

1                    **Evaporative Resistance is of Equal Importance as**  
2                    **Surface Albedo in High Latitude Surface Temperatures**  
3                    **Due to Cloud Feedbacks**

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12                    **Key Points:**

- 13                    • Two Arctic plant types with different properties cause substantial changes to land  
14                    surface temperature through different physical pathways
- 15                    • Reducing land surface evaporative resistance increases low clouds and increases  
16                    shortwave cloud forcing
- 17                    • Albedo directly warms the land surface, while changes in evaporation warms mostly  
18                    by modifying cloud cover

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## Abstract

Arctic vegetation is known to influence Arctic surface temperatures through albedo. However, it is less clear how plant evaporative resistance and albedo independently influence surface climate at high latitudes. We use surface properties derived from two common Arctic tree types to simulate the climate response to a change in land surface albedo and evaporative resistance in factorial combinations. We find that lower evaporative resistances lead to an increase of low clouds. The reflection of light due to the difference in albedos between vegetation types is similar to the loss of incident sunlight due to increased cloud cover resulting from lower evaporative resistance from vegetation change. Our results demonstrate that realistic changes in evaporative resistance can have an equal impact on surface temperature to changes in albedo, and that cloud feedbacks play a first order role in determining the surface climate response to changes in Arctic land cover.

## Plain Language Summary

In the Arctic, darker land surfaces lead to warmer temperatures because they absorb more sunlight. However, there are multiple types of plants that grow in the Arctic, which differ not only in how dark they are but also how easily they release water. We investigate how different Arctic plants' absorption of sunlight and ability to release water to the atmosphere can affect temperature over Arctic land using an Earth System Model. We find that dark trees are capable of absorbing a greater fraction of the incoming sunlight than their brighter counterparts, which tends to warm the surface. In comparison, when the land surface has a harder time releasing water into the atmosphere, a smaller fraction of energy at the land surface is used to evaporate water. This warms the air above the surface, which leads to evaporation of cloud droplets and less cloud cover. As a result, more sunlight is able to reach the surface, and land surface temperatures are warmer even when the surface is relatively bright. In combination, we find that the darkness of the surface and the plants' ability to release water have an equal influence on surface temperatures over land in the Arctic.

## 1 Introduction

As the concentrations of atmospheric CO<sub>2</sub> and other greenhouse gases rise, global temperatures will continue to increase, with even larger increases at high latitudes (Collins et al., 2013). We expect that higher Arctic temperatures will lead to a poleward expansion of the tree-line (Gallimore et al., 2005; Lloyd, 2005; Jeong et al., 2011; Falloon et al., 2012). Tree-line expansion has already been observed in Alaska (Rupp et al., 2000; Lloyd, 2005) and in Sweden (Rundqvist et al., 2011). With continued warming, simulations of future climate scenarios suggest an increase in shrubs and needleleaf trees in the Arctic (Jeong et al., 2011; Falloon et al., 2012).

In addition to the current and future poleward expansion of trees at high northern latitudes, there is evidence from paleo records of expanded forest cover during past warm periods. Climate model simulations set with Mid-Holocene conditions show an expansion of boreal forest into the tundra relative to preindustrial-like conditions (Gallimore et al., 2005). There is also observational evidence for forests in the Arctic in the Late Cretaceous up to the Paleocene with deciduous trees occupying the land above 65°N (Wolfe & Upchurch, 1987; Royer et al., 2006). Along with the increased northward extent of these prehistoric forests, there are indications that these forests were deciduous, rather than the high-latitude needleleaf forests of the present day (G. Bonan, 2015).

Boreal forest communities are comprised of both needleleaf evergreen and broadleaf deciduous tree types (G. B. Bonan et al., 1992). Needleleaf trees have dark leaves (low albedo) and low transpiration rates (Chapin et al., 2000). They comprise the later suc-

67 cessional stages of the ecosystem and are characterized by slow growth rates (Van Cleve  
68 et al., 1996). By contrast, deciduous broadleaf trees in Boreal forests are relatively bright  
69 (higher albedo) with higher transpiration rates (Chapin et al., 2000). They grow back  
70 quickly following disturbance and can dominate the ecosystem following a stand-replacing  
71 disturbance event for a few hundred years (Van Cleve et al., 1996; Rupp et al., 2000),  
72 with consequences for surface energy partitioning (Liu et al., 2005). Although needle-  
73 leaf trees are conceptualized as, and often are, the dominant species in the ecosystem,  
74 pollen records suggest that boreal forests in Alaska were dominated by deciduous broadleaf  
75 vegetation following the last ice age (Edwards et al., 2005). Simulations suggest this could  
76 occur again in the next century as fire frequency and intensity increases and shifts ecosys-  
77 tems towards early successional plant types (Rupp et al., 2000).

78 Vegetation plays a large role in setting the terrestrial surface climate in the Arctic  
79 by altering the surface energy budget through the exchanges of mass and energy be-  
80 tween the land and the atmosphere. It is thought that the conversion of tundra to forests  
81 in the Arctic will contribute substantially to high-latitude warming trends due to a de-  
82 crease in albedo (Chapin et al., 2005), particularly as dark trees cover bright snow. Given  
83 that vegetation is expected to move poleward during warmer climates, we expect this  
84 migration to have a positive feedback on surface temperatures through the effect of albedo.

85 Observations and simulations of an expansion of Arctic trees suggest that warmer  
86 temperatures at high latitudes have been driven primarily by the impact of a darker sur-  
87 face (G. B. Bonan et al., 1992; Thomas & Rowntree, 1992; Foley et al., 1994; Chapin et  
88 al., 2005; G. B. Bonan, 2008; Falloon et al., 2012; Collins et al., 2013; Chae et al., 2015).  
89 However, vegetation in the Arctic can also influence the surface energy budget through  
90 the flux of water to the atmosphere. The magnitude of transpiration can vary substan-  
91 tially across vegetation types – in particular needleleaf evergreen trees have less evap-  
92 otranspiration than leafed-out deciduous broadleaf trees in the Arctic (Chapin et al., 2000).

93 In general, the effect of variations in plant type traits other than albedo, such as  
94 evaporative resistance (the difficulty for the plant to release water), on surface climate  
95 in the Arctic has been less well explored. Climate model simulations of an expansion of  
96 deciduous broadleaf trees (rather than needleleaf evergreen trees) at high latitudes found  
97 approximately equal amounts of warming from two distinct physical processes (Swann  
98 et al., 2010). The warming comes from both the change in albedo and the change in the  
99 greenhouse effect from elevated water vapor concentrations due to higher water flux to  
100 the atmosphere from the deciduous broadleaf trees, along with feedbacks from ocean and  
101 sea-ice (Swann et al., 2010). Other studies have focused on surface roughness and found  
102 that a change in vegetation from grasses and shrubs to needleleaf evergreen trees increases  
103 temperature due to the change in roughness which induces a cloud feedback (Cho et al.,  
104 2018), but they do not explore the changes in evaporative resistance, which we believe  
105 to be an important factor to influencing the Arctic climate (Swann et al., 2010; Laguë  
106 et al., 2019)

107 Swann et al. (2010) demonstrate that trees with different transpiration rates can  
108 have a significant influence on the Arctic climate, however it remains unclear what the  
109 independent relative contributions to Arctic climate are from albedo and transpiration.  
110 With the realistic possibility that tree-line will continue to move poleward as climate warms  
111 and that the forest composition may change to more deciduous broadleaf trees due to  
112 an increase of fire frequency and intensity (Rupp et al., 2000; G. B. Bonan, 2008) and  
113 as indicated by past climates, we need to understand how these changing properties from  
114 one plant species to another can affect the Arctic climate. In this paper we address the  
115 question of how albedo and resistance to surface evapotranspiration influence surface cli-  
116 mate in the Arctic.

## 2 Methods

### 2.1 Model Description

For this study we use climate model simulations with an idealized representation of the land surface to quantify the atmospheric and surface climate response to changes in albedo and evaporative resistance of the vegetation individually. We used the Simple Land Interface Model (SLIM; (Laguë et al., 2019)), which replaces the Community Land Model (CLM) within the Community Earth System Model (CESM; (Hurrell et al., 2013)). SLIM has a number of user controlled prescribed surface properties, including surface albedo and surface evaporative resistance akin to a bulk canopy resistance (see Supplementary Information for more detail) making it useful for independently modifying surface properties in order to analyze the effects of a change in a single surface property. This is in contrast to a complex land surface model such as CLM, where a change in vegetation results in simultaneous changes to many surface properties. For example, we can specify the snow-free surface albedo directly in SLIM, while in CLM the surface albedo is an emergent property of flat leaf albedo values, leaf area, etc.. SLIM conserves energy, and evaluates the surface energy budget at each time step to determine a new surface temperature, soil temperature profile, and net fluxes of shortwave radiation, long-wave radiation, sensible heat, and latent heat. Through CESM, SLIM is then coupled with the Community Atmosphere Model v. 5 (CAM5; (Neale et al., 2012)), a slab ocean model ((Neale et al., 2012)), and an interactive sea ice model (CICE; (Hunke et al., 2017)). We run our simulations globally at a resolution of  $1.9^\circ$  latitude by  $2.5^\circ$  longitude.

It is important to note that there is substantial disagreement between different land models for the robustness of biophysical climate responses to vegetation change (De Noblet-Ducoudré et al., 2012). While land surface models generally agree with each other, as well as with observations, on the effects of vegetation change on radiative fluxes, there is a much larger disagreement on how vegetation change should impact the partitioning of turbulent energy into sensible and latent heat fluxes (Duveiller et al., 2018; De Noblet-Ducoudré et al., 2012). In addition, atmospheric responses to vegetation change are substantial (Laguë et al., 2019), which means that models have a large uncertainty in the impact of vegetation change on near surface climate not only from differences in the land models and their flux representations, but also in the sensitivity of various atmospheric models to changes in land surface fluxes. Both factors (bias in turbulent fluxes and atmospheric responses) contribute to substantial differences in near-surface air temperature over land (Ma et al., 2018).

### 2.2 Simulations

In our simulations we set the land surface in the Arctic (north of  $60^\circ\text{N}$ ) to have uniform prescribed values for evaporative resistance and snow-free albedo corresponding to two plant types: evergreen needleleaf and deciduous broadleaf trees. We chose representative values for the albedo and evaporative resistance for each tree type by estimating them from grid cells dominated by our two plant types in a coupled land-atmosphere simulation using CLM. Needleleaf trees have a lower albedo and a higher evaporative resistance relative to broadleaf trees (Table S1). Albedo values are specified for four streams of radiation (visible direct light, visible diffuse light, near infrared direct light, near infrared diffuse light). Our idealized land model configuration allows us to independently change a single surface property, and therefore run simulations with a factorial combination of different values for albedo and evaporative resistance. For two of our four simulations, we have the plant type traits set to replicate needleleaf and broadleaf trees - that is, one simulation has needleleaf values for both albedo and evaporative resistance, while the other has broadleaf values for both properties (Fig. S1). The two additional simulations have ‘hybrid’ plant types with the albedo of one tree type paired with the evaporative resistance of the other, resulting in a brighter needleleaf tree and a darker

168 broadleaf tree. For simplicity we will refer to our four simulations as “Needleleaf” and  
 169 “Broadleaf”, for tree types with the observed combinations of albedo and resistance, and  
 170 “Bright Needleleaf”, and “Dark Broadleaf” for our hybrid tree types. The surface prop-  
 171 erties (albedo and evaporative resistance) are applied uniformly across all non-glaciated  
 172 land areas north of 60°N. That is, we effectively impose a mono-culture of each tree type  
 173 across the entire Arctic region in each simulation, regardless of the present-day vegeta-  
 174 tion type at each Arctic land location. Outside of the Arctic, surface properties reflect  
 175 those of the present-day vegetation growing in each location and are identical in all sim-  
 176 ulations.

177 We use summertime values derived from a CLM simulation where we take the June-  
 178 July-August surface properties in the Northern Hemisphere and the December-January-  
 179 February surface properties in the Southern Hemisphere. We choose summertime in or-  
 180 der to capture snow-free albedo values and growing season resistance values. The aero-  
 181 dynamic roughness of the land surface, which modulates the exchange of turbulent en-  
 182 ergy (sensible and latent heat) between the land and the atmosphere, is parametrized  
 183 as a function of vegetation height and is held fixed in time (varies spatially) in all sim-  
 184 ulations. SLIM contains a simple snow model which allows for the prescribed bare-ground  
 185 albedo to be masked by snow. Atmospheric CO<sub>2</sub> concentrations are set to a constant  
 186 value of 367 ppm.

187 We run our simulations for 50 years, using the last 30 years for analysis and omit-  
 188 ting the first 20 years to account for spin-up (see Supporting Information, Fig. S2).

### 189 2.3 Analysis

190 We focus our analysis on the summer months of June, July, and August, as these  
 191 months have the least amount of snow cover, thus allowing us to observe the impact of  
 192 the actual snow-free surface properties on the coupled climate system. Results are pre-  
 193 sented as area-weighted averages for all Arctic land surfaces (regions north of 60°N) un-  
 194 less otherwise noted. We report ranges of values of one standard deviation of variabil-  
 195 ity in time. Significance is calculated using a student’s t-test and indicated by stippling.  
 196 To account for lagged autocorrelation of up to two years, we assume  $N/2=15$  degrees of  
 197 freedom for the  $N=30$ -year period; we find this to be a conservative estimate of the ac-  
 198 tual degrees of freedom in the model using methods from Bretherton et al. (1999) (see  
 199 Supporting Information, Fig. S3, S4). A p-value of 0.05 or less indicates a significant dif-  
 200 ference with 95% confidence. Given that we have four experiments but no explicit ‘con-  
 201 trol’ run in the classic sense, we have in some cases compared three of the experiments  
 202 to a baseline of the Needleleaf tree type simulation (needleleaf albedo and needleleaf evap-  
 203 orative resistance), in order to see how the runs compared to one another.

204 To illustrate if changes in moisture or temperature are causing cloud responses we  
 205 use relative humidity as a proxy for cloudiness and analyze the differences in the verti-  
 206 cal profile of relative humidity between simulations. We partition the contribution into  
 207 two parts, one from differences in temperature (T), and another from differences in spe-  
 208 cific humidity (q). We report the change in the contribution of each term relative to the  
 209 normal Needleleaf run as follows:

$$\Delta RH_T = \frac{q_{ctrl}}{qsat_{exp}} - \frac{q_{ctrl}}{qsat_{ctrl}} \quad (1)$$

$$\Delta RH_q = \frac{q_{exp}}{qsat_{ctrl}} - \frac{q_{ctrl}}{qsat_{ctrl}} \quad (2)$$

210 where  $q_{ctrl}$  and  $qsat_{ctrl}$  are the specific humidity and saturated specific humidity  
 211 of the normal Needleleaf run and the  $q_{exp}$  and  $qsat_{exp}$  are the specific humidity and sat-  
 212 urated specific humidity of the other simulations that we are comparing to the normal

213 Needleleaf run. Equation 1 estimates the magnitude and sign of the change in the rel-  
 214 ative humidity profile between the simulations given the change in atmospheric temper-  
 215 ature alone and Equation 2 estimates the impact given the changing specific humidity  
 216 alone. The total change in relative humidity compared to the normal Needleleaf exper-  
 217 iment also includes a small contribution from the sensitivity of actual specific humidity  
 218 to temperature in the simulations. However, we are primarily focused on the dominant,  
 219 independent effects of temperature and specific humidity on the relative humidity pro-  
 220 files in response to changing surface properties described by Equations 1 and 2 (Fig. S5).

### 221 **3 Results & Discussion**

222 Based on prior literature, we expect that higher albedo surfaces (i.e the Broadleaf  
 223 and the Bright Needleleaf) will have cooler temperatures compared to lower albedo sur-  
 224 faces because they should absorb a smaller fraction of shortwave radiation. This assump-  
 225 tion held true for some, but not all of our simulations. The near surface air temperatures  
 226 are  $\sim 2^\circ\text{C}$  cooler for the Broadleaf simulation compared to the Needleleaf (Fig. 1a). How-  
 227 ever, both the Bright Needleleaf and Dark Broadleaf simulations were  $\sim 1^\circ\text{C}$  cooler than  
 228 the normal Needleleaf simulation despite having different surface albedos (Fig. 1a). This  
 229 suggests that additional processes are altering the surface energy budget beyond only  
 230 the change in surface albedo.

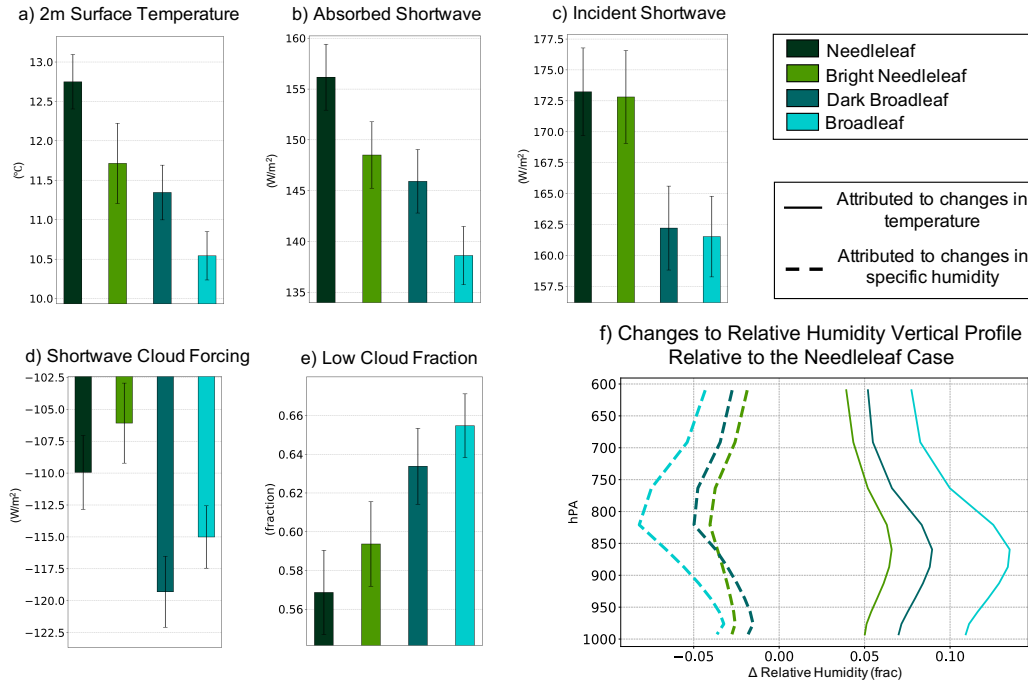
#### 231 **3.1 Surface Energy Budget**

232 The surface energy budget is comprised of five terms which must balance – absorbed  
 233 shortwave (SW) radiation, net long wave radiation, sensible heat flux, latent heat flux,  
 234 and heat storage in the ground. Averaged over the summer months, the total heat stor-  
 235 age in the ground is comparable across all simulations and will not be part of the anal-  
 236 ysis from here on. We find that absorbed SW radiation has a similar pattern to surface  
 237 temperature across experiments (Fig. 1a,b, 2a-c, g-i). Although the albedo directly af-  
 238 fects the fraction of the incident SW radiation that the surface absorbs, the simulations  
 239 with the same albedos (Needleleaf & Dark Broadleaf, Bright Needleleaf & Broadleaf)  
 240 differ from one another in absorbed SW radiation by  $\sim 10\text{Wm}^{-2}$  – a result that can-  
 241 not be explained by changes in albedo alone (Fig. 1b). Since this difference in absorbed  
 242 shortwave radiation is not due to any variation in surface albedo, it must instead be the  
 243 result of changes in the amount of solar radiation reaching the surface.

244 The incident shortwave radiation at the surface varies substantially between the  
 245 experiments with low evaporative resistance (Broadleaf and Dark Broadleaf) and the ex-  
 246 periments with high evaporative resistance (Needleleaf and Bright Needleleaf)(Fig. 1c,  
 247 S6d-f). The difference of incident shortwave radiation between high and low evapora-  
 248 tive resistance experiments suggests that surface evaporative resistance is altering down-  
 249 welling SW radiation.

250 The absorbed shortwave and incident SW results (Fig. 1b,c) indicate that clouds  
 251 are playing a first order role in feeding back on the surface energy budget in response  
 252 to changes in Arctic evaporative resistance. Incident SW radiation is very similar be-  
 253 tween simulations that have the same evaporative resistance but different surface albedo,  
 254 suggesting that evaporative resistance is the dominant factor modifying incoming SW  
 255 through cloud cover. Despite substantial differences in incoming SW radiation between  
 256 the Dark Broadleaf and Bright Needleleaf simulations, they have a similar amount of SW  
 257 radiation absorbed by the surface (Fig. 1b) and a similar change in surface temperature  
 258 (Fig. 1a). This is because the darker surfaces in the Dark Broadleaf simulation absorb  
 259 a larger fraction of the incident SW radiation than the Bright Needleleaf simulation (Fig.  
 260 1b,c), while the Bright Needleleaf simulation has a larger amount of incoming SW ra-  
 261 diation due to less low cloud cover resulting from the high evaporative resistance of the  
 262 land surface (Fig. 1b, 2g-i). The strong impact of changing low cloud cover on the sur-

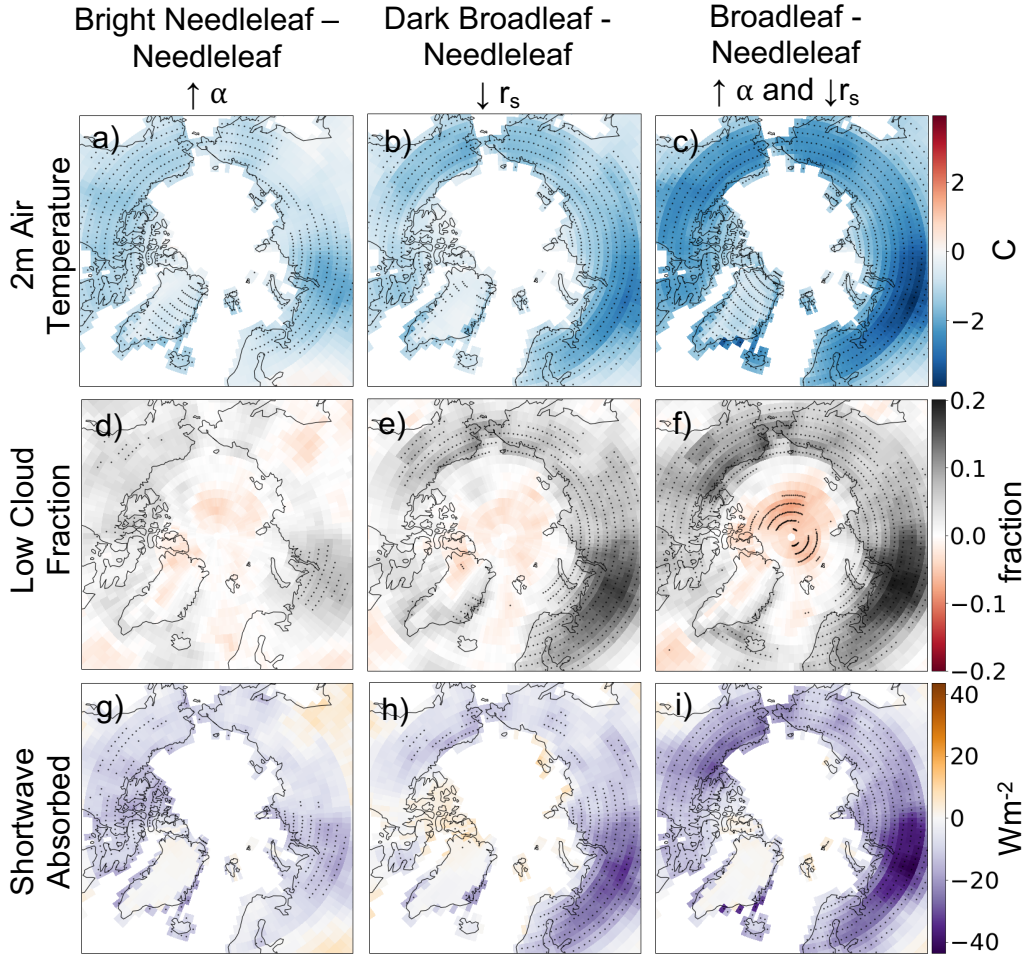




**Figure 1. Arctic average fluxes, states, and cloud changes.** Summertime averages over non-glaciated land North of  $60^{\circ}N$  for each of four different simulations (Needleleaf, Bright Needleleaf, Dark Broadleaf, Broadleaf) for a) 2m surface temperature (C), b) absorbed shortwave radiation at the surface ( $Wm^{-2}$ ), c) incident shortwave radiation at the surface ( $Wm^{-2}$ ), d) shortwave cloud forcing ( $Wm^{-2}$ ) with negative values indicating more clouds, and e) low cloud fraction. The error bars represent one standard deviation of variability in time. Panel f) shows changes in the relative humidity vertical profile relative to the Needleleaf case attributed to changes in the vertical profile of atmospheric temperature (solid lines) and specific humidity (dashed lines).

263 face energy budget in our simulations is consistent with observations from the Surface  
 264 Heat Budget of the Arctic (SHEBA) program and at the North Slope of Alaska Atmo-  
 265 spheric Radiation Measurement (ARM) site, where they show that low clouds provide  
 266 a net cooling in the summer through shading (Schweiger & Key, 1994; Shupe & Intrieri,  
 267 2004; Verlinde et al., 2016).

268 We expect simulations with lower evaporative resistances and higher amounts of  
 269 total absorbed radiation (absorbed SW plus downwelling longwave) to have higher la-  
 270 tent heat fluxes than simulations with high evaporative resistances and less total absorbed  
 271 radiation. In our simulations we see that the latent heat flux is largest ( $\sim 47Wm^{-2}$ )  
 272 for the Dark Broadleaf case, which has a high total absorbed radiation ( $\sim 475Wm^{-2}$ )  
 273 and a low resistance to evaporation (Fig. S7a). The Bright Needleleaf simulation has a  
 274 similar magnitude of total incoming radiation compared to the Dark Broadleaf simula-  
 275 tion, but the resistance to evaporation is larger for the Bright Needleleaf simulation, lead-  
 276 ing to less latent heat flux (Fig S8b). Despite a higher evaporative resistance, the Needle-  
 277 leaf case has the second largest latent heat fluxes ( $\sim 46Wm^{-2}$ ), instead of the Broadleaf  
 278 case, as a result of the largest total incoming radiation  $\sim 485Wm^{-2}$  (Fig. S8a,b). The  
 279 relative amount of sensible heat flux across simulations, which is driven by the gradient  
 280 in temperature from the surface to the atmosphere, shows a similar pattern as surface



**Figure 2.** Spatial patterns of change over the Arctic compared to Needleleaf simulation during the summertime. First row (a-c) shows the difference in 2m air temperature (C), the second row (d-f) shows the change in low cloud fraction, the third row (g-i) shows the change in absorbed shortwave radiation ( $Wm^{-2}$ ) at the surface. Surface temperature and shortwave absorbed are plotted only over land. Column 1 shows the response to increasing albedo ( $\alpha$ ) alone (Bright Needleleaf - Needleleaf), Column 2 shows the response to decreasing evaporative resistance ( $r_s$ ) alone (Dark Broadleaf - Needleleaf), and column 3 shows the response to simultaneously increasing albedo and decreasing evaporative resistance (Broadleaf - Needleleaf). Stippling indicates significance.

281 temperature with larger surface temperatures being associated with larger sensible heat  
 282 flux, but with a greater distinction between the Bright Needleleaf and the Dark Broadleaf  
 283 cases (Fig. S8c).

284 The Needleleaf experiment has the second largest latent heat flux despite having  
 285 a high resistance to evaporation. While at first this seems surprising, it can be readily  
 286 explained by the fraction of the turbulent fluxes occurring as latent heat flux (LH),  
 287 defined as the Evaporative Fraction (EF):

$$EF = \frac{LH}{LH + SH} \quad (3)$$



where SH denotes Sensible Heat fluxes. We find that the low evaporative resistance simulations (Broadleaf and Dark Broadleaf) have relatively higher evaporative fractions of  $\sim 0.64$  compared to the high evaporative resistance simulations (Needleleaf and Bright Needleleaf) which have evaporative fractions of  $\sim 0.58$  (Fig. S8d). The differences in evaporative fraction indicate that simulations with lower evaporative resistance dissipate more energy through latent heat leading to stronger cooling compared with high evaporative resistance simulations regardless of albedo.

Downwelling longwave fluxes emitted by the atmosphere toward the land surface are strongly influenced by surface temperatures (Vargas Zeppetello et al., 2019) and near-surface humidity, and have been observed to be influenced by Arctic cloud cover (Shupe & Intrieri, 2004; Verlinde et al., 2016). In our simulations with warmer surface temperatures, we also find more humid air, and more cloud cover. Thus we expect to see larger downwelling long wave radiation. We find that the greatest downwelling longwave radiation is found in the warmest simulation (Needleleaf with  $\sim 330 W m^{-2}$ ), however the second largest downwelling longwave flux comes from the third warmest experiment (Dark Broadleaf with  $\sim 328 W m^{-2}$ ). Based on surface temperatures alone, we would expect the Bright Needleleaf to have a larger downwelling longwave radiation than the Dark Broadleaf; however, we find that the Dark Broadleaf has more water vapor in the lower parts of the atmosphere and a larger low cloud fraction than the Bright Needleleaf (Fig. 1e, S9c, S10). Thus we hypothesize that the specific humidity and the increase of low clouds may boost the downwelling longwave radiation in the Dark Broadleaf simulation.

### 3.2 Clouds

The experiments with low evaporative resistance (Broadleaf and Dark Broadleaf) have a greater fraction of low clouds than experiments with high resistance (Needleleaf and Bright Needleleaf), and experiments with low albedos (Needleleaf and Dark Broadleaf) have a smaller fraction of low clouds than experiments with high albedos (Broadleaf and Bright Needleleaf) (Fig. 1e). Given the differences we observe in low cloud fraction across these experiments, we infer that changes in both albedo and evaporative resistance influence cloud formation, although the effect from the change in albedo is not as large as from the change in evaporative resistance. Cloud formation depends on the profile of relative humidity, which in turn depends both on the atmospheric temperatures and specific humidity. Both of these factors may respond to altered surface albedo and evaporative resistance. In particular, albedo and evaporative resistance may influence both the temperature of the atmosphere and the total amount of water vapor, both of which are important for cloud formation. To identify which of these factors is responsible for the change in cloud fraction that we observe in our simulations, we look at the vertical structure of relative humidity, a variable that directly describes how close the air is to saturation. We estimated the contribution to changes in the vertical structure of relative humidity from the changes in the profile of temperature and specific humidity in each of our simulations (using Equations 1&2).

Variations in relative humidity profiles between our experiments are dominated by changes in temperature (Fig. 1f). We find that most of the increase of relative humidity can be attributed to cooling of the vertical temperature profile, driven either by changes in surface albedo or evaporative resistance (Fig. 1f). Compared to the Needleleaf simulation, all other experiments show a decrease in specific humidity, which would also act to reduce the relative humidity (Fig. S10), however, this effect is secondary. We thus find that the increase in relative humidity associated with increasing low cloud cover in our lower evaporative resistance cases (Dark Broadleaf and Broadleaf) is largely driven by cooler temperature profiles in simulations with higher evaporative fraction rather than by direct changes in specific humidity. With cooler temperatures and increased low cloud fraction (Fig. 2a-f), we also find a decrease in the 500 hPa geopotential heights (Fig. S11a-c).

340 Cloud feedbacks occur in response to changes in both albedo and evaporative re-  
 341 sistance, resulting in changes to shortwave cloud forcing. Shortwave cloud forcing is de-  
 342 fined as

$$SW_{cloud\ forcing} = netSW_{all\ sky} - netSW_{clear\ sky} \quad (4)$$

343 where  $netSW_{all\ sky}$  is the shortwave radiation at the top of the atmosphere when the ra-  
 344 diative effect of clouds is included (all sky) and  $netSW_{clear\ sky}$  is the same but using a  
 345 solution from the radiative calculations in the atmospheric model as if there were no clouds  
 346 present (clear sky). We find that the simulations with low evaporative resistance (Broadleaf  
 347 & Dark Broadleaf) have a greater magnitude of shortwave cloud forcing than the high  
 348 evaporative resistance simulations (Needleleaf & Bright Needleleaf), on average by  $\sim 9Wm^{-2}$   
 349 (Fig. 1d, S6g-i). When evaporative resistance is held fixed and albedo is changed, there  
 350 is still a change in shortwave cloud forcing of  $\sim 4Wm^{-2}$ .

351 Both evaporative resistance and albedo modify shortwave cloud forcing, but through  
 352 different processes. Changing evaporative resistance modifies shortwave cloud forcing pri-  
 353 marily through the total amount of low cloud cover (Fig. 1c,d), while changing albedo  
 354 modifies shortwave cloud forcing through clear sky fluxes of shortwave radiation (Fig.  
 355 S7b). Thus, in the case of albedo, even a relatively small change in cloud cover can re-  
 356 sult in a substantial change in shortwave cloud forcing because adding clouds above a  
 357 dark surface has a greater impact on the amount of SW absorbed by the land surface  
 358 than it does over a bright surface.

359 Earlier work by Cho et al. (2018) identified that low cloud feedbacks were an im-  
 360 portant factor in determining the surface temperature response to a change in vegeta-  
 361 tion cover in the Arctic. However it is unclear from their study how clouds would respond  
 362 to a change in evaporative resistance alone. Consistent with Cho et al. (2018), we also  
 363 see decreases in low cloud cover and increases in the magnitude of shortwave cloud forc-  
 364 ing in response to a darker surface. They propose two possible explanations for the re-  
 365 duction in low clouds: first, a reduction in relative humidity caused by an increase in tem-  
 366 perature, and second, an increase of roughness causing an increase in the planetary bound-  
 367 ary layer height. In our simulations we see large differences in cloud cover and shortwave  
 368 cloud forcing due to changes in evaporative resistance which we also attribute to changes  
 369 in the vertical profile of temperature; however, in contrast we find a reduction in low cloud  
 370 cover and an increase of surface temperature change to be driven by the evaporative frac-  
 371 tion rather than the surface roughness. We additionally note that our experimental de-  
 372 sign using SLIM allows us to directly separate the effects of surface properties such as  
 373 albedo and evaporative resistance on Arctic climate. Future simulations could potentially  
 374 be used to isolate the impact of surface aerodynamic roughness, but this is not explored  
 375 in this study.

### 376 3.3 Further Considerations

377 In this paper we have identified that changes in evaporative resistance associated  
 378 with a shift in vegetation cover over the Arctic influences the evaporative fraction, re-  
 379 sulting in cloud feedbacks which have the same order of magnitude effect on energy fluxes  
 380 and surface temperature as changes in surface albedo. This explanation holds in the sum-  
 381 mer, but we find that it does not appear in winter due to the accumulated snow cover  
 382 and the lack of incoming sunlight, which changes the turbulent fluxes for our simulations.  
 383 However, the surface temperature pattern that we see in the summertime in our four ex-  
 384 periments appears to persist throughout the year with a smaller magnitude. We hypoth-  
 385 esize that the differences in the temperature over land in the wintertime are a result of  
 386 differences in the simulated amount of seasonal sea ice (Fig. S12a). The differences in  
 387 sea ice in each of the seasons are broadly correlated with the amount of sea ice loss in

388 the summer (Fig. S12b), which we hypothesize to be driven by differences in summer-  
389 time temperatures. Thus the differences year-round could be indirectly driven by sum-  
390 mertime conditions.

391 We note that actual vegetation has seasonal variations in albedo and evaporative  
392 resistance. In this idealized study we have chosen to represent only the summer values  
393 of surface properties and are unable to parse the effect of seasonal variations in leaf area  
394 by masking snow during shoulder seasons (Cook et al., 2008; Swann et al., 2010; Bon-  
395 fils et al., 2012; Luyssaert et al., 2018).

396 Uncertainty in the CLM parameter values used to inform our imposed change in  
397 albedo could modify our results. For example, Majasalmi and Bright (2019) find that  
398 while CLM has a reasonable representation of visible albedo for boreal plant types, the  
399 albedo in the near-infrared is underestimated. A brighter albedo for needleleaf boreal  
400 trees in the near-infrared, similar to the near-infrared albedo of broadleaf boreal trees,  
401 would reduce the total temperature effect of the change in albedo associated with a change  
402 in boreal forest type, although the effect is smaller than it would be for a bias in visi-  
403 ble albedos.

#### 404 **4 Conclusions and Implications**

405 We analyzed the effects of specified albedos and evaporative resistances associated  
406 with two common tree types in the Arctic: needleleaf evergreen trees and broadleaf de-  
407 ciduous trees. We find that evaporative resistance plays a large role in influencing sur-  
408 face air temperature over land in the Arctic, similar in magnitude to the influence of sur-  
409 face albedo. In simulations with lower evaporative resistance we see that there is an in-  
410 crease in the low cloud fraction, which in turn reduces the shortwave radiation incident  
411 at the land surface and enhances shortwave cloud forcing. The difference in incident ra-  
412 diation due to changes in evaporative resistance is then compounded by changes in albedo  
413 in cases where both evaporative resistance and albedo are modified, resulting in dras-  
414 tically different temperatures between experiments with differences in both albedo and  
415 evaporative resistance (Broadleaf and Needleleaf), and similar temperatures when either  
416 one of the surface properties is swapped (Bright Needleleaf and Dark Broadleaf). Our  
417 results show that evaporative resistance is as important in influencing Arctic surface tem-  
418 peratures as surface albedo and needs to be considered in future studies when trying to  
419 understand the effects of vegetation change in the Arctic. These results also demonstrate  
420 the usefulness of idealized approaches to land surface modeling (in our case with SLIM)  
421 and how we can use this modeling approach to isolate individual surface properties to  
422 quantify how changes in specific aspects of the land surface influence the larger climate  
423 system. Further studies focused on the role of specific land surface properties and their  
424 influence on Arctic climate and circulation could advance our understanding of the po-  
425 tential future climate impacts of high-latitude vegetation change.

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