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The effect of soil on the summertime surface energy budget of a humid

Subarctic tundra in northern Quebec, Canada

Georg Lackner*

Department of Civil and Water Engineering, Université Laval, Quebec City, Quebec, Canada

Centre d'études nordiques (CEN), Université Laval, Quebec City, Quebec, Canada

Takuvik Joint International Laboratory, Université Laval (Canada) and CNRS-INSU (France),

Quebec City, Quebec, Canada

CentrEau - Water Research Center, Université Laval, Quebec City, Quebec, Canada

Daniel F. Nadeau

Department of Civil and Water Engineering, Université Laval, Quebec City, Quebec, Canada

CentrEau - Water Research Center, Université Laval, Quebec City, Quebec, Canada

Florent Domine

Department of Chemistry, Université Laval, Quebec City, Quebec, Canada

Department of Geography, Université Laval, Quebec City, Quebec, Canada

Centre d'études nordiques (CEN), Université Laval, Quebec City, Quebec, Canada

Takuvik Joint International Laboratory, Université Laval (Canada) and CNRS-INSU (France),

Quebec City, Quebec, Canada

Annie-Claude Parent

Department of Civil and Water Engineering, Université Laval, Quebec City, Quebec, Canada

Gonzalo Leonardini

Department of Civil and Water Engineering, Université Laval, Quebec City, Quebec, Canada

Aaron Boone

CNRM UMR 3589, Météo-France/CNRS, Toulouse, France

François Anctil

Department of Civil and Water Engineering, Université Laval, Quebec City, Quebec, Canada

CentrEau - Water Research Center, Université Laval, Quebec City, Quebec, Canada

Vincent Fortin

*Meteorological Research Division, Environment and Climate Change Canada, Dorval, Quebec,
Canada*

**Corresponding author: Georg Lackner, georg.lackner@mailbox.org*

ABSTRACT

Rising temperatures in the southern Arctic region are leading to shrub expansion and permafrost degradation. The objective of this study is to analyse the surface energy budget (SEB) of a subarctic shrub tundra site that is subject to these changes, on the east coast of Hudson Bay in eastern Canada. We focus on the turbulent heat fluxes, as they have been poorly quantified in this region. This study is based on data collected by a flux tower using the eddy covariance approach and focused on snow-free periods. Furthermore, we compare our results with those from 6 Fluxnet sites in the Arctic region and analyse the performance of two land surface models, SVS and ISBA, in simulating soil moisture and turbulent heat fluxes. We found that 22.4% of the net radiation was converted into latent heat flux at our site, 33% was used for sensible heat flux and about 15% for ground heat flux. These results were surprising considering our site was by far the wettest site among those studied, and most of the net radiation at the other Arctic sites was consumed by the latent heat flux. We attribute this behaviour to the high hydraulic conductivity of the soil (littoral and intertidal sediments), typical of what is found in the coastal regions of the eastern Canadian Arctic. Land surface models overestimated the surface water content of those soils but were able to accurately simulate the turbulent heat flux, particularly the sensible heat flux and to a lesser extent, the latent heat flux.

47 *Significance statement.* Isostatic uplift after the last deglaciation led to the emergence of beaches,
48 which represent a large area in the Canadian Arctic. We studied the surface energy budget of
49 such a beach emerged 6000 years ago in northeastern Canada. Results suggest that evaporation
50 is up to 20% less than at previously studied sites which is explained by sandy soils which retain
51 little moisture despite abundant precipitation. Deployed numerical models showed difficulties in
52 simulating the soil conditions but proved successful in simulating the surface energy budget after
53 manual adjustments of the soil conditions. These simulation difficulties probably apply to other
54 parts of the Canadian Arctic, possibly leading to errors in meteorological and climate forecasting.

55 **1. Introduction**

56 The increase in global air temperature, which is twice as pronounced in polar regions than in
57 all other parts of the world (Chylek et al. 2009), is changing the distribution of vegetation zones
58 (Myers-Smith and Hik 2018). This change is particularly notable in the forest-tundra ecotone
59 (Payette et al. 2001) at the interface between the boreal forest and the Arctic shrub tundra. This
60 region is characterized by a mosaic of forest patches, usually restricted to humid, wind-sheltered
61 areas, surrounded by shrub tundra that occupies the well-drained and wind-exposed sites. Rising
62 temperatures favor the progressive expansion of these forest patches, which lead to the northward
63 expansion of the boreal forest (Tape et al. 2006).

64 The disappearance of shrub tundra at the southern Arctic border is leading to changes on multiple
65 levels. Heat and moisture exchanges at the interface between the land surface and the atmosphere
66 are being altered. As such, the surface energy budget (SEB) as a whole is also changing due
67 to shifts in leaf area index, land surface albedo, and evapotranspiration. These changes then
68 create a feedback affecting the climate at multiple scales by modifying ecosystem processes and

69 weather patterns (Baldocchi and Ma 2013; Pielke et al. 2011). Due to the circumpolar extent of
70 the forest-tundra ecotone, its potential impact on the climate demands our attention.

71 Moreover, the surface energy budget influences the fate of the permafrost as it controls how much
72 heat is stored in or released from the ground. Permafrost in this region has become discontinuous
73 to sporadic and is degrading quickly (Lemieux et al. 2016). In order to model carbon emissions
74 due to permafrost thaw (Schuur et al. 2015), information on the partitioning of the net radiation
75 into sensible, latent, and soil heat fluxes is key. Thus, an improved understanding of the energy
76 balance can ultimately lead to more accurate predictions of permafrost thawing and the resulting
77 greenhouse gas emissions.

78 Due to the harsh conditions and limited accessibility in the Subarctic region, few in-situ obser-
79 vations have been conducted. Available datasets come from sites located in central and western
80 North America, Russia and Europe (Lafleur et al. 1992; Eugster et al. 2000; Langer et al. 2011;
81 Stiegler et al. 2016). The topography of these sites often consists of plains or gentle slopes, which
82 tend to reduce drainage efficiency and provide wetter soils. The latent heat flux at these sites tends
83 to dominate the SEB over the other terms. For example, Eugster et al. (2000) reported latent heat
84 flux to net radiation ratios in the range of 0.44 to 0.72 at several low Arctic sites in Alaska, and
85 values between 0.48 to 0.65 for treeline shrub tundra across central Canada, Alaska, and Russia.
86 One aspect that has so far been poorly documented is the impact of soil properties on the SEB
87 of subarctic sites, as the vegetation tended to be considered the dominant factor on the issue. In
88 the eastern region of the Canadian Subarctic, for instance, soils are often composed of coarse
89 quaternary deposits such as glacial till and marine sediments (Lemieux et al. 2016). These soils
90 are usually very well-drained even on small inclines, limiting water availability and making this
91 region distinct from those of previous studies.

92 In addition to the quantification of surface fluxes, a pressing issue is whether current land surface
93 models (LSMs) such as ISBA (Noilhan and Planton 1989) and SVS (Husain et al. 2016; Alavi et al.
94 2016) are able to adequately simulate the SEB in Subarctic regions. LSMs are used in a variety
95 of applications, such as numerical weather prediction models and terrestrial ecosystem models.
96 Although the surface of the forest-tundra environment is snow-covered for most of the year, the
97 adequate modeling of the SEB during snow-free period is equally important because this period
98 regulates the growth of vegetation and to a larger extent, the water cycle. Studies have highlighted
99 difficulties in modeling dry sites, stating that the models capture energy limitations better than
100 water limitations (Engstrom et al. 2006), thus leading to increased errors for simulated latent heat
101 fluxes.

102 In this study, we use eddy covariance data to analyse the Arctic surface energy balance and
103 examine the performance of two LSMs: SVS and ISBA. SVS will soon replace the old version
104 of ISBA and be used as an operational weather-forecasting model in Canada. This LSM has
105 been proven to be competent at simulating latent and sensible heat fluxes at six sites in arid,
106 Mediterranean, and tropical climates (Leonardini et al. 2020). This makes it particularly interesting
107 to compare whether SVS performs better in simulating northern Canadian regions than the current
108 version of ISBA. This study presents a comprehensive analysis of observed and modeled surface
109 energy fluxes during three consecutive snow-free periods in a forest-tundra ecotone site in northern
110 Quebec in eastern Canada. The objectives of this study are to (i) use eddy covariance observations
111 to identify key factors that influence the energy partitioning at the surface, (ii) determine the
112 interannual variability of the partitioning, (iii) contrast the observations with those from other
113 Subarctic sites across the globe, and (iv) assess the capability of the current LSMs, ISBA and SVS,
114 to simulate surface heat fluxes in this environment.

2. Methods

a. Study Site

The study site is located in the Tasiapik valley close to the community of Umiujaq, on the eastern shore of Hudson Bay in Quebec, Canada (56°33'31"N, 76°28'56"W; Umiujaq in Figure 1a). The valley is 4.5 km long and 1.3 km wide and surrounded by *cuestas* (steep ridges) with a maximum height of nearly 400 m. The lower part of the valley borders the Tasiujaq Lake at an elevation of 0 m and is connected to Hudson Bay by an inlet. The vegetation cover in the valley is typical of a forest-tundra ecotone, with shrub and some lichen tundra in its upper region and forest tundra in the lower region. While the dwarf birch *Betula glandulosa* is the dominant vegetation type in the shrub tundra, the prevailing tree species in the forest-tundra are black and white spruces (*Picea mariana* [Mill.] BSP and *Picea glauca* [Moench] Voss) (Paradis et al. 2016). Following Gagnon et al. (2019), vegetation in the vicinity of the flux tower could be classified as one of three types: pure lichen cover, small shrubs with lichen understory and a low LAI (presumably <0.5) and small shrubs with moss understory. Lichens covers 20 to 30% of the surface while small shrubs with differing understory make up the rest (See Figure A3).

In the valley, permafrost is discontinuous to sporadic (Lemieux et al. 2016) with rapidly degrading permafrost mounds that are ≈ 100 m in diameter (Fortier and Aubé-Maurice 2008). Close to the study site, permafrost is only present to the south at a distance of about 20 m. The active layer thickness is about 1.5 to 2 m. This active layer depth together with the small inclination of a few degrees of the surface presumably provides a rather undisturbed drainage of precipitation. There are no complete long-term meteorological records available for Umiujaq. Instead, recent observations from the research station in the Tasiapik valley were used to estimate climatological conditions starting in 2012 for air temperature and from 2015 for precipitation. Umiujaq exhibits a

Subarctic climate with a mean annual temperature of -4.0°C . Their winters are long and cold and their summers are cool. There is usually continuous snow cover from late October to early June with a maximum thickness between 0.6 m and 1.0 m. Monthly temperatures and precipitation are shown in Figure 2.

Weather patterns are strongly influenced by Hudson Bay to the west and whether or not it is covered with ice. Hudson Bay usually remains frozen until mid-June, and then stores heat in summer that is released in the fall. This results in relatively lower air temperatures in summer and higher air temperatures in the fall. The average total annual precipitation is about 800 - 1000 mm, out of which about 50% falls as snow and the precipitation cycle shows a strong correlation with the freezing of Hudson Bay. After the freeze-up in mid to late December, precipitation remains low from January to May/June. Precipitation subsequently increases and peaks in late summer and autumn. Fog is common throughout the summer (Robichaud and Mullock 2001). Advection fog forms when warmer air moves over the cold Hudson Bay, mostly in July and August.

The geology of the valley is thoroughly described in Lemieux et al. (2020). The bedrock is covered by a 10- to 50-m thick layer of moraine deposits and fluvio-glacial sediments. Above the fluvio-glacial sediments, there is a layer of marine sediments (mainly silt) that is $\approx 30\text{-m}$ thick, but it does not cover the entire valley. The topsoil layer is sand (littoral and intertidal sediments, see Figure 3), with some limited exceptions.

Gagnon et al. (2019) found that the soil covered with lichen is made up almost entirely of sand (90% for the first 5 cm and almost 100% for deeper layers up to 30 cm). They reported a soil organic content of about 1.4 kg m^{-2} under lichen and 4.3 kg m^{-2} for medium height shrubs tundra with lichen understory. Most of the soil organic content is found in the first 2 cm for the lichen covered areas and 5 cm for medium shrub covered areas.

b. Instrumental Setup

In June 2017, a 10-m tower was equipped with a sonic anemometer and a CO₂/H₂O gas analyzer that was installed 4.2 m above the ground (IRGASON, Campbell Scientific, USA), facing eastwards towards a plateau with a small slope of about 3° (Figure 1 b). The tower holding the instrument and a steep ridge are to the west, in the opposite direction.

The tower also features general meteorological sensors, such as temperature and relative humidity (model HMP45, Vaisala, Finland) and wind direction/speed (model 05103, R.M. Young, USA) probes. Some 10 m west of the flux tower, another meteorological station was installed in 2012, which also measures basic environmental variables such as air temperature and wind speed. It is equipped with a 4-component radiometer (CNR4, Kipp and Zonen, The Netherlands). In the proximity of the two stations, soil temperature and water content (5TM, Meter Group, USA) are measured at several locations and under various vegetation types directly at the tower and several meters distance, at 5 depths from 5 cm to 50 cm, in order to capture spatial variability. Precipitation is also measured next to the tower with a T200B precipitation gauge (GEONOR, USA) equipped with a single Alter shield. Raw precipitation observations were corrected for wind-induced undercatch following Kochendorfer et al. (2018). Flux data was recorded using a CR3000 datalogger while the other data was recorded using a CR1000 datalogger (both Campbell Scientific, USA). The site was visited twice a year for maintenance and data collection.

c. Data Processing

Raw 10-Hz eddy-covariance data were processed using the EddyPro® (version 7.0.3; Li-COR Biosciences, USA) software package. This software includes multiple corrections and QA/QC procedures such as turbulent fluctuation detrending based on a running mean, covariance maximization, density fluctuation compensation (Webb et al. 1980), and analytic correction of high-pass

and low-pass filtering effects (Moncrieff et al. 1997). Turbulence data were rotated into a stream-line coordinate system using the sector-wise planar-fit method (Wilczak et al. 2001). The random uncertainty quantification approach from Finkelstein and Sims (2001) was used to detect outliers, spikes, and other artifacts. The 0-1-2 quality scheme from Mauder et al. (2013) was applied to the data, and only segments that were flagged as 0 and 1 were kept in the study. These values correspond to good- and medium-quality observations, which are both suitable for the analysis of seasonal energy budgets.

Ground heat flux was computed using a time series of a measured soil temperature profile $T(z_1, t)$, $T(z_2, t)$ and $T(z_3, t)$ at three depths $z_1 < z_2 < z_3$ and the one-dimensional heat equation

$$\frac{\partial T(t, z)}{\partial t} = D \frac{\partial^2 T(t, z)}{\partial z^2}, \quad (1)$$

where D is the thermal diffusivity (Westermann et al. 2009). As the thermal diffusivity is defined by the thermal conductivity K and the heat capacity c_h by $D = K/c_h$ we calculated D through K and c_h using empirical equations which are based on the soil density and water content (Chen 2008; Bittelli et al. 2015). Then, the heat equation was solved numerically in Python using two of the temperature sensors in the soil ($T(z_1, t)$ and $T(z_3, t)$) as Dirichlet boundary conditions. The result is a temperature distribution in the considered spatial domain. It was checked against the third soil temperature measurement $T(z_2, t)$ for correctness ($R^2 = 0.99$, $\text{RMSE} = 0.21^\circ\text{C}$). Then, the temperature gradient at the upper boundary z_1 was calculated. The ground heat flux at depth z_1 is then given by the Fourier equation:

$$G_{z_1} = K \left. \frac{\partial T(z, t)}{\partial z} \right|_{z=z_1} \quad (2)$$

The heat storage change G_s in the soil is accounted for by taking the change of the mean temperature of the soil above z_1 . The mean temperature is obtained by extrapolating the temperature distribution given by the heat equation. The final ground heat flux is then given by $G = G_{z_1} + G_s$. This method

was chosen as measurements under different vegetation types were available making it possible to analyze the spatial variability. Furthermore, problems have been reported of alternative methods such as using heat flux plates (Ochsner et al. 2006).

Post-processing was also necessary to sort out the remaining outliers and to fill data gaps. This was achieved using the program PyFluxPro (Isaac et al. 2017). This program has six processing levels, uses EddyPro output files as input and produces a continuous time series for all fluxes. For the first three processing levels, data are read, quality controlled, and finally auxiliary measurements are merged when gaps are present. The quality control includes (i) range checks based on user-defined limits, (ii) spike detection, (iii) manual removal for specific dates and (iv) data rejection based on other variables. For the data rejection, CO₂ and H₂O signal strengths from the IRGA as well as internal error codes from both the sonic anemometer and the IRGA were used to remove erroneous flux data. For the fourth processing level, meteorological variables were gap filled with ERA5 data, a reanalysis product from the European Centre for Medium-Range Weather Forecast. These data provide hourly estimates for various meteorological and soil variables beginning in 1979 at a spatial resolution of 30 km (<https://www.ecmwf.int/en/forecasts/datasets/reanalysis-datasets/era5>). Each variable was bias-corrected using a linear fit between ERA5 and flux tower observations during periods when both were available.

Finally, the fluxes were gap-filled using a self-organizing linear output map (SOLO), a type of artificial neural network (ANN) (see Hsu et al. (2002) and Abramowitz (2005)). SOLO requires a set of environmental drivers such as air temperature, radiation and humidity, as well as the fluxes as inputs. SOLO first constructs relationships between the environmental drivers by applying an ANN-equivalent of a principal component analysis. It then uses an ANN-equivalent of a multiple linear regression to make connections between the drivers and the fluxes. ANN together with

marginal distribution sampling (MDS, Reichstein et al. (2005)) were shown to be the best choices for gap-filling flux data (Moffat et al. 2007).

On average, 56.1% of the hourly values of the sensible heat flux (2017: 50.8%; 2018: 63.5%; 2019: 54.2%) and 47.8% of the latent heat flux (2017: 46.7%; 2018: 54.2%; 2019: 42.4%) passed all quality control steps.

This study focused on the summer period, defined here as snow-free periods. It started on the date that snow disappeared and ended when the first snowfall occurred in order to avoid any effects associated with the presence of snow on the ground. In order to facilitate inter-annual comparisons, we applied these criteria to the three measurement years and defined common start and end dates: July 1st to September 5th (66 days).

d. Comparison Sites

To give the energy balance measurements from the Umiujaq site context, data from six other flux tower sites across the Arctic were selected from the FLUXNET (Pastorello et al. 2020) database. These sites were located in Canada, the United States, and Russia and were selected within latitudes ranging from 56°N to 70°N (Figure 4). FLUXNET is a global network of micrometeorological tower sites that collects eddy covariance observations. We have chosen sites that show some similarities concerning the vegetation (shrubland) and are not considered as permanent wetland.

The comparison sites cover a variety of biomes located in Arctic and Subarctic regions (Figure 4). Their characteristics are detailed in Table 1. Mean annual temperatures range from −14.3°C in eastern Russia (RU-COK), by far the coldest site, to −1°C in Alaska (US-EML). Mean annual precipitation varies considerably, from the dry Alaskan site (122 mm; US-AN) to more humid sites in western Russia (538 mm; RU-VRK) and eastern Canada, with Umiujaq being by far the most humid site (800 - 1000 mm).

251 Vegetation covers at all the sites, except for RU-COK, is classified as open shrubland. Those
252 sites are covered by all evergreen or deciduous woody vegetation less than 2 m tall at coverage
253 between 10% and 60%. The vegetation at RU-COK is closed shrubland, given that the canopy
254 cover exceeds 60%.

255 Data processing was similar to that described in section 2c. Quality assurance and quality control
256 methods follow those in Pastorello et al. (2014), where meteorological variables were gap-filled
257 using ERA-Interim reanalysis data (Vuichard and Papale 2015), and flux data were gap-filled with
258 marginal distribution sampling (Reichstein et al. 2005).

259 As for most of the eddy covariance sites worldwide, all studied sites experienced non-closure of
260 the energy budget. Since not all terms of the energy budget were available for each site and given
261 the ongoing debate on how to best distribute the residual energy budget (De Roo et al. 2018), no
262 corrections were applied to any of the energy budgets.

263 *e. Model Description*

264 To simulate the SEB, two LSMs were used on offline mode: ISBA and SVS. ISBA was included
265 in SURFEX v8.1 (SURFace EXternalisée), a modeling platform developed by Météo-France
266 (<http://www.umr-cnrm.fr/surfex/>) that comprises several other models for various types of
267 land and water surfaces. SVS is a surface model built by Environment and Climate Change Canada
268 for operational hydrometeorological applications.

269 Both models require meteorological data as inputs. These data can be obtained from an atmo-
270 spheric model when operated in coupled mode, or from observations in stand-alone mode (offline).
271 In this study, the stand-alone option was used to ensure the quality of the driving data and to avoid
272 errors that may be introduced by the atmospheric model itself.

1) ISBA-MEB

ISBA (Noilhan and Planton 1989) simulates all water and energy exchanges between the atmosphere, vegetation, and soil. The water and energy budgets of soil are computed by solving the one-dimensional Fourier law and a mixed-form of the Richards equation explicitly (Boone et al. 2000; Decharme et al. 2011). The soil is therefore divided into layers of increasing depth down to 12 m. Each layer has specific hydraulic and thermal properties, which are either calculated internally based on the soil texture or forced externally. The phase change between the solid and liquid phases of water in the soil is computed using the Gibbs free-energy method (Boone et al. 2000; Decharme et al. 2016). This method calculates the temperature that governs the phase change based on the soil texture. During any phase-change, the total amount of water in each layer remains unchanged. Therefore when the ice content increases, the water content decreases by the same amount, and vice versa.

Vegetation characteristics can be automatically selected from 19 available categories using the site coordinates and ECOclimap (<https://opensource.umr-cnrm.fr/projects/ecoclimap-sg/wiki>), or they can be specified by the user (vegetation type, LAI, canopy height and albedo).

In order to specifically include vegetation in ISBA, Boone et al. (2017) recently added what is called the multi-energy balance (MEB) parametrization. MEB largely follows the representation of vegetation detailed in the Rossby Centre Regional Climate Model (Samuelsson et al. 2011). The most important additions (relative to the standard version of ISBA) are the presence of canopy turbulence, a new radiation transfer scheme, and a litter layer. Within MEB, the sensible and latent heat fluxes between multiple components are calculated separately and then combined to form the

295 final heat flux. These components are the vegetation, the bare ground, a ground-based snowpack,
296 and the canopy air space.

297 2) SVS

298 The Soil, Vegetation and Snow scheme (SVS) (Husain et al. 2016; Alavi et al. 2016) is partly
299 based on an early version of ISBA (Bélair et al. 2003a,b), in which a force-restore method is used
300 for the energy budget, as well as a single layer interaction between the canopy and the atmosphere.
301 A notable improvement of this model is the different land surface tiling and vertical layering, as
302 well as having separate energy budgets for bare soil and (low and high) vegetation tiles. Another
303 difference between the two LSMs concerns soil hydrology. The earlier version of ISBA monitored
304 soil moisture at a superficial layer and the rooting-depth layer. In SVS, a certain soil water content
305 is attributed to each of the N soil layers and its evolution is modeled based on Darcian flow.
306 However, in SVS soil temperature is only defined at the soil surface and the associated variation is
307 determined using the single-layer approach mentioned above.

308 3) METEOROLOGICAL FORCING DATA

309 Both ISBA and SVS require the same input data: air temperature, specific humidity, wind speed,
310 incoming shortwave and longwave radiation, atmospheric pressure, and (solid and liquid) precipi-
311 tation rates. Observational data for these variables are available from 2012, except for atmospheric
312 pressure, for which the data start from June 2017 and precipitation where measurements started in
313 2015. Before this date, ERA5 atmospheric pressure and precipitation was used. A simple linear
314 regression was conducted between the measured data and the ERA5 data in order to correct for the
315 atmospheric pressure data from ERA5 and make it consistent with *in situ* observations. A fixed
316 threshold of 0°C was applied to separate the solid and liquid phases of precipitation. This simple

threshold was implemented because we were only studying the summer period, and thus a more elaborate phase partitioning strategy was unnecessary.

4) MODEL SETUP

Both models were set up in a similar fashion to maximize the comparability of the results. However, we have opted to use the soil discretization that is most commonly used for each model and which happen to be different from each other. For ISBA, the soil column has a depth of 12 m and is divided into 20 layers of increasing depth. The soil column for SVS is 10-m-deep and has 7 layers, also of increasing depth.

Models were initialized with a spin-up to ensure an equilibrium of soil moisture and temperature. We used 5 complete years of observations (2012-2017) as the spin-up, which was sufficient for both models to reach an equilibrium. We analyzed the melt and onset of the snow cover in the model to make sure that snow did not interfere with our study period. We used a time step of 1 hour for both models.

The soil composition that was used for both models was 95% sand and 5% silt (Gagnon et al. 2019). Hydraulic and thermal parameters were for the most part calculated by the models themselves, based on the soil composition provided. A modified version of both ISBA and SVS models were also tested where hydraulic parameters were adjusted. The adjusted parameters were the saturated soil water content and the field capacity, both of which were based on soil water content observations at several depths. For both parameters, a profile that decreased with increasing soil depth was selected. The saturated soil water capacity was set to $0.3 \text{ m}^3 \text{ m}^{-3}$ at the surface and $0.08 \text{ m}^3 \text{ m}^{-3}$ at depths of 1 m and more. The field capacity at the surface was $0.18 \text{ m}^3 \text{ m}^{-3}$ in both models and was decreased to a minimal value of $0.05 \text{ m}^3 \text{ m}^{-3}$ at soil layers deeper than 1 m. For

comparison, we also ran both models without changing any of the parameters. This is referred to as the default version of the model.

5) MODEL EVALUATION

Only a subset of the whole dataset was used for model evaluation. We selected data with energy budget closures (EBC) between 0.8 and 1.2, where EBC is defined as:

$$EBC = \frac{H + LE}{R_n + G}. \quad (3)$$

In equation 3, H and LE are the turbulent fluxes of sensible and latent heat, G is the ground heat flux and R_n is the net radiation. We applied this procedure to minimize possible measurement errors and assure comparability between the models, as their energy budget closures are equal to 1 by definition. In order to evaluate and visualize the performance of ISBA and SVS in simulating H and LE , we used Taylor diagrams (Taylor 2001), which can effectively illustrate several performance metrics on the same plot. The normalized standard deviation for a dataset from its mean is given by the radial distance from the origin at (0|0). The angle between the point of interest, and the abscissa defines the correlation of the model with the observations. The normalized RMSD of the model is shown by the distance to the point indicated as Observation at (1|0). The standard deviation and RMSD are both normalized using the standard deviation of the observations.

3. Results

a. General Conditions in Umiujaq

Table 2 shows the conditions during the three summers of interest at the Umiujaq site. These conditions varied considerably in terms of air and soil temperatures, with summer 2019 being the warmest and summer 2018 being the coldest. Also, precipitation differed considerably, although

there was no dry period in neither of the three summer, with a total of 237 mm in 2017 and 369 mm in 2019. Another marked difference, particularly in the beginning of summer, is the amount of fallen snow from the preceding winter. Summer 2018 was delayed by more than two weeks due to an exceptionally thick snow cover (≈ 1.4 m) from the previous winter. This was not the case for the other two summers, as the maximum snow depths from those winters were more typical (0.6-1.0 m).

Studying the one site over three summers allows us to investigate an interesting range of inter-annual variability, with total seasonal precipitation varying by up to 132 mm (factor of 1.56) and mean air temperatures varying by up to 2°C. In general, we can classify the summer of 2017 as dry, the summer of 2018 as cold, and the summer of 2019 as warm and humid.

b. Evolution of the Surface Energy Budget

Figure 5 illustrates the cumulative values of net radiation R_n , sensible heat flux H , latent heat flux LE , ground heat flux G , and residual flux for each of the three summers. Only data with an hourly EBC between 0.2 and 1.5 were used to omit obvious outliers. Our results indicate that H plays a more important role than LE in dissipating the net radiation. Figure 5 shows that over the three summers, H accounted for about 33% of the net radiation while LE represented roughly 22.4%. The ground heat flux played a far less important role, making up around 15% of the net radiation. The residual flux was 29.5% on average and thus comparable to H and LE . For this reason, a sensitivity test confirmed that the relative importance of H and LE does not vary with the applied EBC criterion. The average Bowen ratio BR (H/LE) fluctuated between 1.4 and 1.54 for and EBC ranges between 0.2 to 1.5 and 0.8 to 1.2. The energy budget closure was best in 2019, with a mean closure of 76.7%. Indeed, in 2019, there were more frequent winds from areas less affected by topography, and therefore more favorable to the use of the eddy covariance technique.

Although conditions varied considerably between the three summers, the general seasonal trends for H and LE were comparable (see Figure A2). H exhibited a strong increase directly after snowmelt and peaked in late June, with maximum values of 300 W m^{-2} . It then decreased throughout the summer before settling on a magnitude comparable to that of LE at the end of the summer. Unlike the clear peak and fall seen for H , LE did not follow a clear seasonal trend (Figure A2). It stayed rather constant throughout the season and shows only a small peak towards the middle of the summer in the last two years, reaching maximum values of nearly 200 W m^{-2} . Consequently, BR is greater than 1 at the beginning of summer and subsequently decreases, dropping below 1 at the end of the summer.

By comparing the three years with one another, it is evident that a delayed start to the summer season in 2018 did not have a large impact on the overall patterns of turbulent heat fluxes. H in 2019 showed a similar pattern to 2017, only differing in the second half of July due to an extended warm spell (Figure A1). In summary, it appears that H evolves in a consistent way and follows the same trend after snowmelt, with some differences as a result of meteorological conditions.

c. Evapotranspiration and Soil Properties

In this section, we look at one of the limiting factors for evapotranspiration: water availability. To further investigate whether a lack of water restricts the latent heat flux, Figure 6 displays the precipitation, evapotranspiration (ET), and soil water content in summer 2017 (summer 2018 and 2019 are available in the supplementary material).

Compared to other sites at similar latitudes, Umiujaq experiences high levels of precipitation (Figure 4), meaning there is no general lack of water supply at the surface. Rainfall in Umiujaq is also fairly equally distributed over the entire summer (Figure 2), therefore no distinct dry or wet periods were observed for any of the three summers (the corresponding figures for 2018 and 2019

are available in the supplementary material). However, as the red curve in Figure 6 indicates, the top soil volumetric water content under the lichen θ stayed rather low even during precipitation events, with peaks below $0.15 \text{ m}^3 \text{ m}^{-3}$ in 2017 and $0.2 \text{ m}^3 \text{ m}^{-3}$ for the three summers (the corresponding figures to Figure 6 are available in the supplementary material). Following rainfall, θ dropped rapidly to well below $0.1 \text{ m}^3 \text{ m}^{-3}$, and at times dropped to $0.05 \text{ m}^3 \text{ m}^{-3}$. This demonstrates that even though there is abundant water in the form of rain, the soil can only hold small amounts of the water. This is because the water drains rapidly due to the large particle size (Gagnon et al. 2019) of the soil and the associated high hydraulic conductivity.

Two post-rain periods in Figure 6, one at the end of June and the other in the beginning of July, very clearly demonstrate the influence of soil water content on ET. Approximately two days after the drop in surface water content, ET declined from about 2 mm day^{-1} to less than 0.5 mm day^{-1} and then rose again in response to the next rainfall event. The presence of a ≈ 2 -day lag time between both variables is particularly visible until mid-July as illustrated by the 3-day running mean of ET. The running mean of ET also highlights the influence of soil moisture on ET. Consistently, the opposite behaviour was observed for BR, which increased after the rainy periods.

d. Comparing Umiujaq to Other High Latitude Sites

Figure 7 illustrates the mean net radiation and sensible and latent heat fluxes for the comparison sites (see Table 1), exclusively during daytime ($R_n > 0$). The observed net radiation peaked at the beginning of the study period or shortly after, and subsequently decreased. Maximum net radiation values tend to drop at higher latitudes (CA-NS recorded more than 300 W m^{-2} at 56°N and the maximum mean value at RU-COK was 180 W m^{-2} at nearly 71°N).

The general behavior of H and LE can be roughly divided into three groups: LE that is typically higher than H (BR<1; RU-Cok, US-ICH, RU-VRK, US-EML), LE and H that are approximately

equal ($BR \approx 1$; US-AN and CA-NS) and finally LE that is lower than H ($BR > 1$; CA-UM (Umiujaq)).

We argue that this general behavior of H and LE is usually influenced by three main factors: (i) meteorological conditions, (ii) vegetation cover, and (iii) soil type. Factors (i) and (ii) are discussed in the remainder of this section, while factor (iii) is discussed in section 3e.

(i) There were considerable differences in the mean annual precipitation for all sites analyzed, ranging from 140 mm (US-AN) to more than 800 mm in Umiujaq. However, there is no correlation between the annual precipitation and the Bowen ratio or the magnitude of the latent heat flux. This can be seen at the drier sites in Alaska, which have BRs that are similar to or lower than the much more humid RU-VRK site. Moreover, the highest BR values were observed at Umiujaq, the most humid site. Thus, the annual mean precipitation does not limit LE , although in the aftermath of a rain event, LE levels will rise as seen for Umiujaq before.

(ii) Vegetation cover also alters the BR through transpiration. Through the shade they provide, shrubs cool ground temperature by reducing incoming radiation and consequently influence the relative importance of H and LE . The relative importance G was not analyzed as it was not available for all sites. Even though all the sites were classified as shrubland, the composition of the vegetation at these sites varied. Table 1 provides a brief overview of the vegetation at each site, from the sparse tussock tundra at US-AN with only a few shrubs, to the CA-NS site's dense shrub vegetation with small trees. However, in regards to precipitation, no correlation between the vegetation type and the Bowen ratio was found.

e. Soil Types

As discussed above, the influence of precipitation and vegetation on seasonal LE and its behaviour relative to H is limited. This leaves soil as the dominant factor that controls the BR at Umiujaq. Figure 8 shows the approximate position of each site in a soil texture diagram. The estimate for

Umiujaq is based on Gagnon et al. (2019), while the others are based on the description of the soil from the references presented in Table 1.

Although the positioning of each site within the soil texture triangle is highly uncertain, there is a clear distinction between Umiujaq and the other sites: all the other sites exhibit more clay and silt-rich soils compared to Umiujaq. Sand-rich soils are known to hold less water, drain quicker, and are thus often drier compared to soils rich in clay and silt. For instance, the mean soil moisture for the sites US-ICH ($0.39 - 0.52 \text{ m}^3 \text{ m}^{-3}$) and US-AN ($0.56 \text{ m}^3 \text{ m}^{-3}$) are considerably higher than for Umiujaq ($0.1 - 0.2 \text{ m}^3 \text{ m}^{-3}$). According to Clapp and Hornberger (1978), the saturated volumetric water content of sand is $\approx 0.1 \text{ m}^3 \text{ m}^{-3}$ less than for clay or silty clay. The largest difference is in the saturated hydraulic conductivity: $1.056 \text{ cm min}^{-1}$ for sand compared to $0.0062 \text{ cm min}^{-1}$ for silty clay (Clapp and Hornberger 1978).

Permafrost, which forms an impermeable layer, is also an important factor at each site (see Table 1). Sites that have a shallow active layer, defined as the soil layer that thaws in summer, have a topsoil layer that is often wet, thus favoring *LE*. At the majority of the comparison sites, the active layer is thin, ranging from 15 cm to 70 cm. At Umiujaq however, there is discontinuous/intermittent permafrost and an active layer that is quite deep (100 - 150 cm), which favors the percolation of surface water.

f. Model Comparison

1) SOIL MOISTURE

Results from the previous sections highlight a particularity of the Umiujaq site, that despite its abundant precipitation, the presence of sandy soil produced a low latent heat flux. In light of this, the ability of both models to simulate soil water content was first examined. As mentioned in section 2e, two runs were performed for each model. A default run was performed with no changes

474 to the parameters of the model and another with adjusted hydraulic parameters was performed
475 in order to improve soil simulations. Figure 9 compares the observed soil water content at two
476 different depths with the outputs from ISBA and SVS with default parameters and Figure 10 is the
477 corresponding figure for the model run with adjusted parameters.

478 For the default simulation, only the soil composition was input into the models. ISBA grossly
479 overestimated the soil water content at 6 cm by a margin of $0.2 \text{ m}^3\text{m}^{-3}$ during peak runoff and
480 produced results that were comparable to the observations about one month later in the simulation.
481 Even after the soil water content dropped to a range that was comparable to the observations, the
482 model still slightly overestimated the soil water content (around $0.02 \text{ m}^3\text{m}^{-3}$). The observed soil
483 water content at 6 cm and 50 cm differed considerably, while only minor differences between these
484 two layers were observed in the estimates produced by the two models. The surface water content
485 at 6 cm was overestimated more by SVS than by ISBA, but the default simulation revealed more
486 pronounced differences between values at 6 cm and 50 cm.

487 After adjustments were made (see section 2e), the ISBA outputs more closely matched the obser-
488 vations. When applying the modified soil properties, the simulated SVS results were slightly less
489 accurate than ISBA at the surface when compared to the observations. SVS generally underesti-
490 mated the soil water content by about $0.03 \text{ m}^3\text{m}^{-3}$, while it showed high peaks during rain periods.
491 At 50 cm in depth, SVS produced results that were comparable to ISBA and to the observed value.
492 To simulate the turbulent heat fluxes, the adjusted models were used.

493 2) TURBULENT FLUXES

494 Figures 11 and 12 show observations compared with simulated sensible and latent heat fluxes
495 from ISBA and SVS, respectively.

496 ISBA was able to accurately simulate H with the exception of a few high values that reached up
 497 to 400 W m^{-2} (Figure 11 b). The amount of scatter was generally quite low along the observed
 498 energy scale. Color-coding revealed that model performance was somewhat dependent on the time
 499 of the day for both models. Nighttime fluxes simulated by ISBA (light and dark blue dots) were
 500 systematically overestimated, but only by about 10 W m^{-2} . Also for ISBA, there was a slight
 501 tendency to overestimate H in the afternoon (14-18), while at the other times of the day, the points
 502 were scattered more evenly around the 1:1 line. Otherwise, the simulated fluxes closely followed
 503 the observed values. SVS simulations of H were more challenging. While the distribution of the
 504 points show a slightly more positive trend, SVS still adequately predicted H on average over the
 505 entire energy span, except for high values over 250 W m^{-2} , similar to ISBA. The mean bias for
 506 ISBA was 11.0 W m^{-2} while it was 2.9 W m^{-2} for SVS. The problem of overestimated nighttime
 507 fluxes that was observed with ISBA simulations did not occur with SVS. The modeled surface
 508 temperature compares favourably to the observations ($R^2 = 0.85$ and 0.9 , for ISBA and SVS), but
 509 has a positive bias in the case of ISBA (bias = 1.5°C) and a negative bias in the case of SVS (bias
 510 = -1.3°C).

511 Neither model was able to simulate LE as accurately as H , with the most notable problem for
 512 both models being the nighttime fluxes (Figure 12). ISBA rarely simulated very small negative
 513 fluxes, while SVS did not simulate them at all, despite the observational dataset presenting a regular
 514 occurrence of negative fluxes. A negative LE indicates the condensation of water vapor onto the
 515 soil surface, and this process seems completely absent from the models. Apart from this, both
 516 models yielded a data point distribution that exhibited a more positive trend compared to their H
 517 simulations. For ISBA, the bias was 5.3 W m^{-2} and for SVS, it was 11.1 W m^{-2} . Particularly,
 518 patterns with very low simulated fluxes and high observed fluxes were more frequent for LE than
 519 for H . The q-q plot for ISBA showed only minor deviations from the 1:1 line for fluxes with

520 an amplitude of less than 100 W m^{-2} . In the region above 100 W m^{-2} , the number of points
521 dramatically decreased, with only a few points responsible for the high deviations seen in the q-q
522 plot. For SVS, the datapoints diverge from the 1:1 line at a lower observed flux ($\approx 70 \text{ W m}^{-2}$) and
523 the slope following the divergence is higher than for ISBA. Thus, a significant portion of the *LE*
524 data points were overestimated, with deviations as large as 150 W m^{-2} for observed fluxes in the
525 100 W m^{-2} region.

526 The left side of Figure 13 shows that ISBA performed slightly better in simulating *H* than SVS.
527 The correlation (correlation coefficient of 0.91 for SVS versus 0.97 for ISBA) and the normalized
528 RMSD (0.44 for SVS versus 0.27 for ISBA) show better results for ISBA than for SVS. Also, the
529 standard deviation for ISBA was similar to that of the observations, and ISBA slightly overestimated
530 the variability in *H*, while the overestimation of SVS is a bit higher (1.08 for SVS versus 1.02 for
531 ISBA). These results are consistent with the more dispersed distribution of the data points in the
532 scatter plot that indicate the underestimation of *H* by SVS.

533 The Taylor diagram on the right of Figure 13 illustrates the relatively poor simulation performance
534 of the two models for *LE* compared to *H* which affected both models to a similar extent. (ISBA:
535 correlation coefficient of 0.84, standard deviation of 1.2 and normalized RMSD of 0.64; SVS:
536 correlation coefficient of 0.74, standard deviation of 1.12 and normalized RMSD of 1.64)

537 **4. Discussion**

538 *a. Energy Balance Closure and Measurement Error*

539 A typical issue with eddy covariance measurements is the lack of energy budget closure (Foken
540 2008). Throughout most meteorological conditions, the energy budget remained unclosed at the
541 Umiujaq site (on average 70.5%). However, certain wind directions can lead to better closure than

others (see Appendix Figure A4). The most probable reasons for the SEB imbalance at this site are: (I) the complex topography and (II) the mismatch between the footprint areas of the different flux sensors.

(I): As seen in Figure 1b, the flux tower was located to the east of a steep ridge. The cliff acted as a backward-facing step, potentially leading to flow detachment in the presence of westerly winds. The eddy covariance approach is based on the assumption that there is a surface located upstream that is flat and homogeneous, which is not the case here. However, in recent years, we have noticed an increase in studies conducted in non-ideal terrain (e.g. Geissbühler et al. (2000); Hiller et al. (2008); Etzold et al. (2010); Nadeau et al. (2013); Stiperski and Rotach (2015)). However, considering that the closure of the energy balance exceeded 55% for westerly wind direction (compare Figure A4), most of these data were retained. Foken (2008) argues that in heterogeneous terrains, large eddies can substantially contribute to the energy exchange when generated at the boundaries between different land uses that are excluded from the flux measurements.

(II): The footprint of the eddy covariance system naturally varies depending on wind speed and direction, covering an area of up to a few thousand square meters whereof 20 - 30% is dominated by lichen, and the rest by shrubs (Figure A3). Radiation measurements cover areas on the order of a few square meters, whereas the ground heat flux measurement is a point measurement. Thus, radiation and ground heat flux measurements that do not represent the flux footprint inevitably, which leads to an unbalanced budget. The measurements of R_n and G do not match those of H and LE taken in the footprint area, which may therefore lead to a budget imbalance. The radiation and ground heat flux measurements were derived from lichen-covered surfaces, but the footprint of the eddy covariance system covered a mixture of lichen and shrubs of varying heights, with either lichen or moss understory (Gagnon et al. 2019). This was however, difficult to quantify as knowledge on the distribution of each vegetation type would have been required for every 30-min

flux footprint area. However, we observed a worse energy budget closure (50% to 65%) for periods during which winds blew in from the south, where shrubs tended to be taller and the topography was more complex.

To explore the spatial variability of the surface albedo around the flux tower, point measurements of spectral albedo were made over the two dominant vegetation types (Belke-Brea et al. 2019), lichen and shrubs. The results showed slightly higher albedo levels over lichen, on the order of a few percentage points over the entire spectral range. Additionally, the ground heat flux was calculated under different vegetation types and only minor differences were observed, which alter the overall EBC by 1 or 2%. The 5TM probe did not cover the organic layer at the surface thus an additional error of the ground heat flux is probable. This is because the hydraulic and thermal properties of the organic layer differ from those of the mineral soil. Therefore, although the measurement footprint mismatch certainly contributes to the energy budget imbalance, it is not the sole cause.

Furthermore, measurement errors possibly contributed to the degradation of the energy budget. Mauder et al. (2013) reported errors of 10%-15% for the eddy covariance data for flags 0 and 1, which were used in this study. Radiation and ground heat flux measurements yielded errors of up to 10% (compare <https://www.campbellsci.com/cnr4>). Although the energy balance closure was rather poor, taking climatic conditions into account, it was still in the range of other carefully designed experiments (e.g. Cui and Chui (2019)). The energy budget closure could most probably be improved by additional radiation measurements and by relocating the tower further away from the steep ridge to reduce perturbations of the mean flow for westerly winds.

b. Importance of the Soil

As rather sparse vegetation prevails within the eddy covariance footprint (20 - 30% lichen and 60 - 70% small shrubs with low LAI) we conclude that it is not the dominant factor which controls the

SEB. Thus, the soil properties take a more central role in partitioning the turbulent heat fluxes. The coupling of soil moisture and evapotranspiration is emphasized by similar variation seen in Figure 6. The delay of evapotranspiration relative to soil moisture can likely be explained by variations in air temperature which also show an increased importance once the changes in soil moisture are rather small as observed in the second half of the period depicted in Figure 6.

Lemieux et al. (2020) investigated the groundwater dynamics of the Tasiapik valley and used empirical formulas to estimate that 40% of the total precipitation returned to the atmosphere as evapotranspiration. In our study, 22% to 35% of the precipitated water evaporated in the three summers of interest. There are likely two probable reasons for this difference. One was the use of empirical formulas in the Lemieux et al. (2020) study that were not specifically developed for the climatic conditions in Umiujaq. The second reason is the fact that the watershed contained areas with clay-rich soil, which can retain more water than sandy soil (Clapp and Hornberger 1978). Thus, the evapotranspiration rates for some parts of the Tasiapik valley are likely to be higher than those found around the flux tower used in our study.

ISBA performed slightly better in simulating soil conditions than SVS, revealing weaknesses of using a simplified soil scheme. However, both models exhibited problems when simulating the vertical profile of the soil water content and showed no or only a slight gradient of the volumetric water content in the default configurations (Figure 9). Thus, the models accounted for only a fraction of the actual changes in the soil properties that occurred with depth.

c. Turbulent Fluxes and Summer Energy Balance

Overall, the sensible heat flux is the dominant consumer of the energy from radiation due to the striking dependence of the latent heat flux on the soil water content. On average, the sensible heat flux is 61.3 W m^{-2} during summer, showing a decline from 81.5 W m^{-2} at the beginning of

summer to 48.3 W m^{-2} at the end. As previously discussed, the latent heat flux is more constant throughout summer with an average value of 39 W m^{-2} . The net input of energy from the radiation on average is 187.5 W m^{-2} .

Thus, in Umiujaq, the sensible heat flux makes up 33% of the net radiation on average while 22.4% can be attributed to latent heat flux. At the comparison sites, the proportion of sensible heat flux to radiation is 22% to 40% and the proportion of latent heat flux varies between 29% and 55%. Thus, sensible heat fluxes in Arctic environments seems somewhat less variable, while the differences in the latent heat flux are higher and depend more on local peculiarities such as water availability. This makes latent heat fluxes more difficult to simulate. The poor simulation of the soil at the Umiujaq site by the default configurations in both models, which leads to wetter conditions, impacted simulations of the turbulent fluxes as the latent heat flux was larger due to more water availability. After adjusting the soil properties, the ISBA simulations for these fluxes improved, indicating the influence of soil moisture on the turbulent fluxes in both models. While H was accurately simulated along the entire range of observed values, this was not the case for LE . The simulated values for LE were overestimated, closer to the higher-end values of the observed LE . This difference may be attributed to the fact that the parametrization of LE is less adapted to the dry soils at that site.

Both SVS and ISBA apply the same parametrization for the above-canopy turbulent fluxes. Therefore the reason for the more accurate H simulations in comparison to LE most probably lies in the soil simulation. Both the calculations for the thermal and the hydraulic regime are considerably simpler in SVS. As H and LE depend on the temperature and moisture content at the surface, respectively, an error in those variables directly affects the calculation of H and LE .

5. Conclusions

We analysed the SEB of a Subarctic experimental site in eastern Canada using the eddy covariance technique. We then compared the SEB to other Arctic and Subarctic sites and assessed two current LSMs, ISBA and SVS, in their ability to simulate SEB at the experimental site.

Our results showed that the sensible heat flux was the most important way of dissipating the energy from net radiation. The sensible heat flux accounted for about 33% of net radiation while the latent heat flux made up only 22.4%. The experimental site near Umiujaq consisted of sandy soil that resulted in very low soil water content at the surface, despite abundant precipitation for this latitude. The dry soil conditions most likely constrained the latent heat flux by limiting the water available for evaporation. This condition favored the dissipation of energy through sensible heat flux rather than through latent heat flux. Thus, we conclude that soil has a major influence on the partitioning of the turbulent heat fluxes in contrast to other studies that favor vegetation as the dominant factor for this partitioning.

Other documented sites in the Arctic showed net radiation partitioning that was different. At these sites, the latent heat flux was larger than the sensible heat flux. We attribute this difference to soil properties that retain more water, allowing it to drain slower.

We have also shown that both LSMs either need to be tuned or are not sophisticated enough to account for this particular situation concerning soil content. For this reason, we adjusted the models to be able to account for this factor. However, once the soil water content was better simulated through the modification of the models, the sensible heat flux in particular was very accurately simulated. On the other hand, latent heat flux simulations were not as accurate. Nonetheless, the results confirm that both LSMs are generally suitable for operational hydrometeorological applications in the humid Subarctic tundra ecotone.

Due to the isostatic uplift that followed deglaciation and led to the emergence of beaches, we stress that the sandy soil that covers a significant portion of Arctic Canada causes energy partitioning that is very different from what is presented in other available studies. A priori soil conditions found at the Tasiapik valley site seem uncommon. However, consulting this map (https://ftp.maps.canada.ca/pub/nrcan_rncan/publications/STPublications_PublicationsST/295/295462/gscgcm_195_b_2014_mn01p1.pdf), it can be seen that many regions in the Arctic have similar soil conditions. Indeed, large parts of the coasts of Hudson Bay and Ungava Bay, as well as other parts of the coast of numerous other regions in the eastern and central Canadian Arctic are covered by coarse sediments such as sand and gravel. Moreover, important parts of the Canadian Arctic comprise rocky outcrops that likely show even higher Bowen ratios. The soil water content and latent heat flux might therefore be overestimated by surface models across the Canadian Arctic. Furthermore, the disappearance of permafrost due to climate change may lead to drier soil and subsequently, altered latent heat fluxes.

Data availability statement. Data available on request from the authors.

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APPENDIX

Supplementary Figures

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Table 2.

Hydrometeorological conditions for the three summers of interest. Soil temper-
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Site	CA-UM	CA-NS	US-ICH	US-EML	US-AN	RU-VRK	RU-COK
Location	Quebec, Canada	Manitoba, Canada	Alaska, US	Alaska, US	Alaska, US	Vorkuta, Russia	Chokurdakh, Russia
Coordinates	56.33°N 76.28°E	55.92°N 98.96°E	68.61°N 149.30°E	63.88°N 149.25°E	68.93°N 150.27°E	67.05°N 62.94°W	70.83°N 147.49°W
Elevation (m ASL)	140	244	940	700	600	100	48
MAT (°C)	-4.0	-3.1	-7.4	-1.0	-11.0	-6.1	-14.3
MAP (mm)	800-1000	495	318	378	122	538	212
Vegetation	shrub tundra	heath tundra	tussok tundra	tussok tundra	shrub tundra	shrub tundra	shrub tundra
Soil Type	sand	clay	porous peat	till/ loess	n/a	loamy and peaty	fluvial silt
Permafrost	DC/S	DC/S	C	C	C	DC/S	C
Reference	This study	Goulden (2016)	Bret-Harte et al. (2016)	Belshe et al. (2012)	Rocha et al. (2016)	Biasi and Friborg (2016)	van der Molen et al. (2007)

TABLE 1: Description of the study sites. MAT and MAP are the mean annual temperature and the mean annual precipitation. C: Continuous, DC: Discontinuous, S: Sporadic.

	Period of interest	T_{mean} (°C)	T_{mean}^{Soil} (°C)	Precipitation (mm)	Maximum snow height (cm) in previous winter
2017	1 July - 5 September	9.6	11.5	237	65
2018		8.1	9.3	290	140
2019		10.1	11.7	369	100
Mean		9.3	10.8	298	101

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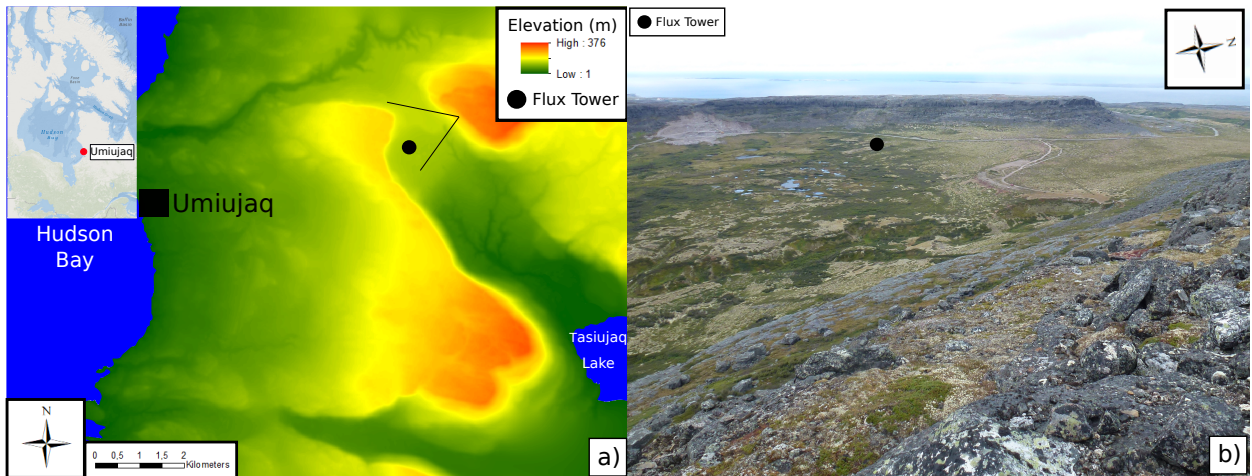


FIG. 1: a) Location of the study site and the experimental setup in the Tasiapik valley close to the village of Umiujaq on the eastern shore of Hudson Bay, Canada and b) photo of the upper portion of the valley where the flux tower is installed. The angle of view in panel b) is shown in panel a). Note the presence of a cuesta west of the tower.

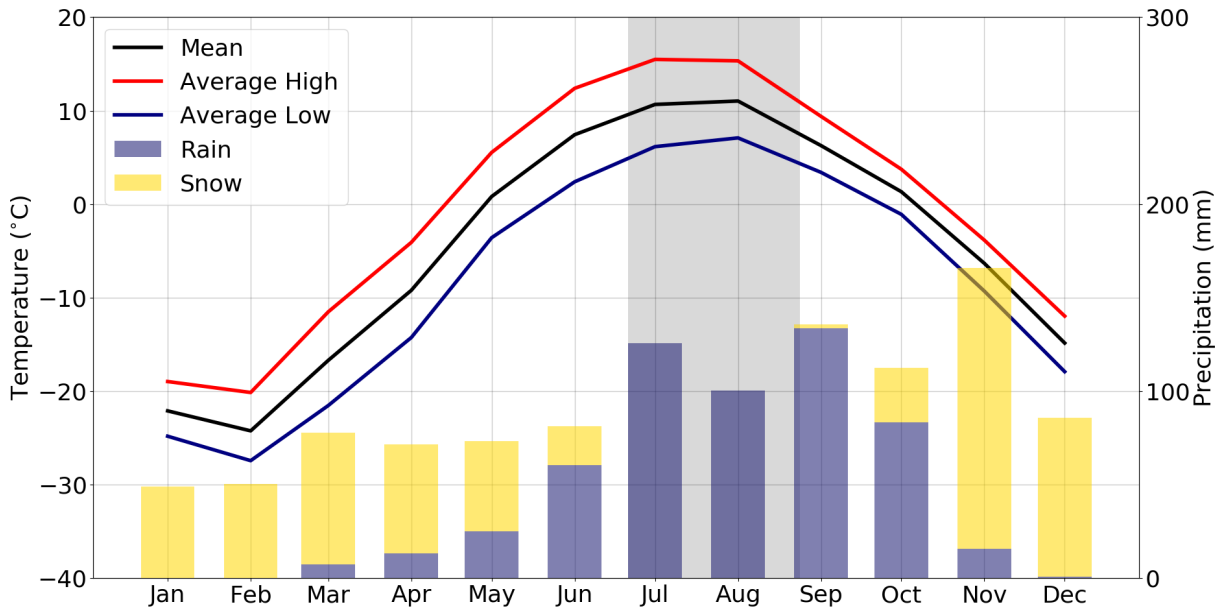


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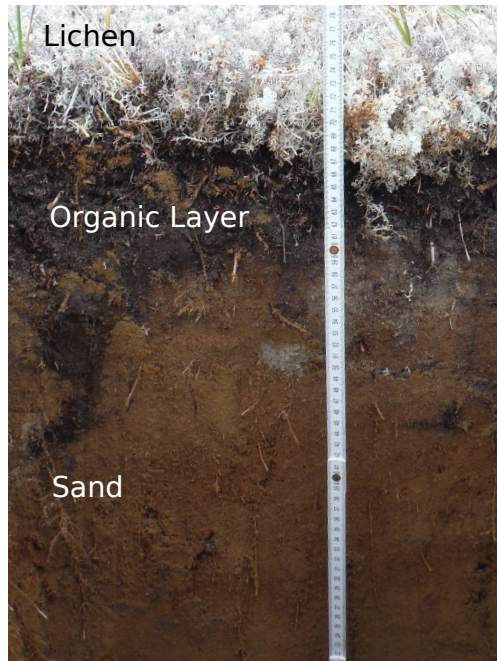


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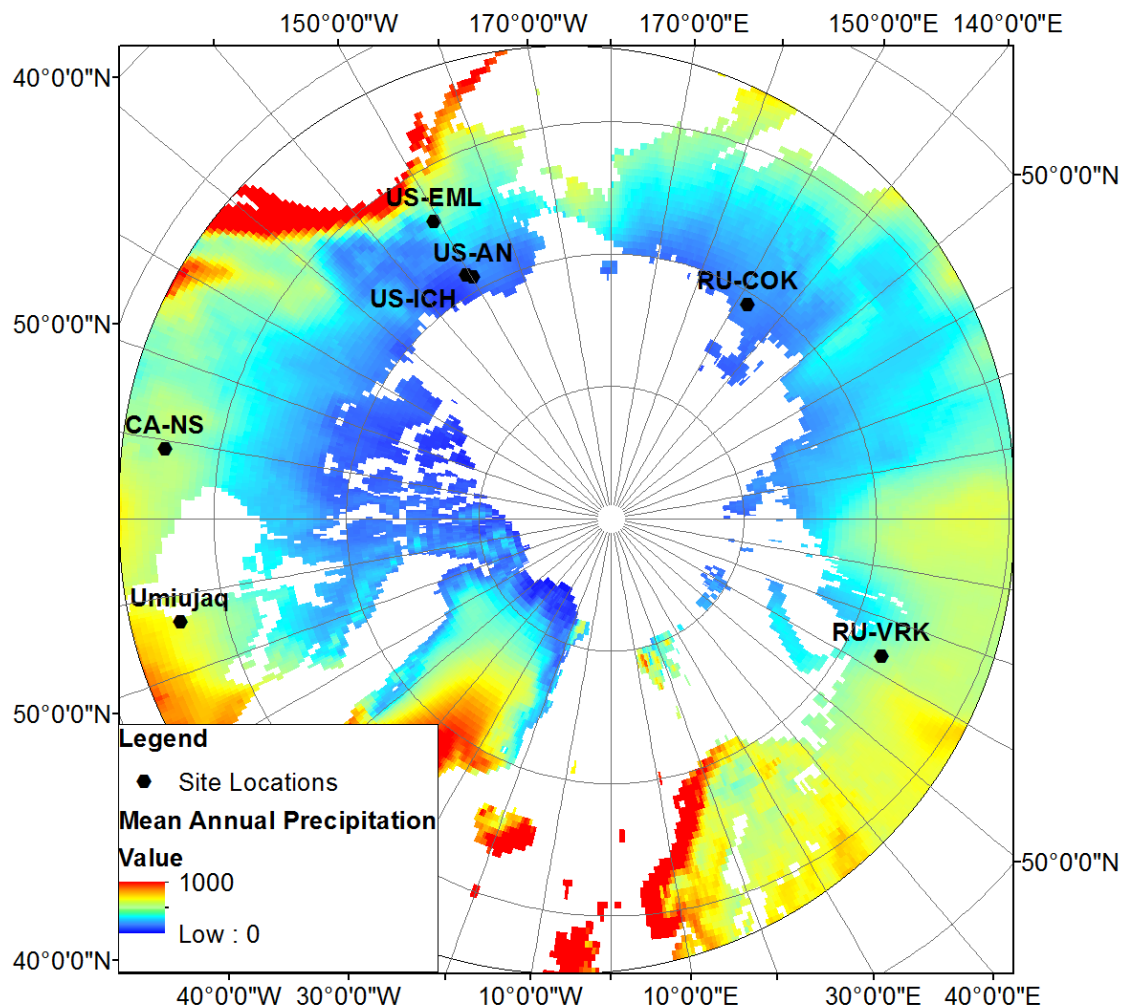


FIG. 4: FLUXNET site locations and spatial pattern of circumpolar yearly precipitation. Source: Harris et al. (2014)

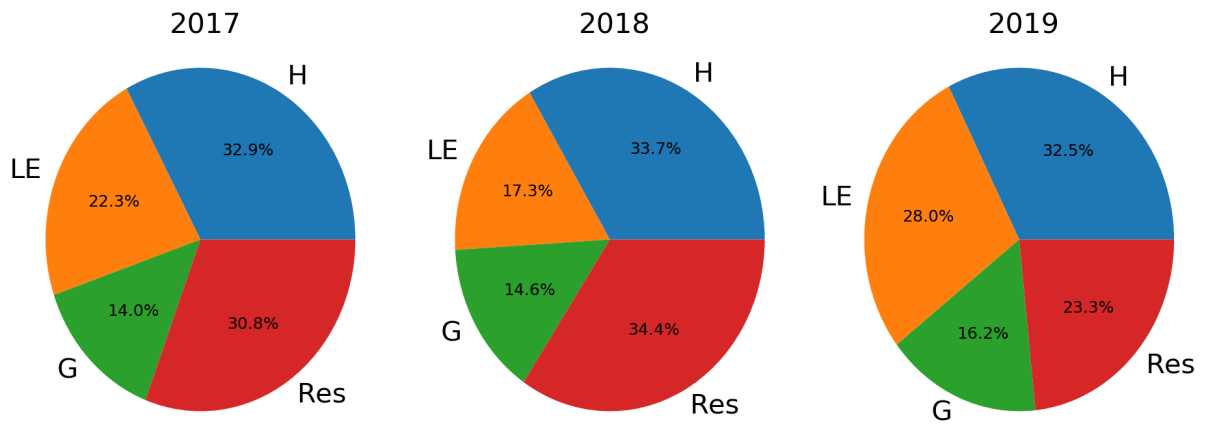


FIG. 5: Partitioning of net radiation into H , LE , and G for the three summers. The residual Res after the partitioning is also given. Only data with EBC between 0.2 and 1.5 were used.

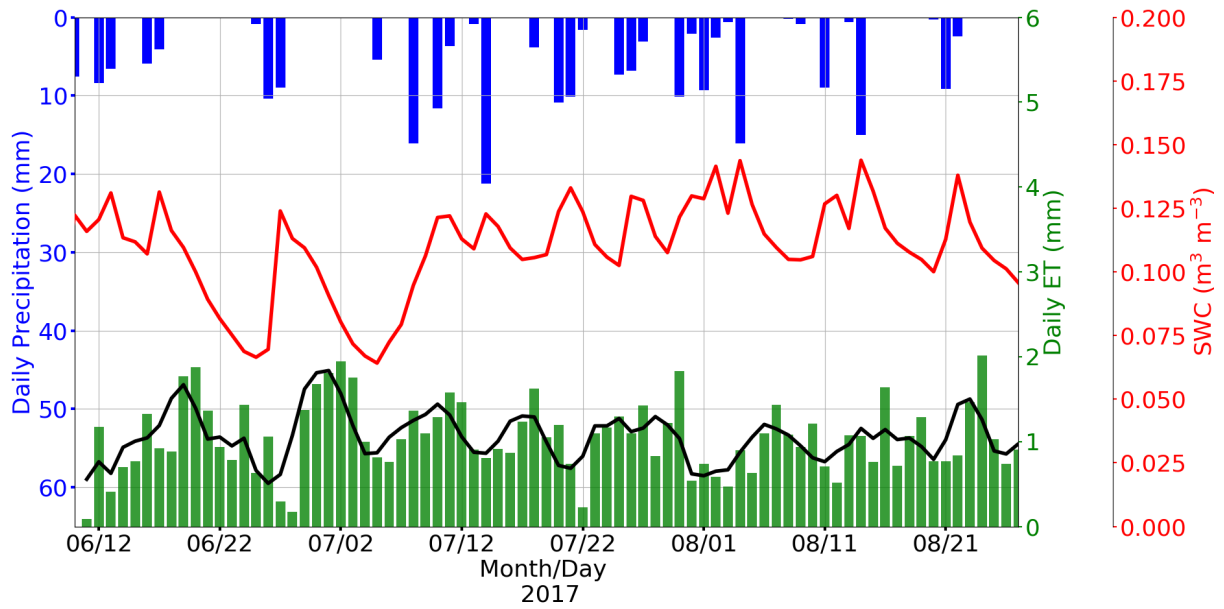


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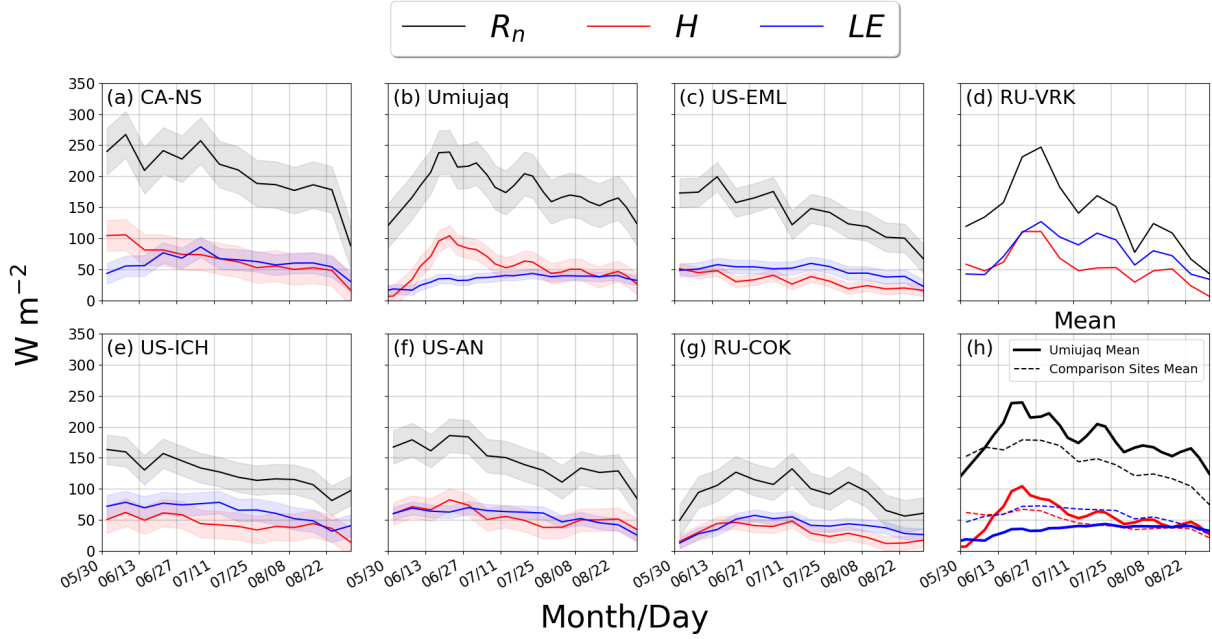


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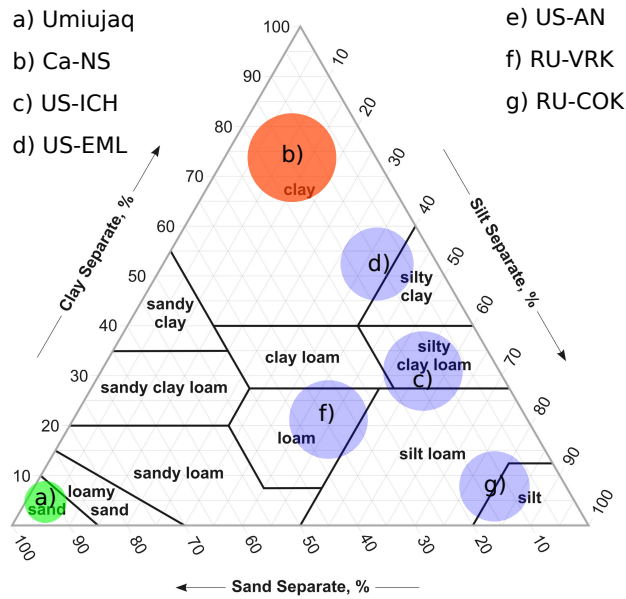


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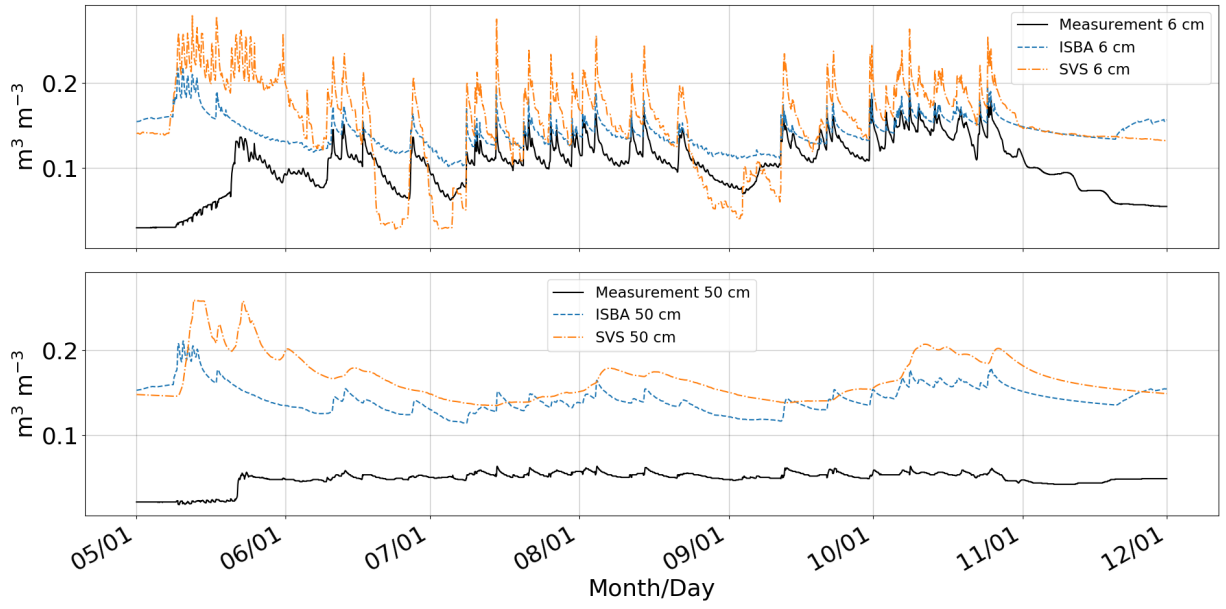


FIG. 9: Observed and modeled soil water content at 2 different depths for 2017 for models with default parameters.

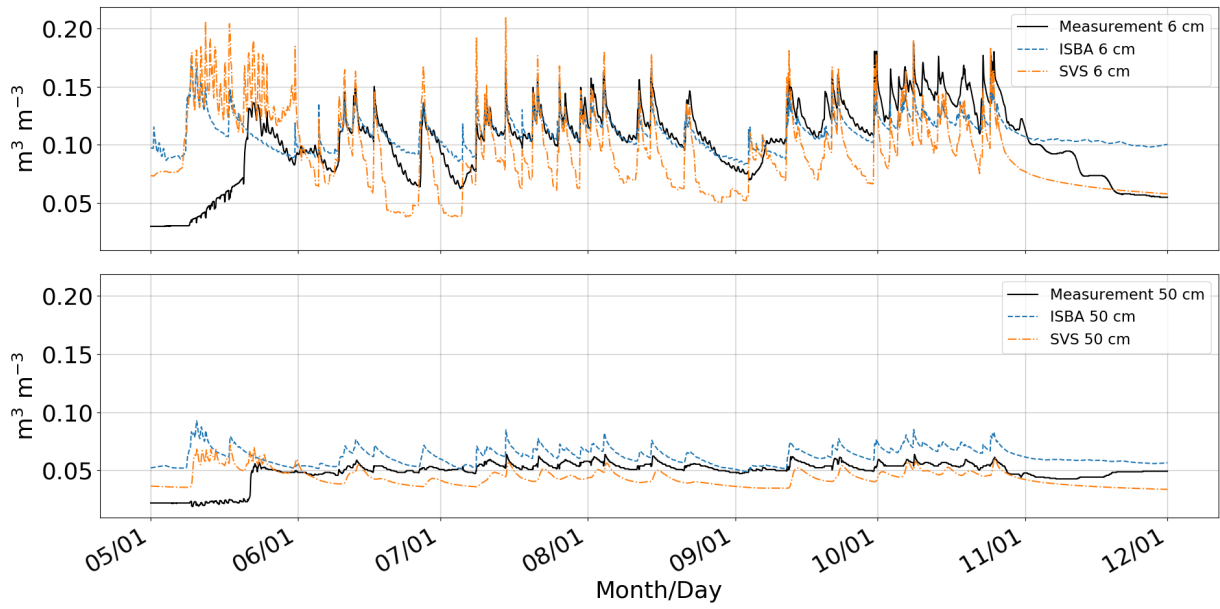


FIG. 10: Observed and modeled soil water content at 2 different depths for 2017 for models with adjusted parameters.

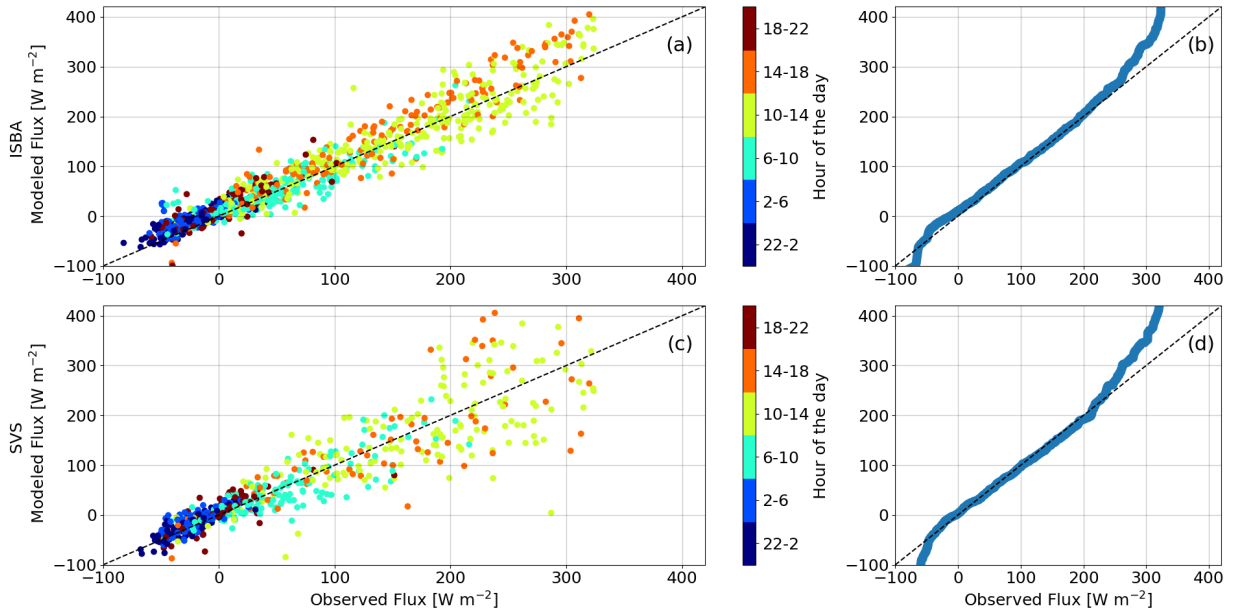


FIG. 11: Left column: Scatter plots comparing observed and modeled sensible heat fluxes at the Umiujaq site for all summers with the colors indicating the hour of day for (a) ISBA and (c) SVS. Right column: q-q plots, for which the quantiles of the distributions are plotted against each other, for (b) ISBA and (d) SVS.

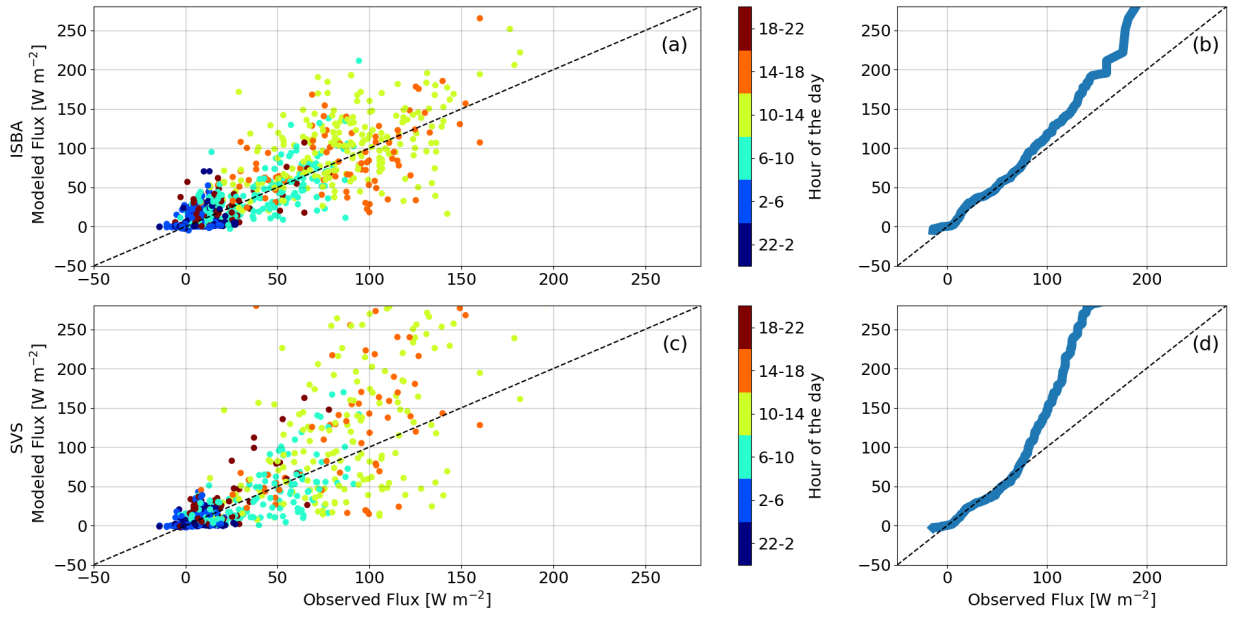


FIG. 12: Same as Figure 11 but for latent heat flux

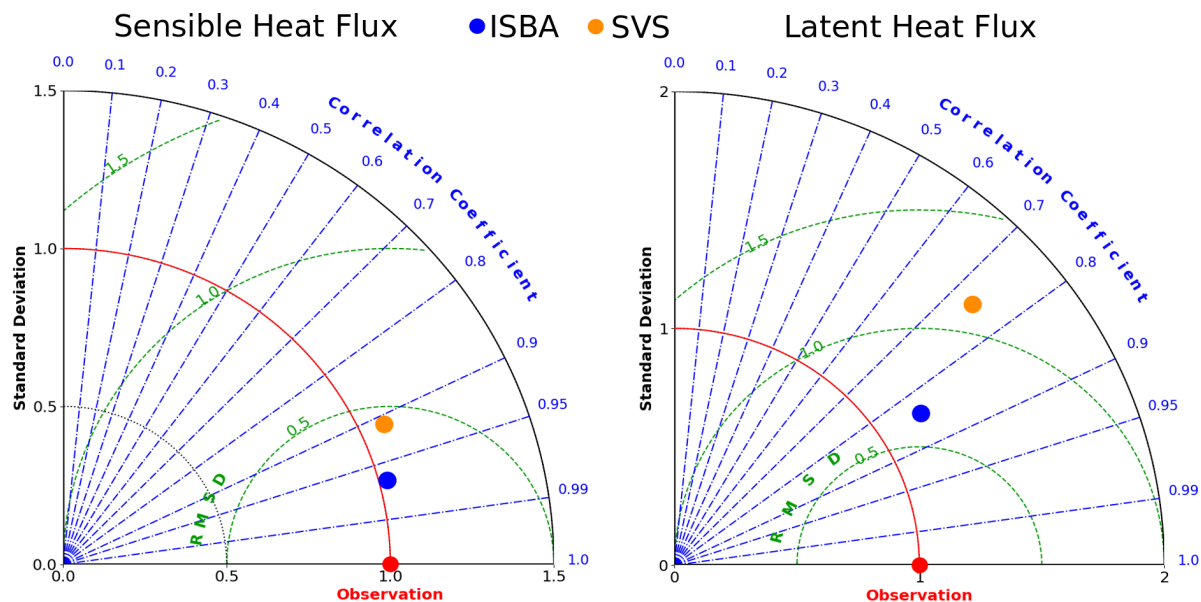


FIG. 13: Taylor diagrams indicating how closely modeled sensible heat flux (left) and latent heat flux (right) match observations.

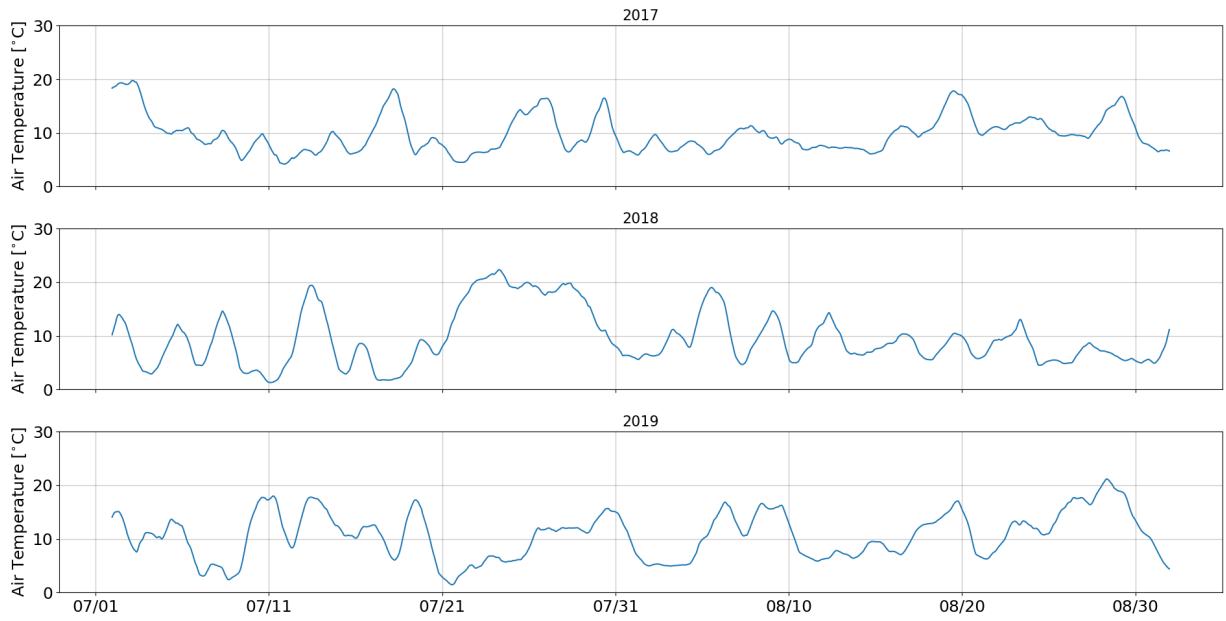


Fig. A1: Weekly running mean of the air temperature for summer 2017, 2018 and 2019.

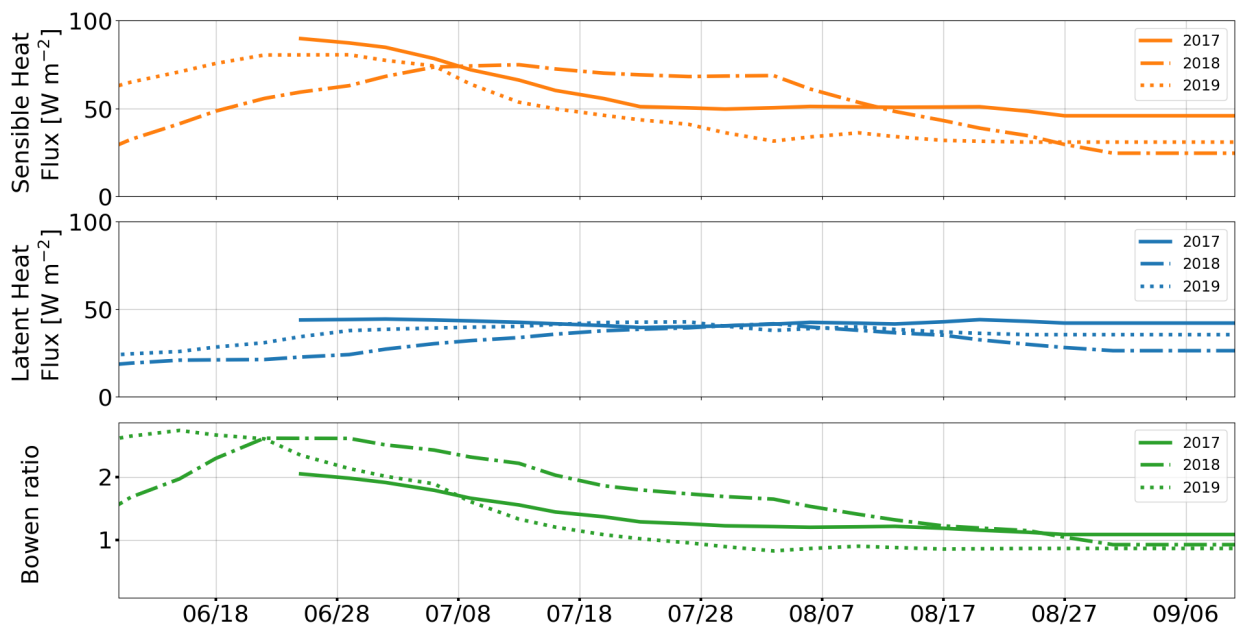


Fig. A2: Weekly running mean of summer sensible and latent heat fluxes for all three study years. Summer 2017 starts later as data were available only from mid June.

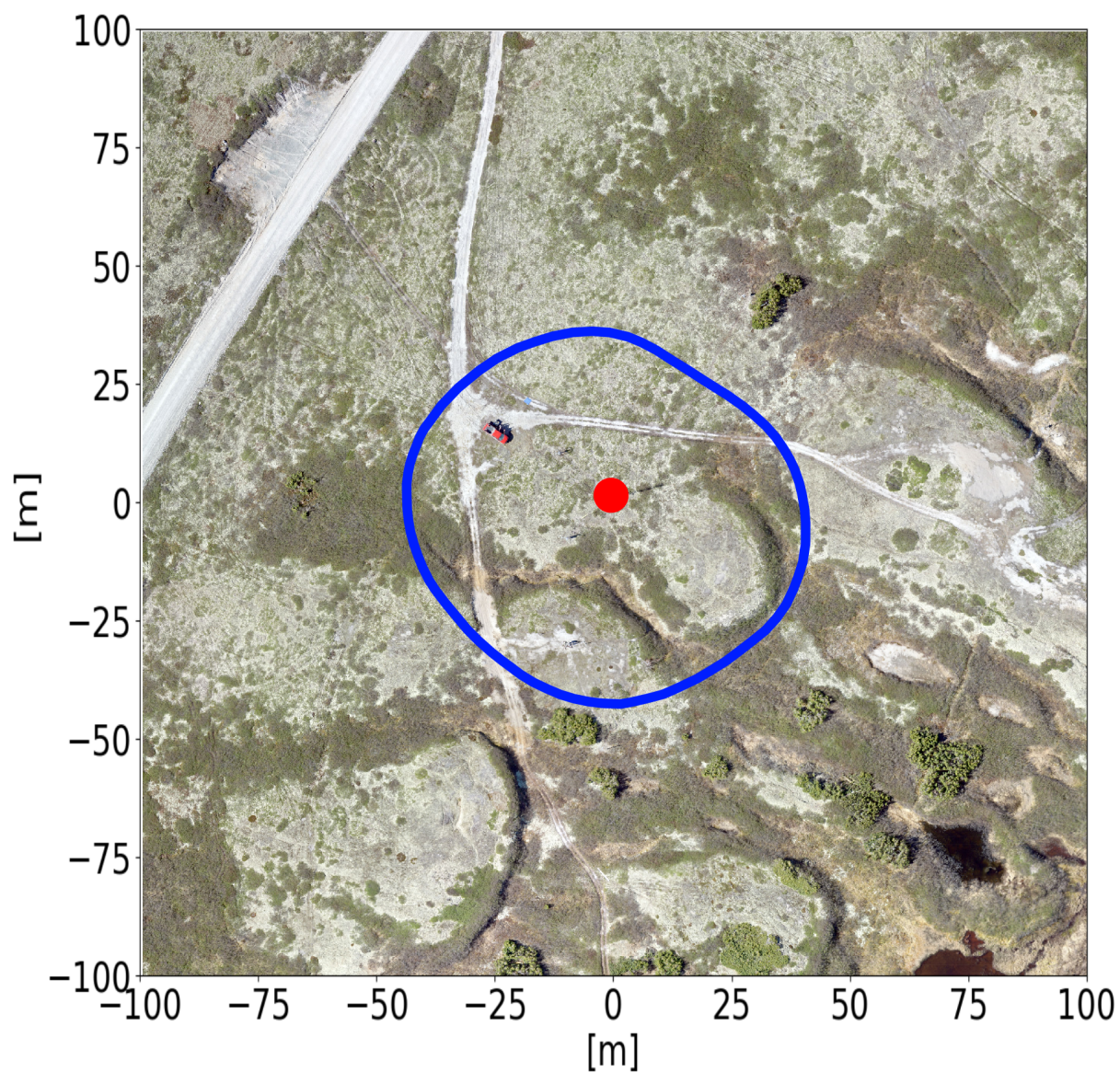


Fig. A3: Drone image of the study site with the flux footprint climatology of summer 2017, indicating 80% of the source area.

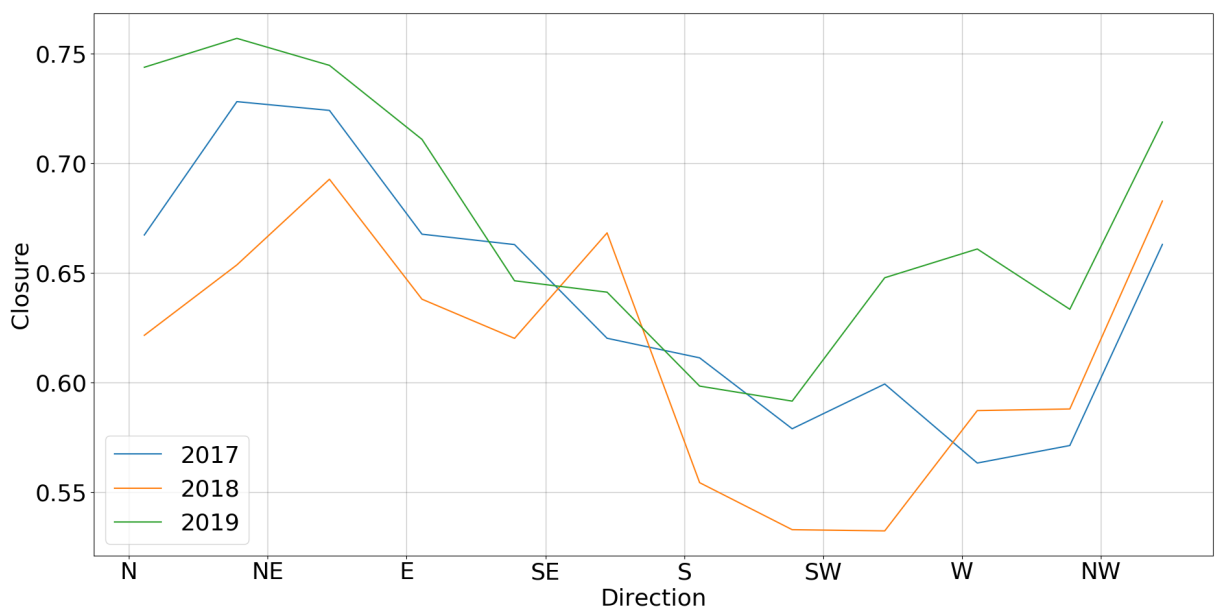


Fig. A4: Dependence of the energy budget closure on the wind direction for all three summers.