed model description in Zeng et al., 2011a, b; Zeng and Su, 2013; Yu et al., 2018; Yu et al., 2020c).

2.2 Snowpack module UEB

The Utah energy balance (UEB) snowpack model (Tarboton and Luce, 1996) is a ~~single-layer~~single-layer physically~~-~~-based snow accumulation and melt model. Two precipitation types, i.e., rainfall and snowfall, are discriminated by its dependence on air temperature. The snowpack is characterized using two primary state variables, snow water equivalent *SWE* and the internal energy *U*. Snowpack temperature is expressed diagnostically as the function of *SWE* and *U*, together with the states of the snowpack (i.e., solid, solid and liquid mixture, and liquid). Given the insulation effect of the snowpack, snow surface temperature differs from the snowpack bulk temperature, which is mathematically considered using the equilibrium method (i.e., balances energy fluxes at the snow surface). The age of the snow surface, as the auxiliary state variable, is utilized to calculate the snow albedo (see Appendix A.3). The melt outflow is calculated using Darcy's law with the liquid fraction as inputs. The conservation of mass and energy~~, as presented in Appendix A.2,~~ forms the physical basis of UEB (Tarboton and Luce, 1996, as presented in Appendix A.2).

UEB is recognized as one simple yet physically-based snowmelt model~~,~~. ~~which can~~It captures the ~~first~~ first-order snow process (e.g., diurnal variation of meltwater outflow rate, snow accumulation, and ablation, see a general overview of UEB model development and applications in Table S3). It requires little effort in parameter calibration and can be easily trans~~port~~ferable and applicable to various locations (e.g., Gardiner et al., 1998; Schulz and de Jong, 2004; Watson et al., 2006; Sultana et al., 2014; Pimentel et al., 2015; Gichamo and Tarboton, 2019), especially for data scarce regions as for example Tibetan Plateau. We thus selected the original parsimonious UEB (Tarboton and Luce, 1996) as the snow module to be coupled with the soil module (STEMMUS-FT).

2.3 Coupling procedure

The coupled process between the snowpack model (UEB) and the soil water model (STEMMUS-FT) was illustrated in Figure 1. The one-way sequential coupling is employed to couple the soil model with the current snowpack model. The role of the snowpack is explicitly considered by altering the water and heat flow of the underlying soil. The snowpack model takes the atmospheric forcing as the input (precipitation, air temperature, wind speed and direction, relative humidity, shortwave and longwave radiation) and solves the snowpack energy and mass balance (Eq. A.8 & A.9, Subroutines: ALBEDO, PARTSNOW, PREDICORR), provides the melt water flux and heat flux as the surface boundary conditions for the soil model STEMMUS-FT (Subroutines: h_sub and Enrgy_sub for ACD models; Diff_Moisture_Heat for BCD model). STEMMUS-FT then solves the energy and mass balance equations of soil layers in one time_step. To highlight the effect of the snowpack on the soil water and vapor transfer process, we constrained the soil surface energy boundary as the Dirichlet type condition (take the specific soil temperature as the surface boundary condition). Surface soil temperature was derived from the soil profile measurements and was not permitted to be higher than zero when there is snowpack. To ensure ~~the~~ numerical convergence, the adapted time_step strategy was used. ~~The h~~Half-hourly meteorological forcing measurements were linearly interpolated to the running timesteps (Subroutine Forcing_PARM). The precipitation rate (validated at 3-hour time intervals) was regarded uniformly within the 3-hour duration (~~see~~ refer to Table S1 for detail). The general description of the ~~main~~ primary subroutines in STEMMUS-UEB was presented in Table 2.~~,~~ It including the main functions, input/output, and ~~its~~ their connection with other subroutines~~, was presented in Table 2~~ (linked with Table S1 and S2 for the description of model input parameters and outputs for this study, see the detailed ~~general~~ description in Tarboton and Luce, 1996; Zeng and Su, 2013; Yu et al., 2018).

2.4 Configurations of numerical experiments

On the basis of the aforementioned STEMMUS-UEB coupling framework, the various ~~complexity~~ complexities of vadose zone physics ~~was~~ were further implemented as three alternative model versions. First, the soil ice effect on soil hydraulic and thermal properties, and the heat flow due to the water phase change were taken into account, while the water and heat transfer is not coupled in STEMMUS-FT and termed the BCD model. Second, the STEMMUS-FT with the fully coupled water and heat transfer physics (i.e., water vapor flow and thermal effect on water flow) was applied and termed the ACD model. Lastly, on top of the ACD model, the air pressure was independently considered as a state variable (therefore, the airflow) and termed the ACD-air model. With the abovementioned model versions (STEMMUS-FT_Snow) and taking into account the no-snow scenarios (STEMMUS-FT_No-Snow), Table 3 lists the configurations of all six designed numerical experiments. The model parameters used for all simulations for the tested experimental site are listed in Table S2.

2.5 Description of the Tested Experimental site

Maqu station, equipped with a catchment scale soil moisture and soil temperature (SMST) monitoring network and micro-meteorological observing system, is situated on the north-eastern edge of the Tibetan Plateau (Su et al., 2011; Dente et al., 2012; Zeng et al., 2016). According to the updated Köppen-Geiger climate Classification System, it can be characterized as a cold climate with dry winter and warm summer. The average annual air temperature is 1.2 ℃, and the mean air temperatures of the coldest month (January) and the warmest month (July) are about -10.0 ℃ and 11.7 ℃, respectively. Alpine meadows (e.g., *Cyperaceae* and *Gramineae*), with a height varying from 5 cm to 15 cm throughout the growing season, are the dominant land cover in this region. The general soil types are sandy loam, silt loam and organic soil for the upper soil layers (Dente et al., 2012; Zheng et al., 2015; Zhao et al., 2018). The soil texture and hydraulic properties were listed in Table S2 and how it was used in STEMMUS-UEB is illustrated in Figure 1 and Table 2.

The Maqu SMST monitoring network spans an area of approximately 40 km×80 km with the elevation ranging from 3200 m to 4200 m a.s.l. (33°30'–34°15'N, 101°38'–102°45'E). SMST profiles are automatically measured by 5TM ECH_2O probes (METER Group, Inc., USA) installed at different soil depths, i.e., 5 cm, 10 cm, 20 cm, 40 cm, and 80 cm. The micro-meteorological observing system consists of a 20 m Planetary Boundary Layer (PBL) tower providing the meteorological measurements at five heights above ground (i.e., wind speed and direction, air temperature and relative humidity), and an eddy-covariance system (EC150, Campbell Scientific, Inc., USA) equipped for measuring the turbulent sensible and latent heat fluxes and carbon fluxes. The equipment for four-component down and upwelling solar and thermal radiation (NR01-L, Campbell Scientific, Inc., USA), and liquid precipitation (T200B, Geonor, Inc., USA) are also deployed. The dataset from December 1, 2015 to March 15, 2016 was utilized in this study. An independent precipitation data (3-hour time interval) during the same testing period from an adjacent meteorological station was used as the mutual validation data.

3. Results: ~~Comparison~~ comparison of simulation results of surface variables with/without snowpack effect

3.1 Albedo

The time series of surface albedo, calculated as the ratio of upwelling shortwave radiation to the downwelling shortwave radiation and estimated using BCD, ACD and ACD-air models, was shown in Figure 2 together with precipitation. As the snowpack has a higher albedo than the underlying surface (e.g., soil, vegetation), compared to the observations, models without snow module presented a relatively flat variation of daily average surface albedo, and lacked the response to the winter precipitation events (Figure 2, Table 4). With the snow module, STEMMUS-UEB models can capture mostly the abrupt increase of surface albedo after winter precipitation events. The mismatches in terms of the magnitude or absence of increased albedo after precipitation events indicated that the model tended to underestimate the ~~dynamics of~~ albedo dynamics. ~~and~~ ~~t~~The shallow snowfall events might be not well captured by the model (see ~~the~~ Sect. 4.1). Three model versions (BCD-Snow, ACD-Snow, and ACD-air-Snow) produced similar fluctuations regarding the presence of snow cover with slight differences in terms of the magnitude of albedo.

3.2 Soil Temperature and Moisture Dynamics

The observed spatial and temporal dynamics of soil temperature from five soil layers was used to verify the performance of different models (Fig. 3). The initial soil temperature state can be characterized as the warm bottom and cool surface soil layers (based on in-situ observations). The freezing front (indicated by the ~~zero degree~~zero-degree isothermal line, ZDIL) developed downwards rapidly until~~l~~ the 70th day after December 1, 2015, reaching its maximum depth. Then the freezing front stabilized as the offset effect of latent heat release (termed as zero-curtain effect). Such ~~effect~~ influence can sustain until all the available water to that layer is frozen, at which point the latent heat effect is negligible compared to the heat conduction. At shallower layers, the atmospheric forcing dominates the fluctuation of thermal states. The isothermal lines (e.g., -2 ^{o}C) had a larger variation than that of ZDIL. At deeper soil layers, the temporal dynamics of isothermal lines were smoother than that of ZDIL, indicating that the effect of fluctuated atmospheric force on soil temperature was damped with the increase of soil depth. Compared to the observations, BCD-Snow model presented an earlier development of the freezing front and arrival of the maximum freezing depth (60th day after December 1, 2015). The deeper and more fluctuated freezing front indicates that a stronger control of atmospheric forcing on soil thermal states was produced by BCD-Snow model. The ACD models can well capture the propagation characteristic of the freezing front in terms of the variation magnitude and maximum freezing depth. There is no significant difference in soil thermal dynamics between the model with and without snow module, except at the surface soil layers (Table 4).

Figure 4 shows the spatial and temporal dynamics of observed and simulated soil water content in the liquid phase (SWCL). The SWCL of active layers ~~is~~ depends to a large~~ly~~ ~~dependent~~extent on the soil freezing/thawing status. Soil is relatively wet at soil layers of 10-60 cm for the starting period. Its temporal development was disrupted by the presence of soil ice and tended to increase wetness during the thawing period. A relatively dry zone ($\theta_L < 0.06\ m^3\ m^{-3}$) above the freezing front was found, indicating the nearly completely frozen soil during the stabilization stage. The initial wet zone of soil moisture was narrowed down and the rewetting zone tended to enlarge from BCD-Snow simulation~~s~~ due to its early freezing and thawing of soil (Fig. 4b). The position of the dry zone occurred earlier as the early reaching of the stabilization period by the BCD-Snow model (Fig. 3b). For the ACD models, the position and development of initial wet zone, rewetting zone and the dry zone is similar to that from the observations, indicating the soil moisture dynamics can be well captured by the ACD models. Compared to the STEMMUS-FT_Snow model, there was no observable difference in the SWCL dynamics at deeper soil layers from STEMMUS-FT_No-Snow simulations. The surface SWCL was found affected from STEMMUS-FT_Snow simulations (Table 4).

3.3 Surface Latent Heat Flux

Figure 5 shows the comparison of time series of observed and model simulated surface cumulative latent heat flux using three models with/without consideration of snow module. Considerable overestimation of latent heat flux was produced by the BCD-Snow model, with 121.79% more than observed. Such overestimations were largely reduced by ACD and ACD-air models. There is a slight underestimation of cumulative latent heat flux by ACD-Snow and ACD-air-Snow models, with -8.33% and -7.05%, respectively. Compared with STEMMUS-FT_Snow simulations, there is less latent heat flux produced by STEMMUS-FT_No-snow simulations. It is mainly due to the sublimation of snow cover, which cannot be simulated by the STEMMUS-FT_No-snow models. The difference in cumulative latent heat flux between STEMMUS-FT with and without snow module increases from BCD to ACD-air schemes, with the values of 2.02%, 7.69%, and 8.97% for BCD, ACD and ACD-air schemes, respectively.

3.4 Liquid/vapor fluxes

To further elaborate the effect of snowpack on LE, we presented the diurnal variations of LE and its components at two typical episodes with precipitation events (freezing and thawing period, respectively). The relative contribution of liquid and vapor flow to the total mass transfer after precipitation events was separately presented in Figure ~~6~~ 8 & ~~7~~9, i.e., the liquid water flux driven by temperature q_{LT}, matric potential q_{Lh} and air pressure q_{La}, water vapor flux driven by temperature q_{VT}, matric potential q_{Vh} and air pressure q_{Va}.

~~3.4.1 Freezing period~~1) LE

Diurnal dynamics of the observed and simulated latent heat flux during the rapid freezing period with the occurrence of precipitation events, from 10th to 14th Days after Dec. 1. 2015, is shown as Fig. 6a, ~~d~~ b &~~g~~c. Compared to the observations, the diurnal variations of latent heat flux ~~was~~were captured by the proposed model with various levels of complexities. Performance of BCD, ACD, and ACD-air models in simulating LE differed mainly regarding the magnitude and response to precipitation events. For the BCD-Snow model, the overestimation of LE was found at 10th and 11th day after December 1 due to relatively high surface soil moisture simulation~~s~~ (Fig. S1b). A certain amount of enhanced surface evaporation was produced shortly after precipitation, which is most probably due to the snow sublimation~~,~~. ~~which~~Snow sublimation~~-~~ presents ~~in the model simulations while~~ appear not intuitively matching with~~in~~ observations. The mismatch in the LE enhancement after precipitation events can be attributed to that the partition process of precipitation into various components (rainfall, snowfall, canopy interception) might not be well captured by the model. Such a response to the winter precipitation events was absent from the BCD-No-Snow simulations.

The overestimation of LE was reduced by ACD and ACD-air models (Fig. ~~6d~~ 6b & ~~g~~c). Compared to the ACD-Snow model simulations, ACD-No-snow model produced a stronger diurnal variation of LE after the ~~precipitation, and~~precipitation and is more approaching ~~to~~ the measured LE. ~~The l~~Lower diurnal variation of LE for the ACD-Snow model can be ascribed to the lower surface SWCL (see Fig. S1d & g). For the ACD-Snow model, precipitation was partitioned into rainfall and snowfall, part of which was directly evaporated as sublimation. The sum of rainfall and the melting part of snowfall reached the soil surface as the incoming water flux, which is less than that for the ACD-No-snow model (took all the precipitation as the incoming water flux). There is no significant difference in the dynamics of LE between simulations by ACD models and ACD-air models.

During the thawing period, the diurnal variations of LE were well simulated by the models (Figure 7). There are some discrepancies regarding the peak values of LE. For the BCD-Snow model, overestimations were found in 10~~1~~0~~st,~~th, 101st, and 102nd ~~, and 103rd~~ day after December 1, 2015. The high LE values on 100th~~101st~~ and 101st ~~102nd~~ day are probably due to the high surface soil moisture by the thawing water (Fig. S2b). ~~w~~While on the 102nd~~103rd~~ day, it is due to the snow sublimation (Fig. 7a). The peak values were reproduced but shifted by BCD-No-Snow simulation~~s~~, which occurred on 100th and at the end of 102nd, indicating the shift of surface soil moisture states (Fig. S2b).

For the ACD model, the difference in latent heat flux between snow and no-snow simulations was noticeable two days after precipitation. The larger values of LE from the ACD-No-~~S~~snow model occurred earlier than that from the ACD-~~No-s~~Snow model, as the earlier response of surface soil moisture to the precipitation event (Fig. S2). While compared to the observations, the enhancement of LE advanced from the ACD-Snow simulations (Fig. 7~~d~~b). This enhanced evaporation can be attributed to the snow sublimation and increased surface soil moisture content. Similar lag behavior of precipitation-enhanced evaporation was produced by the ACD-air-Snow models (Figure 7c). There are mismatches in the time and magnitude of LE enhancement between ACD-Snow model simulations and observations (Fig. 7b). This discrepancy lies in the uncertainties of snowpack simulations, which can be ~~due~~attributed to either the inaccurate precipitation measurements (Barrere et al., 2017; Günther et al., 2019) or that the precipitation partition process is not well described by the model (Harder and Pomeroy, 2014; Ding et al., 2017).

~~For ACD-Snow model, the precipitation induced evaporation enhancement was lagged compared to ACD-No-Snow model. This enhanced evaporation can be attributed to the snow sublimation and increased surface soil moisture content. The similar lag behavior of precipitation-enhanced evaporation was produced by the ACD-air-Snow models.~~

2) LE and decomposition of surface mass transfer~~3.4.2 Thawing period~~

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During the freezing period, the soil water vapor, instead of liquid water flux, dominated the surface mass transfer process. Missing the description of the vapor diffusion process hindered the BCD models to realistically depict the decomposition of surface mass transfer dynamics (Fig. 8a &b).

There is a visible diurnal variation of thermal vapor flux q_{VT} from~~From~~ the ACD model simulation (Fig. 8c &d~~,~~). ~~there is a visible diurnal variation of thermal vapor flux~~ ~~q_{VT}~~ ~~. It can be clearly identified that t~~The isothermal vapor flux q_{Vh} contributed to most of the mass transfer during the freezing period. ~~After winter precipitation during the nighttime, there is a certain amount of isothermal vapor flux driven by the downward matric potential gradient. The reason is that the precipitation water immediately freezes on the soil surface.~~ It should be noted that the sum of water/vapor fluxes at 0.1cm soil layer cannot balance the surface evaporation, especially after the precipitation events (Fig. ~~6e~~8c). We assumed and attributed it to the surface ice sublimation process. Precipitation water was frozen on the soil surface, and only vapor fluxes are active~~exists~~ in the topsoil layers. Sublimation of surface ice may contribute to the gaps between liquid/vapor fluxes and LE (Yu et al., 2018). As more precipitation water was frozen on the soil surface from the ACD-No-Snow model (Fig. 8d), the difference between the sum of water/vapor fluxes at the top 0.1cm soil layer and the surface evaporative water enlarged compared to ACD-Snow simulations ~~(Fig. 6f)~~. Thermal liquid water flux q_{LT} appears negligible to the total mass flux during the whole simulation period. There is no significant difference recognized in the mass transfer between the ACD-air and ACD during the freezing period.

~~3.4.2 Thawing period~~

~~During the thawing period, the diurnal variations of LE were well simulated by the models. There are some discrepancies regarding the peak values of LE. For BCD-Snow model, overestimations were found in 101st, 102nd, and 103rd day after December 1, 2015. The high LE values on 101st and 102nd day are probably due to the thawing water (Fig. S2b) while on 103rd day it is the snow sublimation (Fig. 7a). The peak values were reproduced but shifted by BCD-No-Snow simulations, which occurred on 100th and at the end of 102nd, indicating the shift of surface soil moisture states (Fig. S2b).~~

~~For the ACD model, the difference in latent heat flux between snow and no-snow simulations was noticeable two days after precipitation. The larger values of LE from ACD-Snow model occurred earlier than that from ACD-No-snow model, as the earlier response of surface soil moisture to the precipitation event (Fig. S2). While compared to the observations, the enhancement of LE advanced from the ACD-Snow simulations (Fig. 7d). There are mismatches in the time and magnitude of LE enhancement between ACD-Snow model simulations and observations. This discrepancy lies in the uncertainties of snowpack simulations, which can be due to either the inaccurate precipitation measurements (Barrere et al., 2017; Günther et al., 2019) or that the precipitation partition process is not well described by the model (Harder and Pomeroy, 2014; Ding et al., 2017).~~

~~For ACD-Snow model, the precipitation induced evaporation enhancement was lagged compared to ACD-No-Snow model. This enhanced evaporation can be attributed to the snow sublimation and increased surface soil moisture content. The similar lag behavior of precipitation-enhanced evaporation was produced by the ACD-air-Snow models.~~

During the thawing period, a certain amount of upward liquid water flux was produced by the BCD model, supplying the water to the topsoil and evaporate into the atmosphere (Fig. ~~7b &c~~9a &b). Compared to the isothermal liquid flux q_{Lh}, the thermal liquid flux q_{LT} was negligible to the total mass flux.

For the ACD model, the diurnal variation of thermal vapor flux q_{VT} was enhanced after precipitation, producing a larger amount of upward/downward vapor flux during the night/~~day time~~daytime (e.g., Fig. ~~7e~~9c). As the surface soil is relatively dry, the isothermal vapor flux q_{Vh} contributes nearly ~~all of~~all the mass flux during the selected thawing period. Driven by the ~~large downward~~ matric potential gradient, a large amount of isothermal water vapor flux q_{Vh}, accompanied by downward liquid water flux q_{Lh}, can be found after the nighttime precipitation event (Fig. 9c, d, e, f). ~~This~~ These ~~precipitation~~ precipitation-induced isothermal liquid/vapor fluxes ~~was~~ were lagged and less intense from the ACD-Snow model than that from the ACD-No-Snow model simulation (e.g., Fig. 9c vs. Fig. 9d). It is explained that ~~T~~the snowpack reduces the instant ~~liquid phase of~~ precipitation infiltration process and enables the snowmelt ~~will occur~~ afterwards~~. It resulted in~~, which led to the lagged and weaker response of surface SWCL to the precipitation (Fig. S2). It breaks the balance between isothermal vapor flux and evaporative LE (around 103rd day after Dec. 1, 2015). Compared to the ACD-No-Snow model, such imbalance was enlarged for the ACD-Snow model during the thawing period (Fig. ~~7e~~ 9c &~~f~~d).

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Compared to the ACD-No-Snow simulations, the upward thermal vapor flux q_{VT} was enhanced after precipitation for the ACD-air-No-Snow model (Fig. ~~7i~~9f). This enhanced upward vapor flux reduced the soil liquid water content at 0.1cm (Fig. S2f~~), and~~) and decreased the soil hydraulic conductivity and then the downward isothermal liquid/vapor flux (q_{Lh}, q_{Vh}). Other than that, there is no significant difference between the ACD-air model and the ACD model during the thawing period.

4. Discussion

4.1 Uncertainties ~~of~~ in simulations of surface albedo ~~simulations~~ and ~~Limitations~~limitations

After a winter precipitation event, land surface albedo increases considerably (Fig. 2), indicating the ~~formation~~ presence of the snowpack. ~~While~~ However, such snowfall events were ~~isolated~~ episodic with small magnitude, which ~~are~~ is difficult to be well captured. Such difficulties can be partially attributed to the inherent uncertainties in ~~the representativity of~~ precipitation measurements (both the precipitation amount and types). Due to the spatial variability of precipitation, the accurate observation of winter precipitation is proved to be a challenge, especially during windy winters (Barrere et al., 2017; Pan et al., 2017). It is necessary to have more snowpack-relevant measurements (e.g., the high-resolution measurements of the spatiotemporal field of wind speed, precipitation, and snowpack variations) to understand the dynamics of snowpack and its effect on energy and water fluxes. ~~On the other hand~~Furthermore, the temporal resolution of precipitation measurements adopted in this study is relatively coarse (3-hour). In the current precipitation partition parameterization, the amount of snowfall was determined as a function of precipitation and air temperature thresholds. Given the coarse temporal resolution of precipitation measurements, the model may produce a time shift of snowfall events or even the mal-identification of snowfall~~ taken into account the effect of air temperature~~. The simple relation between the air temperature and precipitation types may be not suitable to this region, because air temperature is not the best indicator of precipitation types, as argued by Ding et al. (2014). Other factors, i.e., relative humidity, surface elevation, and wet-bulb temperature, are also very relevant and should be taken into account for the discrimination of precipitation types. The other uncertainty lies in the representation of the snow process. For example, the wind-blow effect and canopy snow interception, which have been recognized as important to the accurate simulation of snowpack dynamics (Mahat and Tarboton, 2014), are not taken into account in detail. Last but not least, the interpretation of surface albedo dynamics needs to be adapted to the specific site, especially regarding the shallow snow situations (Ueno et al., 2007; Ueno et al., 2012; Ding et al., 2017; Wang et al., 2017). The albedo of the underlying surface should also be properly accommodated to this Tibetan meadow system. Regardless of the aforementioned uncertainties, our proposed model was capable to capture the surface albedo variations with precipitation (Fig. 2) and can be seen as acceptable to ~~conduct the~~ analy~~sis~~ze ~~of~~ snow cover effects in such a harsh ~~condition~~environment.

4.2 ~~Benefits from STEMMUS-UEB: The effect of s~~Snow cover-induced evaporation enhancement

Different from the rainfall, precipitation water from snowfall enters the soil considerably lagged in time due to the water storage by~~ the~~ snow cover (You et al., 2019). With ~~consideration of~~the snow module, precipitation was partitioned into rainfall and snowfall. Part of the snowfall evaporated into the atmosphere as sublimation and the other part together with the rainfall infiltrated into the underlying soil. It resulted in the delay of incoming water to the soil with a less amount compared to that without consideration of the snow module~~ (Fig. 8 e &f)~~. This amount of incoming water increased the evaporation after precipitation (Fig. ~~8 d~~6 & 7). The other source for the enhanced evaporation flux after precipitation is snow sublimation, which is absent from the model without the snow module. Sublimation~~, although not easy to observe,~~ occurs readily under certain weather conditions (e.g., with freezing temperatures, enough energy). It can be ~~further sped up~~more active ~~at~~ in regions with low relative humidity, low air pressure and dry winds. Such amount of sublimation has been reported important from the perspective of climate and hydrology (e.g., Strasser et al., 2008; Jambon-Puillet et al., 2018), especially at high altitude regions with the low air pressure. During the freezing period, the evaporation enhancement can be also sourced from the sublimation of surface ice. The amount of the ice sublimation appeared to ~~be~~ decrease~~d~~ during the freezing period ~~as~~ in the presence of a transient snowpack (e.g., Fig. 8c vs. 8d~~6~~). This is consistent with the results of Hagedorn et al. (2007), who investigated the effect of snow cover on the mass balance of ground ice with an artificially continuous annual snow cover. ~~Their~~ According to their results, ~~indicated that~~the snow cover enhanced the vapor transfer into the soil and thus reduced the ~~long~~ long-term ice sublimation. The relative contribution of increased surface soil moisture, snow sublimation, and surface ice sublimation to the enhanced evaporation is dependent on the pre-precipitation soil moisture/temperature states, air temperature, and the time and magnitude of precipitation events. Under the conditions of the low pre-precipitation SWCL ~~under~~ with the freezing soil temperature (e.g., Fig. ~~6e~~8e, 11th vs. 12th Days after 1 December), the precipitation falls ~~down~~ on the surface as snowfall and rainfall (most~~ly~~ freezes as ice). The sublimation from surface ice can contribute to most of the total mass transfer (e.g., Fig. ~~6e~~8e, 11th Days after 1 December). If the soil temperature rises above the freezing temperature, there will be no sublimation of surface ice, in terms of contributing to the enhanced evaporation (e.g., Fig. 9e, 102nd Days after 1 December). ~~.~~

4.3 ~~Benefits from STEMMUS-UEB: Responses among different complexity of soil model~~Snow cover impacts with different soil model complexities

The model with ~~various~~ different complexity of soil mass and energy transfer physics behaves differently in response to the winter precipitation events. During the freezing period, there is no significant difference in the BCD model simulated soil moisture with/without the snow module. The precipitation water freezes at the soil surface, which cannot be transferred downwards with the BCD model physics. The sublimation, from either the snow or the surface ice, contributes all to the precipitation-enhanced evaporation for the BCD model. As with ~~consideration of~~ vapor flow, the surface ice increases the soil moisture at lower layers via the downward isothermal vapor flux (Fig. ~~6~~8). The surface ice sublimation and increased soil moisture-induced evaporation enhancement can be ~~clearly~~ identified from the ACD model simulation~~s~~. The role of air flow was negligible to the mass transfer during the freezing period.

When it comes to the thawing period, BCD model produced a certain amount of liquid water flow, contributing considerably to the mass transfer. The obvious fluctuation of SWCL was noticed due to the thawing water and precipitation event. The main source for the increased evaporation was interpreted as isothermal liquid water flow. While for the ACD model, the situation becomes more complex. Thawing surface ice and snowmelt water may coexist at the soil surface, resulting in different soil moisture response to precipitation events. The ice sublimation, snow sublimation, and increased soil moisture ~~all~~ contribute to the evaporation enhancement after precipitation. When considering air flow, dry air interacts with soil ice, liquid/vapor water in soil pores (Yu et al., 2018) and alters the soil moisture states. It thus considerably changes the relative contribution of each component to the mass transfer (Fig. ~~7~~9).

5. Conclusions

~~Rendering from~~With the aim to investigate the hydrothermal effect of the snowpack on the underlying soil system, we developed the integrated process-based soil-snow-atmosphere model, STEMMUS-UEB v1.0.0, which is ~~dedicated for~~based on the easily ~~transportable~~ transferable and physically-based description of the snowpack process and the detailed interpretation of the soil physical process with various complexities. From STEMMUS-UEB simulations, snowpack affects not only the soil surface conditions (surface ice and SWCL~~),~~) and energy-related states (albedo, latent heat flux)~~,~~ but also the transfer patterns of subsurface soil liquid/vapor flow. ~~With consideration of the snow module,~~ STEMMUS-FT model can capture mostly the abrupt increase of surface albedo after winter precipitation events with consideration of the snow module. There is a significant overestimation of cumulative surface latent heat flux by the BCD model. ACD and ACD-air model produces a slight underestimation of cumulative LE compared to the observations. ~~Given no~~Without sublimation from snowpack, there is a less latent heat flux produced by STEMMUS-FT_No-snow simulations compared with ~~STEMMUS-FT_S~~snow simulations. The presence of snowpack alters the partition process of precipitation and thus the surface SWCL. BCD models with/without snowpack produced ~~the~~ similar surface SWCL during the freezing period while resulted in the abrupt increase of soil moisture in response to the precipitation during the thawing period. ACD-Snow model simulated a less intensive and lagged soil moisture variation in response to precipitation compared to the ACD-No-Snow model during both the freezing and thawing period, respectively. ACD-air model ~~affects~~ affected the intensity of increased surface soil moisture, especially during the thawing period.

Three mechanisms, surface ice sublimation, snow sublimation and increased soil moisture, can contribute to ~~the~~ enhanced latent heat flux after winter precipitation events. The relative role of each mechanism in the total mass transfer can be affected by the time and magnitude of precipitation and pre-precipitation soil moisture/temperature states (see Sect. 4.3). The simple BCD model cannot provide a realistic partitioning of mass transfer. ACD model, with consideration of vapor diffusion and thermal effect on water flow and snowpack can produce a reasonable analysis of the relative contributions of different water flux components. With consideration of air flow, the relative contribution of each component to the mass transfer was ~~considerably~~ substantially altered during the thawing period. Further work will take into account the thermal interactive effects between snowpack and the underlying soil. Such work will inevitably enhance our confidence in interpreting the underlying mechanisms and physically elaborating on the role of snowpack in cold regions.

Code and data availability. The coupled Soil-Snow model (STEMMUS-UEB v1.0.0) with three levels of complexity of soil water and heat transfer physics was developed based on STEMMUS-FT (Simultaneous Transfer of Energy, Momentum and Mass in Unsaturated Soils with Freeze and Thaw) and UEB (Utah Energy Balance) model. The original STEMMUS source code is available from the GitHub website via https://github.com/yijianzeng/STEMMUS. The snowmelt module is based on the code of (Tarboton and Luce, 1996). The coupled STEMMUS-UEB v1.0.0 code is archived on Zenodo (Yu et al., 2020b), licensed under the Apache License, Version 2.0. The current code is tested by MATLAB 2019b using an Intel Core i7 processor (Intel® Core™ i7-6700HQ CPU @ 2.60GHz 2.59 GHz), an installed memory (RAM, 16.0 GB), and a 64-bit Windows 10 Enterprise operating system. The relevant data can be accessed from 4TU. Center for Research Data (https://doi.org/10.4121/uuid:cc69b7f2-2448-4379-b638-09327012ce9b; https://doi.org/10.4121/uuid:61db65b1-b2aa-4ada-b41e-61ef70e57e4a).

Author contribution. ZS, YZ, and LY designed and conceptualized this study; YZ and ZS provided the original version of STEMMUS model code and supervised the further modelling development; LY developed the STEMMUS-UEB model coupling framework with the contribution from YZ; LY and YZ prepared the original draft of the paper, LY, YZ, and ZS all contributed to the reviewing and editing of the final paper.

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

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Appendix A

A.1 STEMMUS-FT model with three levels of complexity

A.1.1 Uncoupled soil water and heat transfer physics

The Richard equation which describes the water flow under gravity and capillary forces in isothermal conditions, is solved for variably saturated soils.

$$\frac{\partial\theta}{\partial t} = -\frac{\partial q}{\partial z} - S = \rho_L \frac{\partial}{\partial z}\left[K\left(\frac{\partial\psi}{\partial z} + 1\right)\right] - S \tag{A.1}$$

where θ (m^3 m^{-3}) is the volumetric water content; q (kg m^{-2} s^{-1}) is the water flux; z (m) is the vertical direction coordinate (positive upwards); S (s^{-1}) is the sink term for root water uptake; ρ_L (kg m^{-3}) is the soil liquid water density; K (m s^{-1}) is the soil hydraulic conductivity; ψ (m) is the soil water potential; t (s) is the time.

The heat conservation equation, considering the latent heat due to water phase change, can be expressed as:

$$C_{soil}\frac{\partial T}{\partial t} - \rho_i L_f \frac{\partial\theta_i}{\partial t} = \frac{\partial}{\partial z}\left(\lambda_{eff}\frac{\partial T}{\partial z}\right) \tag{A.2}$$

where C_{soil} (J kg^{-1} °C^{-1}) is the specific heat capacity of bulk soil; T (°C) is the soil temperature; ρ_i (kg m^{-3}) is the density of soil ice; L_f (J kg^{-1}) is the latent heat of fusion; θ_i (m^3 m^{-3}) is the soil ice volumetric water content. λ_{eff} (W m^{-1} °C^{-1}) is the effective thermal conductivity of the soil.

A.1.2 Coupled water and heat transfer

For the coupled water and heat transfer physics, the liquid water flow is non-isothermal and affected by soil temperature regimes. The movement of water vapor, as the linkage between soil water and heat flow, is explicitly characterized. With modifications made by Milly (1982), the extended version of Richards (1931) equation with consideration of the liquid and vapor flow is written as:

$$\begin{aligned}&\frac{\partial}{\partial t}(\rho_L\theta_L + \rho_V\theta_V + \rho_i\theta_i) = -\frac{\partial}{\partial z}(q_L + q_V) - S\\ &= -\frac{\partial}{\partial z}(q_{Lh} + q_{LT} + q_{Vh} + q_{VT}) - S\\ &= \rho_L\frac{\partial}{\partial z}\left[K_{Lh}\left(\frac{\partial\psi}{\partial z} + 1\right) + K_{LT}\frac{\partial T}{\partial z}\right] + \frac{\partial}{\partial z}\left[D_{Vh}\frac{\partial\psi}{\partial z} + D_{VT}\frac{\partial T}{\partial z}\right] - S\end{aligned} \tag{A.3}$$

where ρ_V and ρ_i (kg m^{-3}) are the density of water vapor and ice, respectively; θ_L and θ_V (m^3 m^{-3}) are the volumetric water content (liquid and vapor, respectively); q_L and q_V (kg m^{-2} s^{-1}) are the soil water fluxes of liquid water and water vapor (positive upwards), respectively. K_{Lh} (m s^{-1}) and K_{LT} (m^2 s^{-1} °C^{-1}) are the isothermal and thermal hydraulic conductivities, respectively. D_{Vh} (kg m^{-2} s^{-1}) is the isothermal vapor conductivity; and D_{VT} (kg m^{-1} s^{-1} °C^{-1}) is the thermal vapor diffusion coefficient.

On the basis of De Vries (1958) and Hansson et al. (2004)'s work, the heat transport function in frozen soils, considering the fully coupled water and heat transport physics, can be expressed as:

$$\begin{aligned}&\frac{\partial}{\partial t}\left[(\rho_s\theta_s C_s + \rho_L\theta_L C_L + \rho_V\theta_V C_V + \rho_i\theta_i C_i)(T - T_r) + \rho_V\theta_V L_0 - \rho_i\theta_i L_f\right] - \rho_L W\frac{\partial\theta_L}{\partial t}\\ &\quad = \frac{\partial}{\partial z}\left(\lambda_{eff}\frac{\partial T}{\partial z}\right) - \frac{\partial}{\partial z}\left[q_L C_L(T - T_r) + q_V(L_0 + C_V(T - T_r))\right] - C_L S(T - T_r)\end{aligned} \tag{A.4}$$

where C_s, C_L, C_V and C_i (J kg^{-1} °C^{-1}) are the specific heat capacities of solids, liquid and water vapor and ice, respectively; ρ_s (kg m^{-3}) is the density of solids; θ_s is the volumetric fraction of solids in the soil; T_r (°C) is the arbitrary reference temperature; L_0 (J kg^{-1}) is the latent heat of vaporization of water at the reference temperature T_r; W (J kg^{-1}) is the differential heat of wetting (the amount of heat released when a small amount of free water is added to the soil matrix).

A.1.3 Coupled mass and heat physics with air flow

In STEMMUS-FT, the temporal dynamics of three phases of water (liquid, vapor and ice), together with the soil dry air component are explicitly presented and simultaneously solved by spatially discretizing the corresponding governing equations of liquid water flow, vapor flow and air flow.

$$\begin{aligned} &\frac{\partial}{\partial t}(\rho_L\theta_L+\rho_V\theta_V+\rho_i\theta_{ice})=-\frac{\partial}{\partial z}(q_{Lh}+q_{LT}+q_{La}+q_{Vh}+q_{VT}+q_{Va})-S \\ &=\rho_L\frac{\partial}{\partial z}\left[K\left(\frac{\partial\psi}{\partial z}+1\right)+D_{TD}\frac{\partial T}{\partial z}+\frac{K}{\gamma_w}\frac{\partial P_g}{\partial z}\right]+\frac{\partial}{\partial z}\left[D_{Vh}\frac{\partial\psi}{\partial z}+D_{VT}\frac{\partial T}{\partial z}+D_{Va}\frac{\partial P_g}{\partial z}\right]-S \end{aligned} \tag{A.5}$$

where q_{Lh}, q_{LT}, and q_{La} (kg m^{-2} s^{-1}) are the liquid water fluxes driven by the gradient of matric potential $\frac{\partial\psi}{\partial z}$, temperature $\frac{\partial T}{\partial z}$, and air pressure $\frac{\partial P_g}{\partial z}$, respectively. q_{Vh}, q_{VT}, and q_{Va} (kg m^{-2} s^{-1}) are the water vapor fluxes driven by the gradient of matric potential $\frac{\partial\psi}{\partial z}$, temperature $\frac{\partial T}{\partial z}$, and air pressure $\frac{\partial P_g}{\partial z}$, respectively. P_g (Pa) is the mixed pore-air pressure. γ_W (kg m^{-2} s^{-2}) is the specific weight of water; D_{TD} (kg m^{-1} s^{-1} °C^{-1}) is the transport coefficient for adsorbed liquid flow due to temperature gradient; D_{Vh} (kg m^{-2} s^{-1}) is the isothermal vapor conductivity; and D_{VT} (kg m^{-1} s^{-1} °C^{-1}) is the thermal vapor diffusion coefficient; D_{Va} is the advective vapor transfer coefficient (Zeng et al., 2011a, b).

STEMMUS-FT takes into account different heat transfer mechanisms, including heat conduction ($\lambda_{eff}\frac{\partial T}{\partial z}$), convective heat transferred by liquid flux ($-C_Lq_L(T-T_r)$, $-C_LS(T-T_r)$), vapor flux ($-[L_0q_V+C_Vq_V(T-T_r)]$) and air flow ($q_aC_a(T-T_r)$). The latent heat of vaporization ($\rho_V\theta_VL_0$), the latent heat of freezing/thawing ($-\rho_i\theta_iL_f$) and a source term associated with the exothermic process of wetting of a porous medium (integral heat of wetting) ($-\rho_LW\frac{\partial\theta_L}{\partial t}$).

$$\begin{aligned} &\frac{\partial}{\partial t}\left[(\rho_s\theta_sC_s+\rho_L\theta_LC_L+\rho_V\theta_VC_V+\rho_{da}\theta_aC_a+\rho_i\theta_iC_i)(T-T_r)+\rho_V\theta_VL_0-\rho_i\theta_iL_f\right]- \\ &\rho_LW\frac{\partial\theta_L}{\partial t} \\ &=\frac{\partial}{\partial z}\left(\lambda_{eff}\frac{\partial T}{\partial z}\right)-\frac{\partial}{\partial z}[q_LC_L(T-T_r)+q_V(L_0+C_V(T-T_r))+q_aC_a(T-T_r)]-C_LS(T-T_r) \end{aligned} \tag{A.6}$$

where ρ_{da} (kg m^{-3}) is the density of dry air; C_a (J kg^{-1} °C^{-1}) is the specific heat capacity of dry air; q_a (kg m^{-2} s^{-1}) is the air flux. The air flow balance equation for solving the coupled water and heat equations is written as Zeng et al. (2011a, b) and Zeng and Su (2013):

$$\frac{\partial}{\partial t}[\varepsilon\rho_{da}(S_a+H_cS_L)]=\frac{\partial}{\partial z}\left[D_e\frac{\partial\rho_{da}}{\partial z}+\rho_{da}\frac{S_aK_g}{\mu_a}\frac{\partial P_g}{\partial z}-H_c\rho_{da}\frac{q_L}{\rho_L}+(\theta_aD_{Vg})\frac{\partial\rho_{da}}{\partial z}\right] \tag{A.7}$$

where ε is the porosity; S_a (=1-S_L) is the degree of air saturation in the soil; S_L (=θ_L/ε) is the degree of saturation in the soil; H_c is Henry's constant; D_e (m^2 s^{-1}) is the molecular diffusivity of water vapor in soil; K_g (m^2) is the intrinsic air permeability; μ_a (kg m^{-2} s^{-1}) is the air viscosity; θ_a (=θ_V) is the volumetric fraction of dry air in the soil; and D_{Vg} (m^2 s^{-1}) is the gas phase longitudinal dispersion coefficient.

A.2 Snowpack module UEB

A.2.1 Mass balance equation

The increase or decrease of snow water equivalence with time equals the difference of income and outgoing water flux:

$$\frac{dSWE}{dt} = P_r + P_s - M_r - E \tag{A.8}$$

where *SWE* (m) is the snow water equivalent; P_r (m/s) is the rainfall rate; P_s (m/s) is the snowfall rate; M_r (m/s) is the meltwater outflow from the snowpack; and E is the sublimation from the snowpack.

A.2.2 Energy balance equation

The energy balance of snowpack can be expressed as:

$$\frac{dU}{dt} = Q_{sn} + Q_{li} + Q_p + Q_g - Q_{le} + Q_h + Q_e - Q_m \tag{A.9}$$

where Q_{sn} (W/m^2) is the net shortwave radiation; Q_{li} (W/m^2) is the incoming longwave radiation; Q_p (W/m^2) is the advected heat from precipitation; Q_g (W/m^2) is the ground heat flux; Q_{le} (W/m^2) is the outgoing longwave radiation; Q_h (W/m^2) is the sensible heat flux; Q_e (W/m^2) is the latent heat flux due to sublimation/condensation; and Q_m (W/m^2) is the advected heat removed by meltwater.

Equations (8) and (9) form a coupled set of first order, nonlinear ordinary differential equations. Euler predictor-corrector approach was employed in UEB model to solve the initial value problems of these equations (Tarboton and Luce, 1996).

A.3 Albedo calculation

A.3.1 Ground albedo

Instead of the constant bare soil albedo in the original UEB model, the bare soil albedo is expressed as a decreasing linear function of soil moisture in STEMMUS-UEB.

$$\alpha_{g,v} = \alpha_{sat} + \min\{\alpha_{sat}, \max[(0.11 - 0.4\theta), 0]\} \tag{A.10}$$

$$\alpha_{g,ir} = 2\alpha_{g,v} \tag{A.11}$$

where $\alpha_{g,v}$ and $\alpha_{g,ir}$ are the bare soil/ground albedo for the visible and infrared band, respectively. α_{sat} is the saturated soil albedo, depending on local soil color. θ is the surface volumetric soil moisture.

A.3.2 Vegetation albedo

The calculation of vegetation albedo is developed to capture the essential features of a two-stream approximation model using asymptotic equation. It approaches the underlying surface albedo $\alpha_{g,\lambda}$ or the thick canopy albedo $\alpha_{c,\lambda}$ when the L_{SAI} is close to zero or infinity.

$$\alpha_{Veg,b,\lambda} = \alpha_{c,\lambda}\left[1 - \exp\left(-\frac{\omega_\lambda \beta L_{SAI}}{\mu \alpha_{c,\lambda}}\right)\right] + \alpha_{g,\lambda}\exp[-\left(1 + \frac{0.5}{\mu}\right)L_{SAI}] \tag{A.12}$$

$$\alpha_{Veg,d,\lambda} = \alpha_{c,\lambda}\left[1 - \exp\left(-\frac{2\omega_\lambda \beta L_{SAI}}{\alpha_{c,\lambda}}\right)\right] + \alpha_{g,\lambda}\exp[-2\,L_{SAI}] \tag{A.13}$$

where subscripts Veg, b, d, c, g and λ represent vegetation, direct beam, diffuse radiation, thick canopy, ground, and spectrum bands of either visible or infrared bands. μ is the cosine of solar zenith angle; ω_λ is the single scattering albedo, 0.15 for visible and 0.85 for infrared band, respectively; β is assigned as 0.5; L_{SAI} is the sum of leaf area index LAI and stem area index SAI; $\alpha_{c,\lambda}$ is the thick canopy albedo dependent on vegetation types.

The bulk snow-free surface albedo, averaged between bare ground albedo and vegetation albedo, then is written as:

$$\alpha_{\eta,\lambda} = \alpha_{Veg,\lambda} f_{Veg} + \alpha_{g,\lambda}(1 - f_{Veg}) \tag{A.14}$$

where $\alpha_{\eta,\lambda}$ is the averaged bulk snow-free surface albedo; f_{Veg} is the fraction of vegetation cover.

A.3.3 Snow albedo

According to Dickinson et al. (1993), snow albedo can be expressed as a function of snow surface age and solar illumination angle. The snow surface age, which is dependent on snow surface temperature and snowfall, is updated with each time step in UEB. Visible and near infrared bands are separately treated when calculating reflectance, which are further averaged as the albedo with modifications of illumination angle and snow age. The reflectance in the visible and near infrared bands can be written as:

$$\alpha_{vd} = \left(1 - C_v S_{age}\right)\alpha_{vo} \tag{A.15}$$

$$\alpha_{ird} = \left(1 - C_{ir} S_{age}\right)\alpha_{iro} \tag{A.16}$$

where α_{vd} and α_{ird} represent diffuse reflectance in the visible and near infrared bands, respectively. C_v (= 0.2) and C_{ir} (=0.5) are parameters that quantify the sensitivity of the visible and infrared band albedo to snow surface aging (grain size growth), α_{vo} (=0.85) and α_{iro} (=0.65) are fresh snow reflectance in visible and infrared bands, respectively. S_{age} is a function to account for aging of the snow surface, and is given by:

$$S_{age} = \frac{\tau}{1 + \tau} \tag{A.17}$$

where τ is the non-dimensional snow surface age that is incremented at each time step by the quantity designed to emulate the effect of the growth of surface grain sizes.

$$\Delta\tau = \frac{r_1 + r_2 + r_3}{\tau_o}\Delta t \tag{A.18}$$

where Δt is the time step in seconds with $\tau_o = 10^6$s. r_1 is the parameter to represent the effect of grain growth due to vapor diffusion, and is dependent on snow surface temperature:

$$r_1 = \exp\left[5000\left(\frac{1}{273.16} - \frac{1}{T_s}\right)\right] \tag{A.19}$$

r_2 describes the additional effect near and at the freezing point due to melt and refreeze:

$$r_2 = \min\left(r_1^{10}, 1\right) \tag{A.20}$$

r_3=0.03 (0.01 in Antarctica) represents the effect of dirt and soot.

The reflectance of radiation with illumination angle (measured relative to the surface normal) is computed as:

$$\alpha_v = \alpha_{vd} + 0.4\, f(\varphi)(1 - \alpha_{vd}) \tag{A.21}$$

$$\alpha_{ir} = \alpha_{ird} + 0.4\, f(\varphi)(1 - \alpha_{ird}) \tag{A.22}$$

where $f(\varphi) = \begin{cases} \frac{1}{b}\left[\frac{b+1}{1+2b\cos(\varphi)} - 1\right], & for \cos(\varphi) < 0.5 \\ 0, & otherwise \end{cases}$

where b is a parameter set at 2 as Dickinson et al. (1993).

When the snowpack is shallow (depth z<h=0.01m), the albedo is calculated by interpolating between the snow albedo and bare ground albedo with the exponential term approximating the exponential extinction of radiation penetration of snow.

$$A_{v/ir} = r\alpha_{g,v/ir} + (1 - r)\alpha_{v/ir} \tag{A.23}$$

where $r = \left(1 - \frac{z}{h}\right) e^{-z/2h}$.

Notation

Symbol	Parameter	Unit	Value
Main inputs			
Soil model component (STEMMUS-FT)			
a	Fitted parameter for soil surface resistance	-	0.3565
$b(z)$	Normalized water uptake distribution	m^{-1}	
C_a	Specific heat capacity of dry air	J kg^{-1} $°C^{-1}$	1.005
C_{app}	Apparent heat capacity	J kg^{-1} $°C^{-1}$	
C_i	Specific heat capacity of ice	J kg^{-1} $°C^{-1}$	2.0455
C_L	Specific heat capacity of liquid	J kg^{-1} $°C^{-1}$	4.186
C_s	Specific heat capacity of soil solids	J kg^{-1} $°C^{-1}$	
C_{soil}	Heat capacity of the bulk soil	J kg^{-1} $°C^{-1}$	
C_V	Specific heat capacity of water vapor	J kg^{-1} $°C^{-1}$	1.87
c_p	Specific heat capacity of air	J kg^{-1} K^{-1}	
D_e	Molecular diffusivity of water vapor in soil	m^2 s^{-1}	
D_{TD}	Transport coefficient for adsorbed liquid flow due to temperature gradient	kg m^{-1} s^{-1} $°C^{-1}$	
D_{Va}	Advective vapor transfer coefficient	s	
D_{Vg}	Gas phase longitudinal dispersion coefficient	m^2 s^{-1}	
D_{Vh}	Isothermal vapor conductivity	kg m^{-2} s^{-1}	
D_{VT}	Thermal vapor diffusion coefficient	kg m^{-1} s^{-1} $°C^{-1}$	
H_c	Henry's constant	-	0.02
K	Hydraulic conductivity	m s^{-1}	
K_g	Intrinsic air permeability	m^2	
K_{Lh}	Isothermal hydraulic conductivities	m s^{-1}	
K_{LT}	Thermal hydraulic conductivities	m^2 s^{-1} $°C^{-1}$	
K_s	Soil saturated hydraulic conductivity	m s^{-1}	
L_0	Latent heat of vaporization of water at the reference temperature	J kg^{-1}	
LAI_{eff}	Effective leaf area index	-	
L_f	Latent heat of fusion	J kg^{-1}	3.34E+05
n	Van Genuchten fitting parameters	-	
r_a^c	Aerodynamic resistance for canopy surface	s m^{-1}	
r_a^s	Aerodynamic resistance for bare soil	s m^{-1}	
$r_{c,min}$	Minimum canopy surface resistance	s m^{-1}	
$r_{l,min}$	Minimum leaf stomatal resistance	s m^{-1}	
r_s	Soil surface resistance	s m^{-1}	
r_{sl}	Resistance to molecular diffusion of the water surface	s m^{-1}	10
R_n	Net radiation	MJ m^{-2} day^{-1}	
R_n^c	Net radiation at the canopy surface	MJ m^{-2} day^{-1}	
R_n^s	Net radiation at the soil surface	MJ m^{-2} day^{-1}	
S_a	Degree of saturation of the soil air	-	$=1-S_L$
S_L	Degree of water saturation in the soil	-	$=\theta_L/\varepsilon$
S_p	Potential water uptake rate	s^{-1}	

t	Time	s	
T_p	Potential transpiration	m s^{-1}	
T_r	Arbitrary reference temperature	°C	20
W	Differential heat of wetting	J kg^{-1}	
z	Vertical space coordinate (positive upwards)	m	
α	Air entry value of soil	m^{-1}	
$a(h)$	Reduction coefficient related to soil water potential	-	
ε	Porosity	-	
λ_{eff}	Effective thermal conductivity of the soil	W m^{-1} °C^{-1}	
θ_s	Volumetric fraction of solids in the soil	m^3 m^{-3}	
θ_{sat}	Saturated soil water content	m^3 m^{-3}	
θ_{r}	Residual soil water content	m^3 m^{-3}	
θ_1	Topsoil water content	m^3 m^{-3}	
θ_{min}	Minimum water content above which soil is able to deliver vapor at a potential rate	m^3 m^{-3}	
ρ_a	Air density	kg m^{-3}	
ρ_{da}	Density of dry air	kg m^{-3}	
ρ_i	Density of ice	kg m^{-3}	920
ρ_L	Density of soil liquid water	kg m^{-3}	1000
ρ_s	Density of solids	kg m^{-3}	
ρ_V	Density of water vapor	kg m^{-3}	
γ_W	Specific weight of water	kg m^{-2} s^{-2}	
μ_a	Air viscosity	kg m^{-2} s^{-1}	
Snow model component (UEB)			
T_{r}	Air temperature above which precipitation is all rain	°C	3.5
T_{sn}	Air temperature below which precipitation is all snow	°C	0
$\varepsilon_{\mathrm{sn}}$	Emissivity of snow	-	0.99
C_{g}	Ground heat capacity	J kg^{-1} °C^{-1}	2.09
z_{o}	Snow surface aerodynamic roughness	m	0.001
L_{c}	Liquid holding capacity of snow	-	0.05
K_{sn}	Snow saturated hydraulic conductivity	m h^{-1}	160
α_{vo}	Visual new snow albedo	-	0.95
α_{iro}	Near-infrared new snow albedo	-	0.65
α_{bg}	Bare ground albedo	-	Eqs. A10 - A14
D_{e}	Thermally active depth of soil	m	0.4
λ_{sn}	Snow surface thermal conductivity	m h^{-1}	0.02
ρ_{sn}	Snow density	kg m^{-3}	450
A_{ed}	Albedo extinction depth	m	0.0001
F_{c}	Forest cover fraction	-	0
D_{f}	Drift factor	-	1
ρ_{s}	Soil density	kg m^{-3}	1700
Main outputs			
Soil model component (STEMMUS-FT)			
ψ	Soil water potential	m	

P_g	Mixed pore-air pressure	Pa
T	Soil temperature	°C
θ	Volumetric water content	$m^3\ m^{-3}$
θ_i	Soil ice volumetric water content	$m^3\ m^{-3}$
θ_L	Soil liquid volumetric water content	$m^3\ m^{-3}$
θ_V	Soil vapor volumetric water content	$m^3\ m^{-3}$
θ_a	Volumetric fraction of dry air in the soil	$m^3\ m^{-3}$
q	Water flux	$kg\ m^{-2}\ s^{-1}$
q_a	Dry air flux	$kg\ m^{-2}\ s^{-1}$
q_L	Soil liquid water fluxes (positive upwards)	$kg\ m^{-2}\ s^{-1}$
q_{La}	Liquid water flux driven by the gradient of air pressure	$kg\ m^{-2}\ s^{-1}$
q_{Lh}	Liquid water flux driven by the gradient of matric potential	$kg\ m^{-2}\ s^{-1}$
q_{LT}	Liquid water flux driven by the gradient of temperature	$kg\ m^{-2}\ s^{-1}$
q_V	Soil water vapor fluxes (positive upwards)	$kg\ m^{-2}\ s^{-1}$
q_{Va}	Water vapor flux driven by the gradient of air pressure	$kg\ m^{-2}\ s^{-1}$
q_{Vh}	Water vapor flux driven by the gradient of matric potential	$kg\ m^{-2}\ s^{-1}$
q_{VT}	Water vapor flux driven by the gradient of temperature	$kg\ m^{-2}\ s^{-1}$
S	Sink term for transpiration	s^{-1}
S_h	Latent heat flux density	$W\ m^{-3}$
Snow model component (UEB)		
P_r	Precipitation in the form of rain	$m\ s^{-1}$
P_s	Precipitation in the form of snow	$m\ s^{-1}$
SWE	Snow water equivalent	m
Q_h	Surface Sensible Heat Flux	$W\ m^{-2}$
Q_e	Surface Latent Heat Flux	$W\ m^{-2}$
E	Surface Sublimation	$m\ s^{-1}$
T_{surf}	Snow Surface Temperature	°C
U	Energy Content	
M_r	Melt outflow rate	$m\ s^{-1}$
$A_{v/ir}$	Surface Albedo	-
Q_m	Heat advected by melt outflow	$W\ m^{-2}$
Q_{sn}	Net shortwave radiation	$W\ m^{-2}$
Q_{li}	Net longwave radiation	$W\ m^{-2}$
τ	No-dimensional snow age	-

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Tables and Figures

Table 1. Brief overview of current soil-snow modelling efforts.

Model	Soil					Snow						Relevant reference
	Water balance	Energy balance	Air balance	Water-heat coupled	Others (vapor, freeze-thaw, convective heat)	Snow layer	Snow energy budget	Water flow	Snow albedo	Snow density	Other processes (snow compaction, wind, and vegetation effect)	
CABLE-SLI	Richards	HT_cond, Advc	No	Yes	Vapor; HT_convect (liquid)	Multilayer	HT_cond, Advc	Mass conservation	Albedo_SNW_1A	Density_SNW_1	Snow compaction (overburden and metamorphism)	Cuntz and Haverd (2018)
CLASS	Richards	HT_cond	No	No	No vapor; LH_phas	Single	HT_cond	Snowfall, energy driven snow melting	Albedo_SNW_1B	Density_SNW_2A	-	Barlett et al. (2006)
CLM5	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer up to five	HT_cond	Mass conservation	Albedo_SNW_2	Density_SNW_4A	Snow compaction (metamorphism, overburden, melting, wind-drift)	Lawrence et al., (2019)
HTESSEL	Richards	HT_cond	No	No	No vapor; LH_phas	Single	HT_cond	Mass conservation	Albedo_SNW_3B	Density_SNW_4B	Snow compaction (overburden and metamorphism)	Dutra et al. (2010)
HTESSEL-ML	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer up to 3	HT_cond	Mass conservation	Albedo_SNW_3B	Density_SNW_4B	Snow compaction (overburden and metamorphism)	Dutra et al. (2012)
SURFEX-ISBA-ES01	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer, 3	HT_cond	Mass conservation	Albedo_SNW_3A	Density_SNW_4C	Snow compaction and settling	Boone and Etchevers (2001)
SURFEX-ISBA-ES16	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer, 12	HT_cond	Mass conservation	Albedo_SNW_3C	Density_SNW_4D	Snow compaction; wind-induced densification	Decharme et al. (2016)
SURFEX-ISBA-MEB	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer, 12	HT_cond	Mass conservation	Albedo_SNW_3C	Density_SNW_4D	Snow compaction; wind-induced densification; Vegetation effect (interception/ unloading; snow fraction); litter layer; Multi-component energy balance	Boone et al. (2017)
SURFEX-Crocus	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer (dynamic)	HT_cond	Mass conservation	Albedo_SNW_3D	Density_SNW_4F	Snow metamorphism; compaction; wind drift; sublimation/ hoar deposition	Vionnet et al. (2012)
JSBACH	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer up to 5	HT_cond	Mass conservation	Constant	Constant	-	Ekici et al. (2014)
JULES	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer up to 5	HT_cond	Mass conservation	Albedo_SNW_3A	Density_SNW_4B	Snow compaction	Best (2011)
Noah-MP	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer up to 3	HT_cond	Mass conservation	Albedo_SNW_2	Density_SNW_2B	-	Niu et al. (2011)
ORCHIDEE-ES	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer, 3	HT_cond	Mass conservation	Albedo_SNW_3E	Density_SNW_4B	Snow compaction (overburden and metamorphism)	Wang et al. (2013)
SNOWPACK	Richards	HT_cond	No	Yes	Vapor; HT_convect (liquid)	Multilayer	HT_cond	Mass conservation, vapor	Albedo_SNW_3D	Density_SNW_4G	Explicit prognostic settlement; Snow metamorphism; compaction; wind drift; sublimation	Lehning et al. (1999)
WEB-DHM	Richards	HT_cond	No	No	No vapor; LH_phas	Single	HT_cond	Mass conservation	Albedo_SNW_1B	Constant	Vegetation interception	Wang et al. (2009)
WEB-DHM-S	Richards	HT_cond	No	No	No vapor; LH_phas	Multilayer up to 3	HT_cond	Mass conservation	Albedo_SNW_3F	Density_SNW_4B	Snow compaction	Shrestha et al. (2010)
HydroSiB2-SF	Richards	HT_cond	No	Yes	Vapor; enthalpy-based FT; LH_phas	Multilayer up to 3	HT_cond, Advc	Mass conservation	Albedo_SNW_3F	Density_SNW_4B	Snow compaction	Wang et al. (2017)
WEB-GM	-	-	-	-	-	Multilayer, vary with snow depth	Enthalpy based heat transfer	Mass conservation	Albedo_SNW_4	Density_SNW_3	Snow compaction (metamorphism, snow densification, melting);	Ding et al. (2017)

SWAP	Richards	HT_cond	No	No	No vapor; LH_phas	Single	-	Mass conservation	Constant	Density_SNW_4H	Vegetation interception	Gusev and Nasonova (2003)
COUP	Richards	HT_cond, Advc	No	Yes	Vapor; HT_convect (liquid)	Single	HT_cond	Mass conservation	Albedo_SNW_1A	Density_SNW_2C	Snow compaction	Jansson (2012)
SHAW	Richards	HT_cond, Advc	No	Yes	Vapor; HT_convect (liquid, vapor)	Multilayer	HT_cond, Advc	Mass conservation, vapor	Albedo_SNW_1C	Density_SNW_4E	Snow compaction, settling	Flerchinger and Saxton (1989); Flerchinger (2017)
HYDRUS	Richards	HT_cond, Advc	No	Yes	Vapor; HT_convect (liquid, vapor)	-	-	-	-	-	-	Hansson et al. (2004); Šimůnek et al. (2008)
STEMMUS-UEB	Richards	HT_cond, Advc	Yes	Yes	Vapor; LH_phas; HT_convect (liquid, vapor, dry air); Various complexity of SHP	Single	HT_cond, Advc	Mass conservation	Albedo_SNW_3F	Constant	Empirical wind drift and vegetation interception	This study

Note:

HT_cond, Heat conduction;

Advc, Advection;

LH_phas, Latent heat due to phase change;

HT_Convect, Convective heat due to liquid;

SHP, soil physical process;

Albedo_SNW_1A, Snow albedo 1A, Function of snow age;

Albedo_SNW_1B, Snow albedo 1B, Empirical function, considering dry/wet states;

Albedo_SNW_1C, Snow albedo 1C, Function of extinction coefficient, grain-size, and solar zenith angle;

Albedo_SNW_2, Snow albedo 2, Two-stream radiative transfer solution, considering snow aging, solar zenith angle, optical parameters, and impurity;

Albedo_SNW_3A, Snow albedo 3A, Prognostic snow albedo, considering aging effect;

Albedo_SNW_3B, Snow albedo 3B, Prognostic snow albedo, considering aging effect and vegetation type dependent;

Albedo_SNW_3C, Snow albedo 3C, Prognostic snow albedo, considering aging and optical diameter;

Albedo_SNW_3D, Snow albedo 3D, Prognostic snow albedo, considering age and microstructure;

Albedo_SNW_3E, Snow albedo 3E, Prognostic snow albedo, considering aging effect and dry/wet states;

Albedo_SNW_3F, Snow albedo 3F, Prognostic snow albedo considering aging effect, solar zenith angle;

Albedo_SNW_4, Snow albedo 4, Diagnostic snow albedo, considering snow aging, sleet/snowfall fraction, grain diameter, cloud fraction, and solar elevation effect;

Density_SNW_1, Snow density 1, relying on in situ measurements;

Density_SNW_2A, Snow density 2A, function of air temperature;

Density_SNW_2B, Snow density 2B, Function of extinction coefficient and grain-size;

Density_SNW_2C, Snow density 2C, Function of old (densification), new-fallen (air temperature) snow pack density, and snow depth;

Density_SNW_3, Snow density 3, Diagnostic density, considering wet-bulb temperature;

Density_SNW_4A, Snow density 4A, Prognostic density, considering temperature, wind effect, snow compaction, water/ice states;

Density_SNW_4B, Snow density 4B, Prognostic density, considering overburden and thermal metamorphisms;

Density_SNW_4C, Snow density 4C, Prognostic snow density, considering snow compaction and settling;

Density_SNW_4D, Snow density 4D, Prognostic snow density, considering snow compaction and wind-induced densification;

Density_SNW_4E, Snow density 4E, Prognostic snow density, considering snow compaction, settling, and vapor transfer;

Density_SNW_4F, Snow density 4F, Prognostic density, function of wind speed and air temperature;

Density_SNW_4G, Snow density 4G, Prognostic density, function of stress state and microstructure;

Density_SNW_4H, Snow density 4H, Prognostic density, considering snow temperature.

Table 2. Main subroutines in STEMMUS-UEB

Model Subroutines	Main functions	Main inputs	Main outputs	Subroutine-Connections
Soil module				
Air_sub	Solves soil dry air balance equation	Water vapor density, diffusivity, dispersion coefficient; dry air density, gas conductivity, flux; liquid water flux; top and bottom boundary conditions	Soil air pressure profile	CondV_DVg, CondL_h, Condg_k_g, Density_V, h_sub --->; --> Enrgy_sub,
CondL_h	Calculates soil hydraulic conductivity	Soil hydraulic parameters; soil matric potential; soil temperature	Soil hydraulic conductivity; soil water content	StartInit --->; --> h_sub; Air_sub; Enrgy_sub,
CondT_coeff	Calculates soil thermal capacity and conductivity	Thermal properties of soil constituents; soil texture; soil water content; volumetric fraction of dry air; dry air density; vapor density	Soil thermal capacity and conductivity	StartInit, CondL_h, Density_V, Density_DA, EfeCapCond --->; --> Enrgy_sub,
CondV_DVg	Calculates flux of dry air and vapor dispersity	Gas conductivity, dry air pressure, volumetric fraction of dry air; saturated soil water content	Dry air flux and vapor dispersion coefficient	StartInit, CondL_h, Condg_k_g --->; --> h_sub; Air_sub; Enrgy_sub,
CondL_Tdisp	Calculates transport coefficient for adsorbed liquid flow	Soil porosity, soil water content, temperature, matric potential, volumetric fraction of dry air	Transport coefficient for adsorbed liquid flow and the heat of wetting	StartInit, CondL_h, Condg_k_g --->; --> h_sub; Enrgy_sub,
Condg_k_g	Calculates gas conductivity	Soil porosity, saturated hydraulic conductivity, volumetric fraction of dry air	Gas conductivity	StartInit, CondL_h --->; --> CondV_DVg,
Density_DA	Calculates dry air density	Soil temperature, matric potential, dry air pressure; vapor density and its derivative with respect to temperature and matric potential	Density of dry air	StartInit, CondL_h, Density_V --->; --> CondT_coeff, Air_sub, Enrgy_sub,
Density_V	Calculates vapor density and its derivative with respect to temperature and matric potential	Soil temperature, matric potential	Vapor density and its derivative with respect to temperature and matric potential	CondL_h --->; --> Density_DA, CondT_coeff, h_sub, Air_sub, Enrgy_sub,
EfeCapCond	Calculates soil thermal capacity and conductivity	Thermal properties of soil constituents; soil texture; soil water content; volumetric fraction of dry air; dry air density; vapor density	Soil heat capacity; thermal conductivity	StartInit, CondL_h, Density_V, Density_DA --->; --> CondT_coeff,
Enrgy_sub	Solves soil energy balance equation	Soil thermal properties, soil hydraulic conductivity, soil matric potential, soil water content, soil temperature, soil dry air pressure, density of dry air, heat of wetting, vapor density, liquid water flux, vapor flux, dry air flux, meterological forcing, top and bottom boundary conditions	Soil temperature profile, liquid water flux, vapor flux, and dry air flux, surface and bottom energy fluxes	Air_sub, h_sub, CondL_h, CondV_DVg, CondL_Tdisp, CondT_coeff, Density_D, Density_DA, PREDICORR -->,
Forcing_PARM	Disaggregates the meteorological forcing into the required time steps	Observed meteorological forcing at hourly/daily time scale	Meteorological forcings at model required time scale	StartInit --->; --> h_sub, Enrgy_sub,
h_sub	Solves soil water balance equation	Soil temperature, soil water content, matric potential, soil hydraulic conductivity, heat of wetting, soil dry air pressure, vapor density, diffusivity, dispersivity, volumetric fraction of vapor, meteorological forcing, top and bottom boundary conditions	Soil matric potential profile, top and bottom water fluxes, evaporation	StartInit, CondV_DVg, CondL_h, CondV_DE, CondL_Tdisp, Condg_k_g, Density_V, Forcing_PARM, ALBEDO, PARTSNOW, PREDICORR --->; --> Air_sub, Enrgy_sub,
StartInit	Initializes model setup	Soil texture, thermal properties of soil constituents, initial soil water content and temperature, top and bottom boundary condition settings	-	--> CondV_DVg, CondL_h, CondV_DE, CondL_Tdisp, Condg_k_g, Density_DA, EfeCapCond, Forcing_PARM, h_sub,
Diff_Moisture_Heat	Solves soil water and energy balance equations independently	Soil thermal properties, soil hydraulic conductivity, soil matric potential, soil water content, soil temperature, meteorological forcing, top and bottom boundary conditions	Soil water content and temperature profile, liquid water flux, surface and bottom water and energy fluxes	StartInit, CondT_coeff, Forcing_PARM, ALBEDO, PARTSNOW, PREDICORR -->,
Snowpack module				
agesn	Calculates snow age	Snow surface temperature, snowfall	Updated snow age	PARTSNOW, PREDICORR -->; --> ALBEDO,
ALBEDO	Calculates snow albedo	Fresh snow reflectance at visible and near infrared bands, snow age, bare ground albedo, albedo extinction parameter, snow water equivalent	Snow albedo	agesn --->; --> PREDICORR,
PARTSNOW	Partitions precipitation into rainfall and snowfall	Precipitation, air temperature, temperature thresholds for rainfall/snowfall	Rainfall, snowfall	Forcing_PARM --->; --> PREDICORR,

PREDICORR	Solves the snow mass and energy balance equations and updates state variables SWE and U	Air temperature, snow albedo, wind speed, relative humidity, rainfall/snowfall, shortwave/longwave radiation, site parameters	Snow energy content, water equivalent, snow albedo, snow surface temperature, meltwater outflow rate, snow sublimation, snowfall/rainfall	Forcing_PARM --->; --> agesn[2], ALBEDO[2].

Note:

---> means the relevant subroutines which are incoming to the current one, --> means the relevant subroutines for which the current subroutine is output to;

agesn[2] and ALBEDO[2], means the use of subroutines agesn and ALBEDO after solving the snowpack energy and mass conservation equations, to update the snow age and albedo.

Table 3. Numerical experiments with various mass and energy transfer schemes with/without explicit consideration of snow cover (Eqs. A1-A7 are listed in Appendix A.1; Eqs. A8-A9 are listed in Appendix A.2).

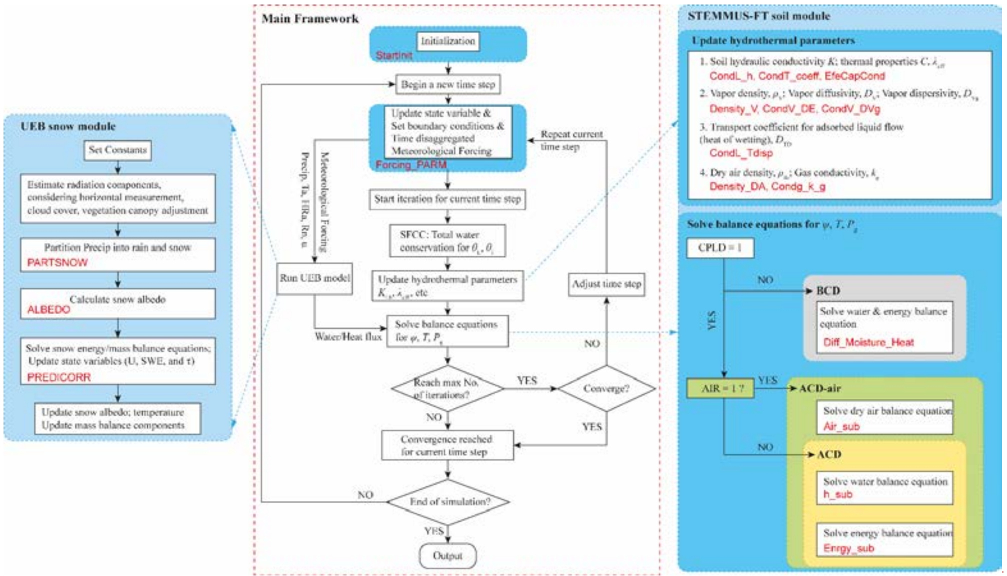
Processes		Experiments	
Snowpack (SNW)	Mass and energy transfer in soils (SMETr)		
SNW =1: UEB (Eqs. A.8 & A.9)	SMETr=1: basic coupled water-heat transfer (Eqs. A.1 & A.2)	BCD-Snow	STEMMUS-FT_Snow
	SMETr=2: advanced coupled water-heat transfer without air flow (Eqs. A.3 & A.4)	ACD-Snow	
	SMETr=3: advanced coupled water-heat transfer with air flow (Eqs. A.5, A.6 & A.7)	ACD-air-Snow	
SNW =0: No discrimination of snow and rainfall	SMETr=1: basic coupled water-heat transfer (Eqs. A.1 & A.2)	BCD-No-Snow	STEMMUS-FT_No-snow
	SMETr=2: advanced coupled water-heat transfer without air flow (Eqs. A.3 & A.4)	ACD-No-Snow	
	SMETr=3: advanced coupled water-heat transfer with air flow (Eqs. A.5, A.6 & A.7)	ACD-air-No-Snow	

Table 4. Comparative statistics values of various model versions for snow albedo, LE, soil temperature, and soil moisture. The best statistical performance is highlighted by bold fonts, while the values with poor statistical model performance is underlined with the italic fonts.

Experiments		Statistics	Snow albedo	LE (mm/d)	Soil temperature (ºC)					Soil moisture ($cm^3\ cm^{-3}$)				
					5cm	10cm	20cm	40cm	80cm	5cm	10cm	20cm	40cm	80cm
STEMMUS-FT_Snow	BCD	BIAS	-0.0100	*0.162*	**-0.071**	*0.150*	-0.048	-1.127	-0.1390	0.0064	0.0091	0.0048	**0.0031**	1.80E-03
		R^2	0.296	0.278	**0.976**	0.958	0.881	0.626	0.810	0.704	*0.586*	*0.310*	*0.387*	*0.237*
		RMSE	0.033	0.579	0.4697	0.415	0.544	1.548	0.5352	0.0194	0.0223	0.0307	0.0322	0.0118
	ACD	BIAS	-0.0049	-0.020	-0.224	**0.054**	-0.032	-0.982	**0.0129**	-0.0014	**0.0024**	**0.0001**	0.0045	7.57E-04
		R^2	0.253	0.232	0.964	0.969	0.971	0.944	0.995	0.878	**0.960**	**0.991**	0.992	0.982
		RMSE	0.032	0.305	0.4462	0.374	0.209	1.190	0.1201	0.0087	**0.0041**	0.0028	0.0055	0.0019
	ACD-air	BIAS	**-0.0048**	**-0.019**	-0.223	0.055	-0.032	-0.982	0.0130	-0.0013	0.0025	0.0001	0.0045	**7.55E-04**
		R^2	**0.338**	0.217	0.963	**0.969**	0.971	**0.944**	**0.995**	0.883	0.960	0.990	**0.992**	**0.982**
		RMSE	**0.031**	0.314	0.4464	0.374	0.210	1.190	0.1200	0.0084	0.0042	**0.0028**	**0.0055**	**0.0019**
STEMMUS-FT_No-snow	BCD	BIAS	-0.0123	*0.157*	-0.073	*0.149*	-0.048	-1.128	-0.1397	0.0099	0.0092	0.0048	0.0031	1.70E-03
		R^2	-	0.303	0.976	0.958	0.881	0.627	0.810	0.771	*0.581*	*0.309*	*0.386*	*0.240*
		RMSE	0.038	0.565	0.4673	0.415	0.544	1.548	0.5354	0.0261	0.0224	0.0307	0.0322	0.0117
	ACD	BIAS	-0.0079	-0.031	-0.213	0.065	-0.023	-0.977	0.0154	**-0.0010**	0.0026	0.0002	0.0046	8.29E-04
		R^2	-	**0.363**	0.964	0.969	**0.973**	0.943	0.995	**0.887**	0.959	0.991	0.991	0.979
		RMSE	0.037	**0.242**	0.4352	**0.370**	**0.201**	1.186	0.1210	**0.0081**	0.0044	0.0028	0.0058	0.0020
	ACD-air	BIAS	-0.0079	-0.031	-0.210	0.072	**-0.014**	**-0.968**	0.0222	-0.0011	0.0026	0.0003	0.0049	9.13E-04
		R^2	-	0.358	0.965	0.969	0.972	0.943	0.995	0.886	0.960	0.991	0.990	0.979
		RMSE	0.037	0.243	**0.4349**	0.374	0.202	**1.180**	**0.1198**	0.0082	0.0041	0.0028	0.0061	0.0020

Note: $BIAS = \frac{\sum_{i=1}^{n}(y_i - \hat{y}_i)}{n}$, $R^2 = 1 - \frac{\sum_{i=1}^{n}(y_i - \hat{y}_i)^2}{\sum_{i=1}^{n}(y_i - \bar{y})^2}$, $RMSE = \sqrt{\frac{\sum_{i=1}^{n}(y_i - \hat{y}_i)^2}{n}}$, where y_i, $\hat{y}_i$, are the measured and model simulated values of the selected variable (snow albedo, LE, soil temperature/moisture); $\bar{y}$ is the mean values of the measurements of the selected variable (snow albedo, LE, soil temperature/moisture); n is the number of data points.

The correlation is all significant at the 0.01 level, except for "-", which indicates that the correlation is not significant.



UEB snow module
Set Constants
Estimate radiation components, considering horizontal measurement, cloud cover, vegetation canopy adjustment
Partition Precip into rain and snow
PARTSNOW
Calculate snow albedo
ALBEDO
Solve snow energy/mass balance equations; Update state variables (U, SWE, and τ)
PREDICORR
Update snow albedo; temperature
Update mass balance components
Main Framework
Initialization
StartInit
Begin a new time step
Update state variable & Set boundary conditions & Time disaggregated Meteorological Forcing
Forcing_PARM
Meteorological Forcing
Precip, Ta, HRa, Rn, u
Run UEB model
Water/Heat flux
Start iteration for current time step
SFCC: Total water conservation for θ_L, θ_i
Update hydrothermal parameters K_{Lh}, λ_{eff}, etc
Solve balance equations for φ, T, P_g
Reach max No. of iterations?
YES
NO
Converge?
NO
YES
Adjust time step
Repeat current time step
Convergence reached for current time step
End of simulation?
NO
YES
Output
STEMMUS-FT soil module
Update hydrothermal parameters
1. Soil hydraulic conductivity K; thermal properties C, λ_{eff}
CondL_h, CondT_coeff, EfeCapCond
2. Vapor density, ρ_v; Vapor diffusivity, D_v; Vapor dispersivity, D_{Vg}
Density_V, CondV_DE, CondV_DVg
3. Transport coefficient for adsorbed liquid flow (heat of wetting), D_{TD}
CondL_Tdisp
4. Dry air density, ρ_{da}; Gas conductivity, k_g
Density_DA, Condg_k_g
Solve balance equations for φ, T, P_g
CPLD = 1
NO
YES
BCD
Solve water & energy balance equation
Diff_Moisture_Heat
AIR = 1 ?
YES
ACD-air
Solve dry air balance equation
Air_sub
NO
ACD
Solve water balance equation
h_sub
Solve energy balance equation
Enrgy_sub

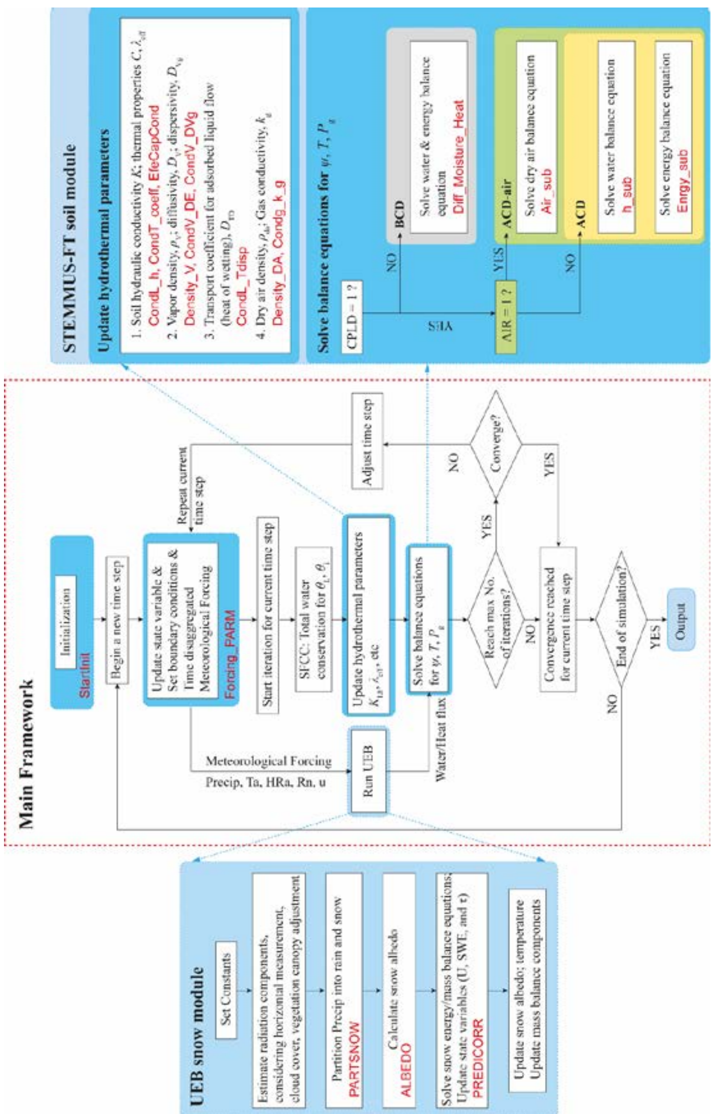
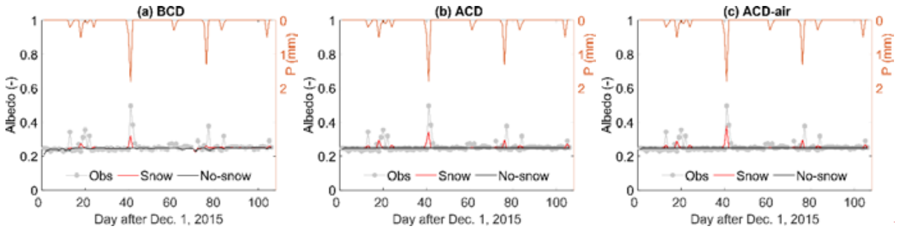


Figure 1. The overview of the coupled STEMMUS-FT and UEB model framework and model structure. SFCC is soil freezing characteristic curve; θ_L and θ_i are soil liquid water and ice content; K_{Lh} is soil hydraulic conductivity; λ_{eff} is thermal conductivity. ψ, T, P_g are the state variables for soil module STEMMUS-FT (matric potential, temperature, and air pressure, respectively). U, SWE, and τ are the state variables for snow module UEB (snow energy content, snow water equivalent, and snow age, respectively). UEB, Utah Energy Balance module. Precip, Ta, HRa, Rn, and u are the meteorological inputs (precipitation, air temperature, relative humidity, ~~radiation~~radiation, and wind speed). Model subroutines are in red fonts.



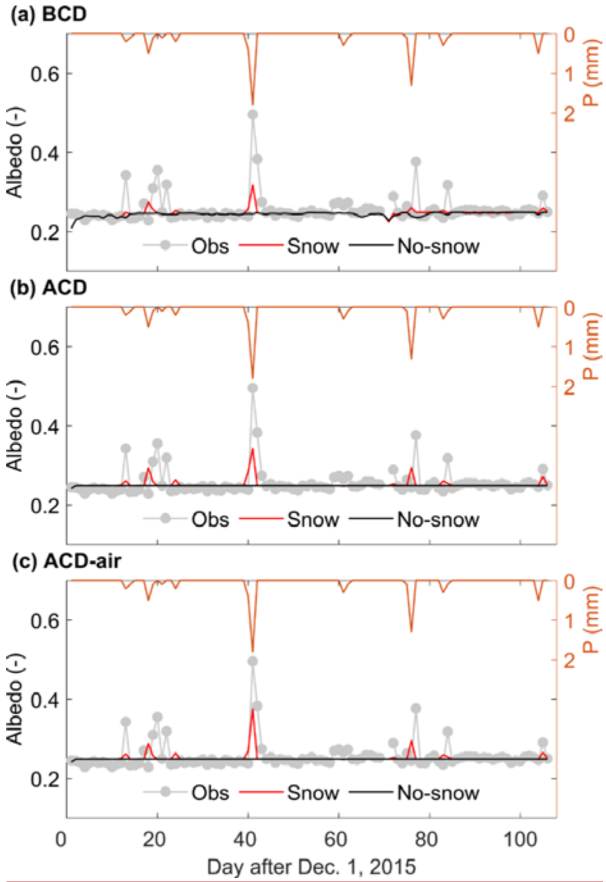


Figure 2. Time series of observed and model simulated daily average albedo using (a) BCD, (b) ACD, and (c) ACD-air soil model with/without consideration of snow module, with the precipitation.

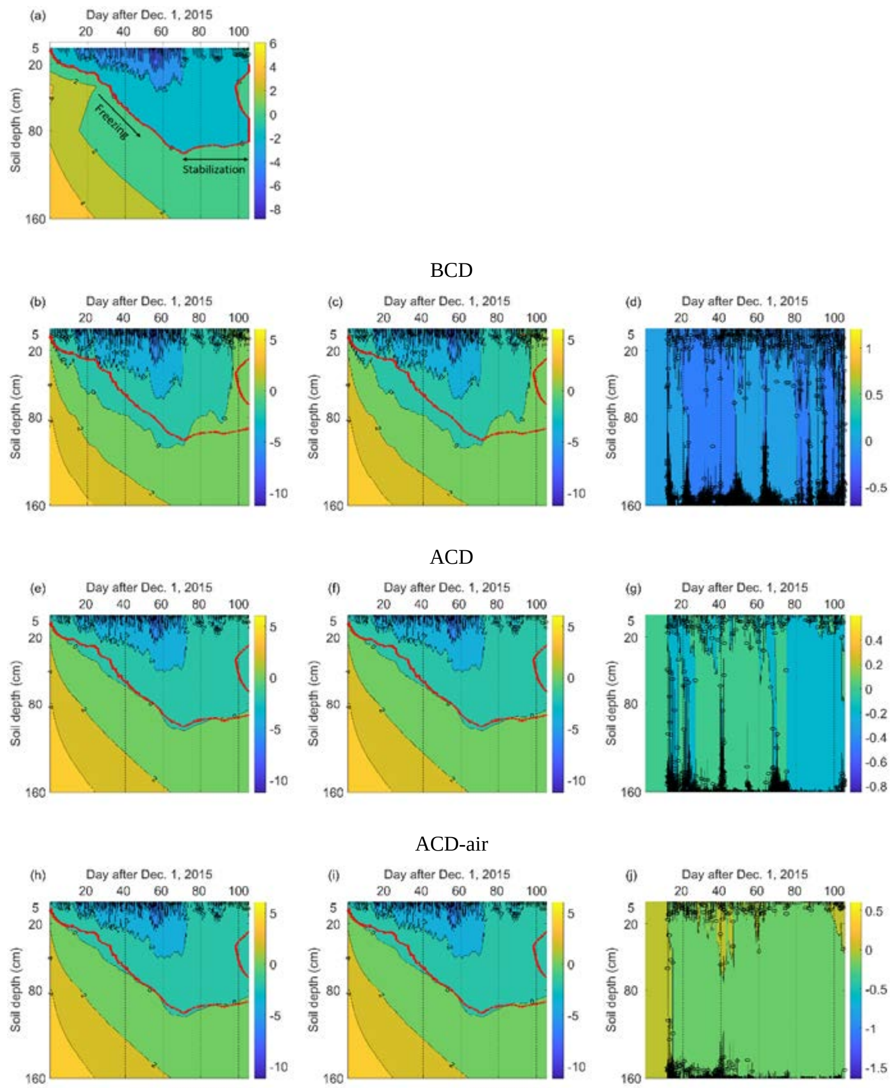


Figure 3. The spatial and temporal dynamics of observed (a) and simulated soil temperature using BCD, ACD, and ACD-air soil model, with and without consideration of snow module (Snow: b, e, h and No-Snow: c, f, i) and the difference (d, g, j) (simulations with snow minus simulations without snow). The red line indicates the ~~zero~~ zero-degree isothermal line (ZDIL) from the measured soil temperature. The observed soil freezing stage and stabilization stage was marked in Fig. 3a.

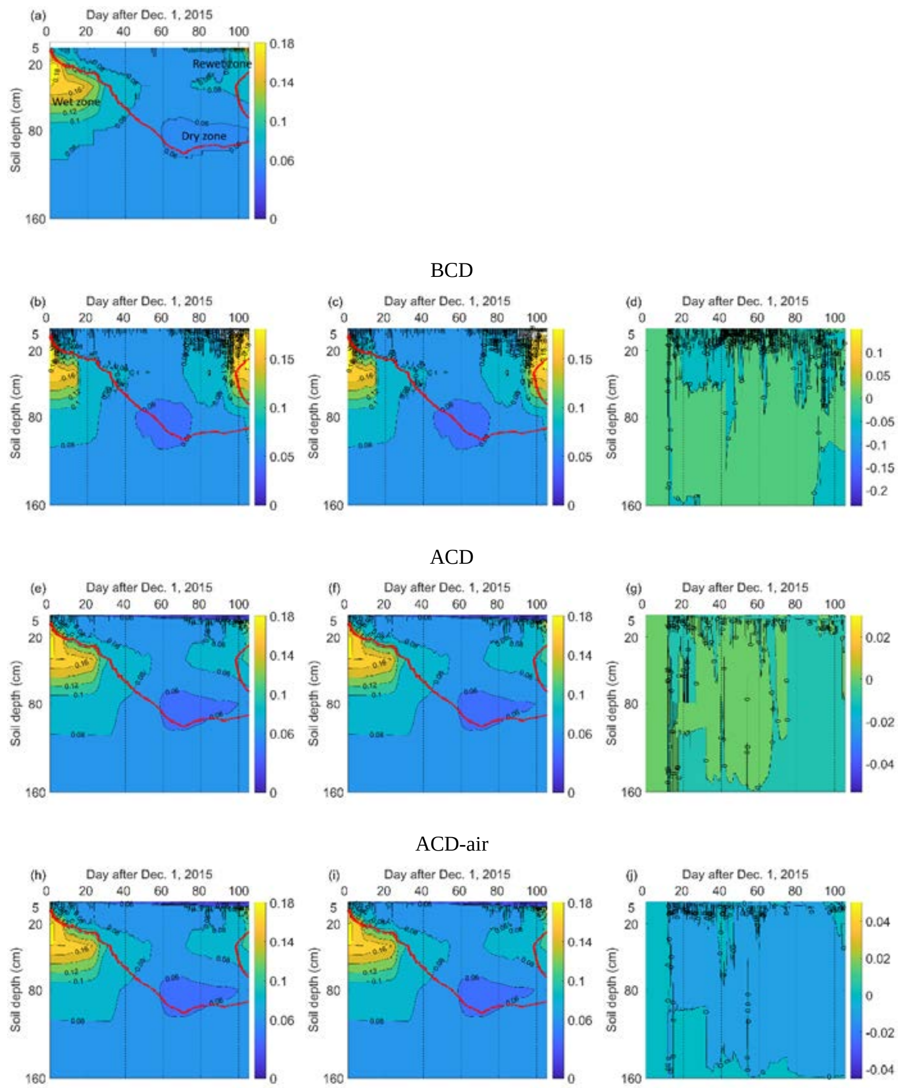


Figure 4. The spatial and temporal dynamics of observed (a) and simulated soil volumetric water content using BCD, ACD, and ACD-air soil model, with and without consideration of snow module (Snow: b, e, h and No-Snow: c, f, i) and the difference (d, g, j) (simulations with snow minus simulations without snow). The red line indicates the ~~zero~~ zero-degree isothermal line from the measured soil temperature. The observed wet zone, dry zone and rewet zone of soil moisture was indicated in Fig. 4a.

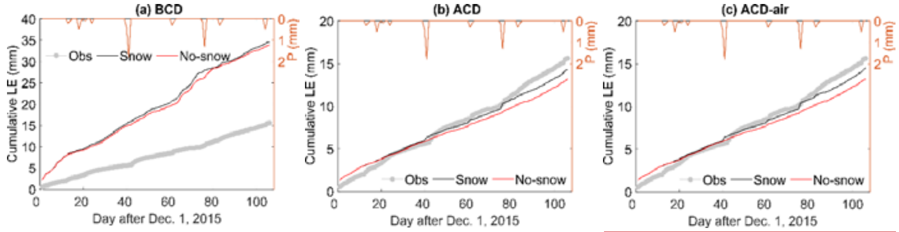
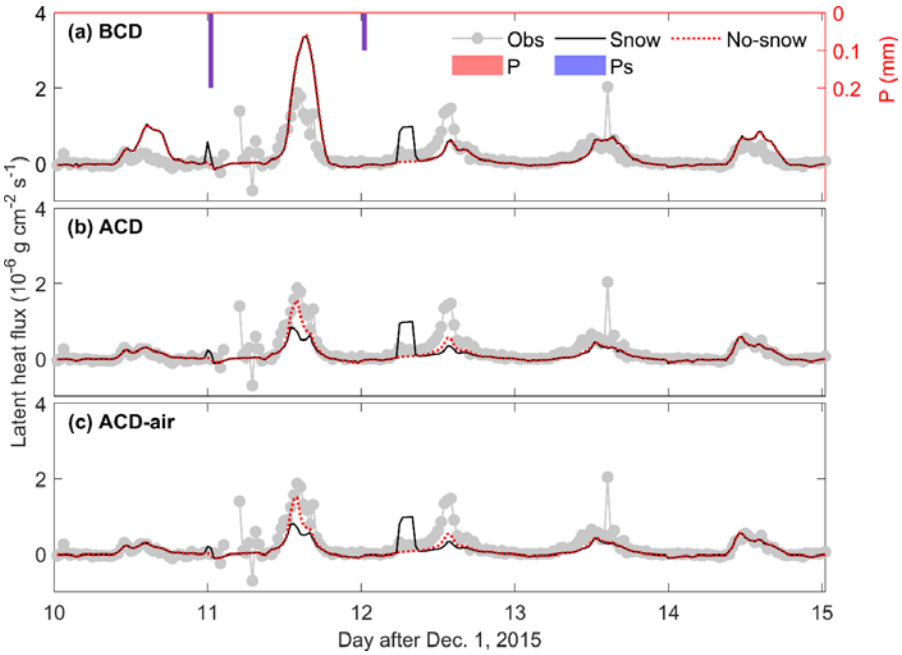


Figure 5. Time series of observed and model simulated surface cumulative latent heat flux (LE) using (a) BCD, (b) ACD, and (c) ACD-air soil model with/without consideration of snow module, with the precipitation.



940 **Figure 6. Observed and model simulated latent heat flux, using (a) BCD, (b) ACD, and (c) ACD-air soil model with/without snow module, of a typical five-day freezing period (from 10th to 14th Days after Dec. 1. 2015). P is the precipitation and Ps is the snowfall. All precipitation is in the form of snowfall.**

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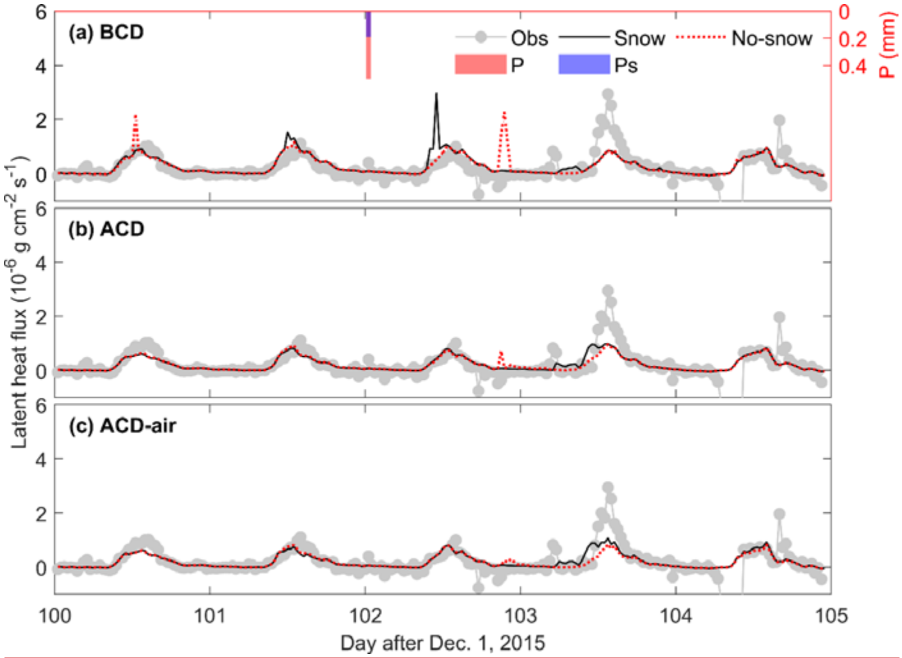
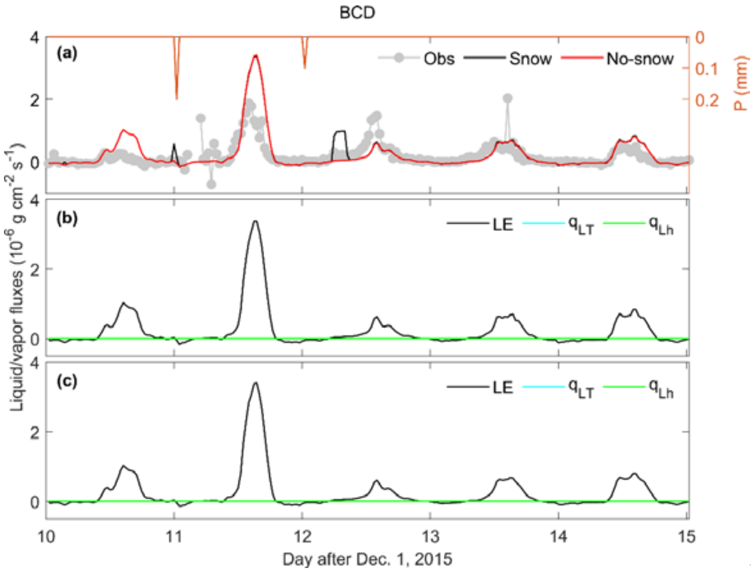
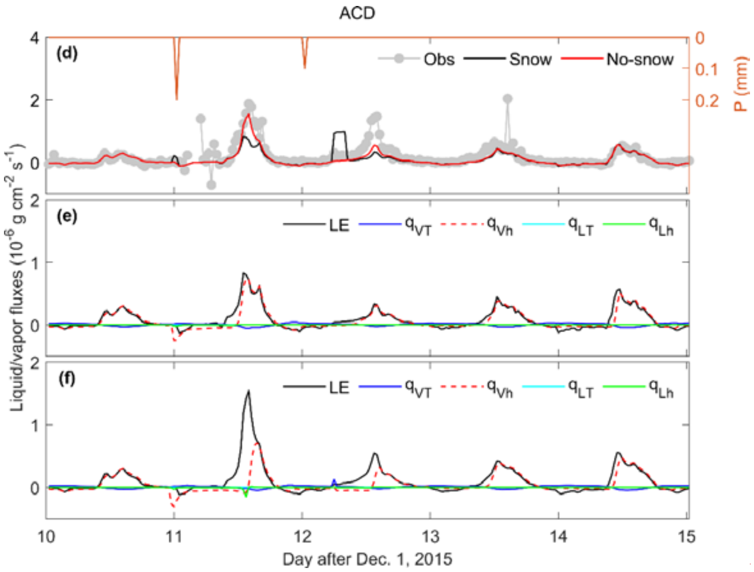
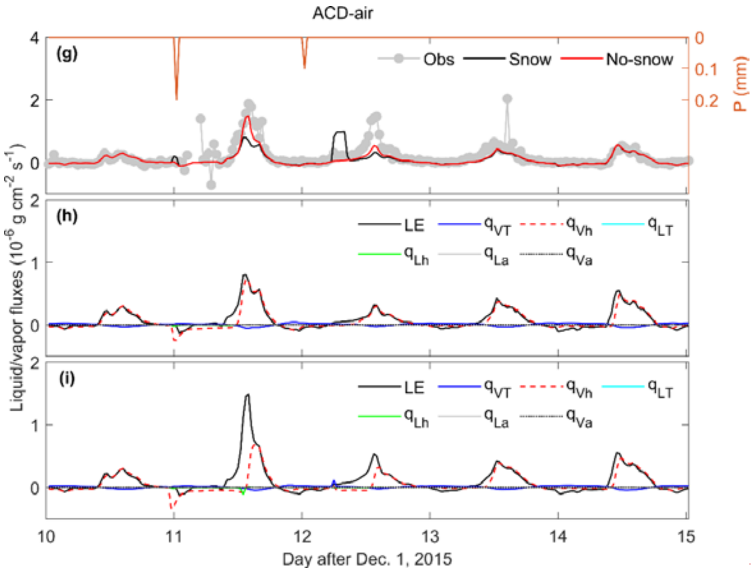


Figure 7. Observed and model simulated latent heat flux, using (a) BCD, (b) ACD, and (c) ACD-air soil model with/without snow module, of a typical five-day thawing period (from 100th to 104th Days after Dec. 1. 2015). P is the precipitation and Ps is the snowfall.




ACD
(d)
Obs
Snow
No-snow
P (mm)
(e)
LE
q_{VT}
q_{Vh}
q_{LT}
q_{Lh}
(f)
LE
q_{VT}
q_{Vh}
q_{LT}
q_{Lh}
Liquid/vapor fluxes (10^{-6} g cm^{-2} s^{-1})
Day after Dec. 1, 2015


ACD-air
(g)
Obs
Snow
No-snow
P (mm)
Liquid/vapor fluxes (10^{-6} g cm^{-2} s^{-1})
(h)
LE
q_{VT}
q_{Vh}
q_{LT}
q_{Lh}
q_{La}
q_{Va}
(i)
Day after Dec. 1, 2015

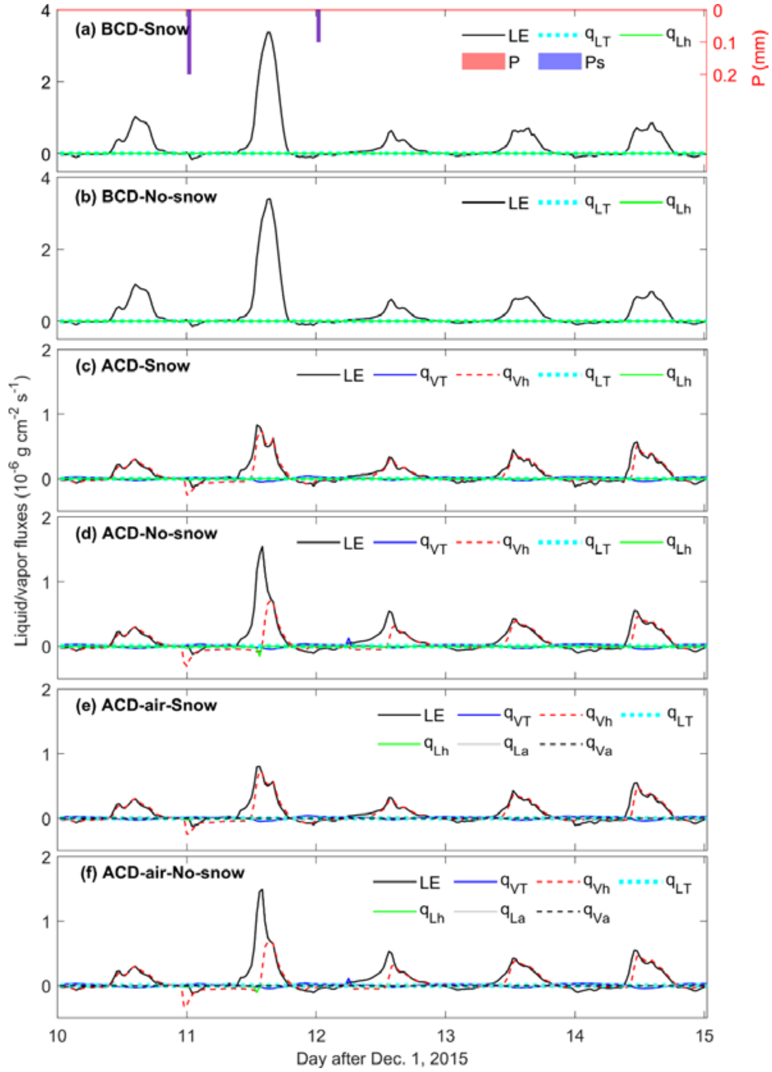
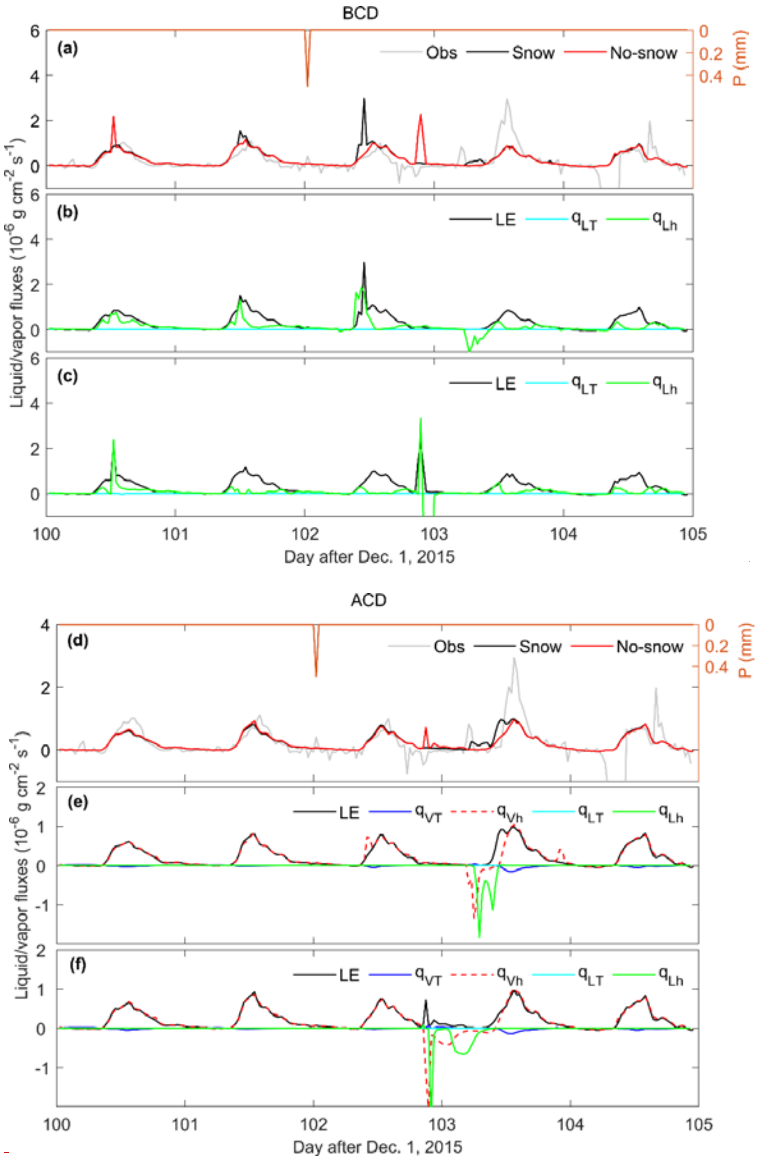
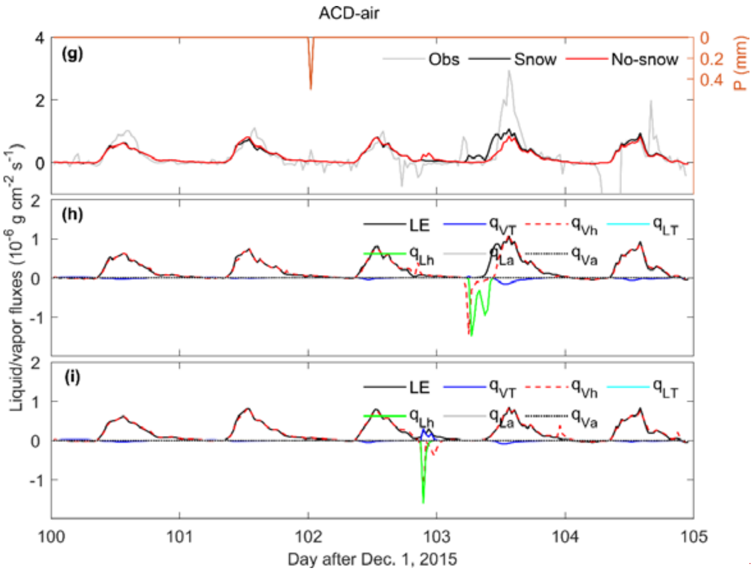


Figure 8. Model simulated latent heat flux and surface soil (0.1cm) thermal and isothermal liquid water and vapor fluxes (LE, q_{VT}, q_{Vh}, q_{LT}, q_{Lh}, q_{La}, q_{Va}), with and without snow module, of a typical five-day freezing period (from 10th to 14th Days after Dec. 1. 2015). a, c, and e are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-Snow, ACD-Snow, and ACD-air-Snow model, respectively. b, d, and f are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-No-Snow, ACD-No-Snow, and ACD-air-No-Snow model, respectively. LE is the latent heat flux, q_{VT}, q_{Vh} are the water vapor fluxes driven by temperature and matric potential gradients, q_{LT}, q_{Lh} are the liquid water fluxes driven by temperature and matric potential gradients, q_{La}, q_{Va} are the liquid and vapor water fluxes driven by air pressure gradients. Positive/negative values indicate upward/downward fluxes. Note that the surface LE fluxes without snow sublimation were presented. P is the precipitation and Ps is the snowfall. All precipitation is in the form of snowfall.~~Figure 6. Observed latent heat flux and simulated latent heat flux and surface soil (0.1cm) thermal and isothermal liquid water and vapor fluxes~~

~~**(LE, q_{VT}, q_{Vh}, q_{LT}, q_{Lh}, q_{La}, q_{Va}) with snow module and without snow module of a typical five-day freezing period (from 10th to 14th Days after Dec. 1. 2015). a, d, g are the comparison results of LE for BCD, ACD, and ACD-air soil model with/without snow module, respectively; b, e, h are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-Snow, ACD-Snow, and ACD-air-Snow model, respectively. c, f, i are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-No-Snow, ACD-No-Snow, and ACD-air-No-Snow model, respectively. LE is the latent heat flux, q_{VT}, q_{Vh} are the water vapor fluxes driven by temperature and matric potential gradients, q_{LT}, q_{Lh} are the liquid water fluxes driven by temperature and matric potential gradients, q_{La}, q_{Va} are the liquid and vapor water fluxes driven by air pressure gradients. Positive/negative values indicate upward/downward fluxes. Note that the surface LE fluxes without snow sublimation were presented in e & h.**~~



BCD
(a)
Obs
Snow
No-snow
P (mm)
(b)
LE
q_{LT}
q_{Lh}
(c)
Liquid/vapor fluxes (10^{-6} g cm^{-2} s^{-1})
Day after Dec. 1, 2015
ACD
(d)
(e)
q_{VT}
q_{Vh}
(f)


ACD-air
(g)
Obs
Snow
No-snow
P (mm)
0
0.2
0.4
4
2
0
(h)
LE
q_{VT}
q_{Vh}
q_{LT}
q_{Lh}
q_{La}
q_{Va}
2
1
0
-1
(i)
Liquid/vapor fluxes (10^{-6} g cm^{-2} s^{-1})
100
101
102
103
104
105
Day after Dec. 1, 2015

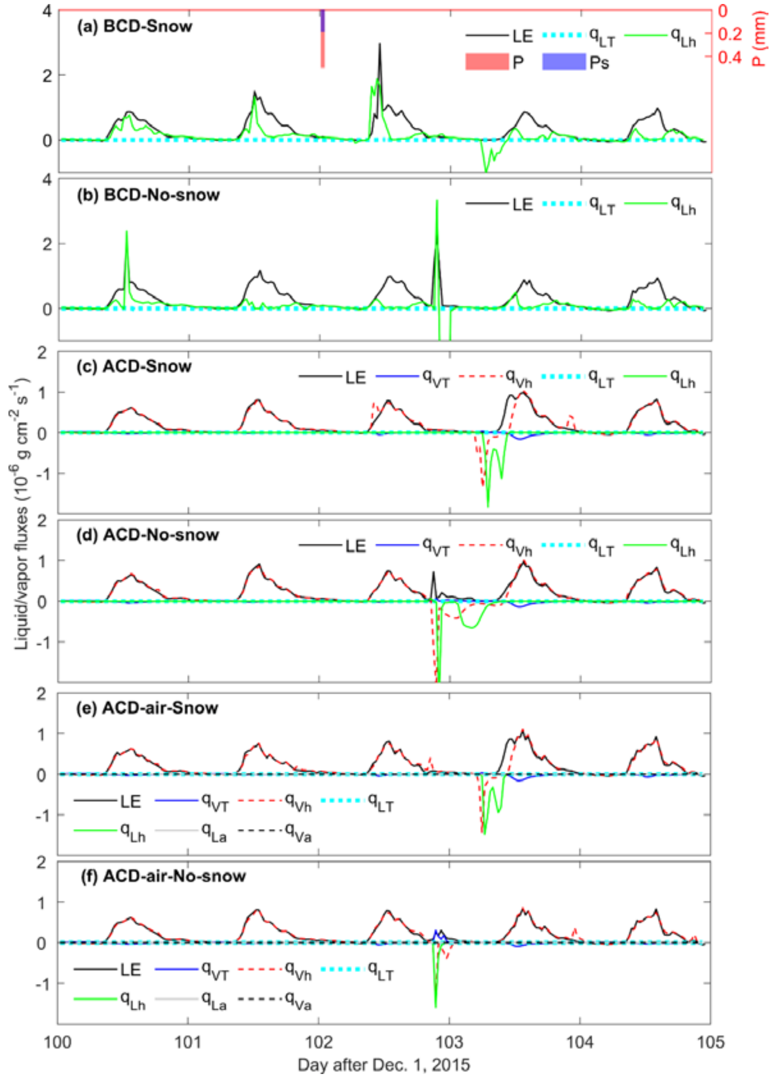


Figure 9. Model simulated latent heat flux and surface soil (0.1cm) thermal and isothermal liquid water and vapor fluxes (LE, q_{VT}, q_{Vh}, q_{LT}, q_{Lh}, q_{La}, q_{Va}) using BCD (a, b), ACD (c, d), and ACD-air (e, f) simulations with and without snow module, respectively, during the typical 5-day thawing periods (from 100th to 104th Days after Dec. 1. 2015). a, c, and e are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-Snow, ACD-Snow, and ACD-air-Snow model, respectively. b, d, and f are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-No-Snow, ACD-No-Snow, and ACD-air-No-Snow model, respectively. LE is the latent heat flux, q_{VT}, q_{Vh} are the water vapor fluxes driven by temperature and matric potential gradients, q_{LT}, q_{Lh} are the liquid water fluxes driven by temperature and matric potential gradients, q_{La}, q_{Va} are the liquid and vapor water fluxes driven by air pressure gradients. Positive/negative values indicate upward/downward fluxes. Note that the surface LE fluxes without snow sublimation were presented. P is the precipitation and Ps is the snowfall.~~**Figure 7. Observed latent heat flux and model simulated latent heat flux and**~~

~~surface soil (0.1cm) thermal and isothermal liquid water and vapor fluxes (LE, q_{VT}, q_{Vh}, q_{LT}, q_{Lh}, q_{La}, q_{Va}) using BCD (a, b, c), ACD (d, e, f), and ACD-air (g, h, i) simulations with and without snow module, respectively, during the typical 5-day thawing periods (from 100th to 105th Days after Dec. 1. 2015). a, d, g are the comparison results of LE for BCD, ACD, and ACD-air soil model with/without snow module, respectively; b, e, h are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-Snow, ACD-Snow, and ACD-air-Snow model, respectively. c, f, i are the surface soil thermal/isothermal liquid water and vapor fluxes simulated by BCD-No-Snow, ACD-No-Snow, and ACD-air-No-Snow model, respectively. LE is the latent heat flux, q_{VT}, q_{Vh} are the water vapor fluxes driven by temperature and matric potential gradients, q_{LT}, q_{Lh} are the liquid water fluxes driven by temperature and matric potential gradients, q_{La}, q_{Va} are the liquid and vapor water fluxes driven by air pressure gradients. Positive/negative values indicate upward/downward fluxes. Note that the surface LE fluxes without snow sublimation were presented in e & h.~~

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