

**Snow thickness  
retrieval over thick  
Arctic sea ice using  
SMOS satellite data**

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# Snow thickness retrieval over thick Arctic sea ice using SMOS satellite data

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## Abstract

The microwave interferometric radiometer of the European Space Agency's Soil Moisture and Ocean Salinity (SMOS) mission measures at a frequency of 1.4 GHz in the L-band. In contrast to other microwave satellites, low frequency measurements in L-band have a large penetration depth in sea ice and thus contain information on the ice thickness. Previous ice thickness retrievals have neglected a snow layer on top of the ice. Here, we implement a snow layer in our emission model and investigate how snow influences L-band brightness temperatures and whether it is possible to retrieve snow thickness over thick Arctic sea ice from SMOS data.

We find that the brightness temperatures above snow-covered sea ice are higher than above bare sea ice and that horizontal polarisation is more affected by the snow layer than vertical polarisation. In accordance with our theoretical investigations, the root mean square deviation between simulated and observed horizontally polarised brightness temperatures decreases from 20.0 K to 4.4 K, when we include the snow layer in the simulations. Under cold Arctic conditions we find brightness temperatures to increase with increasing snow thickness. Because dry snow is almost transparent in L-band, this brightness temperature's dependence on snow thickness originates from the thermal insulation of snow and its dependence on the snow layer thickness. This temperature effect allows us to retrieve snow thickness over thick sea ice. For the best simulation scenario and snow thicknesses up to 35 cm, the average snow thickness retrieved from horizontally polarised SMOS brightness temperatures agrees within 0.7 cm with the average snow thickness measured during the IceBridge flight campaign in the Arctic in spring 2012. The corresponding root mean square deviation is 6.3 cm, and the correlation coefficient is  $r^2 = 0.55$ .

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# 1 Introduction

The Soil Moisture and Ocean Salinity (SMOS) mission carries the first satellite-based passive microwave radiometer that measures radiation emitted from the Earth at a frequency of 1.4 GHz in the L-band. The mission was launched successfully in 2009 and since spring 2010 observations have been made available to scientific and operational users (Mecklenburg et al., 2012). Although designed to provide global estimates of soil moisture and ocean salinity, L-band brightness temperatures measured by SMOS can be used to retrieve thin sea ice thickness (Kaleschke et al., 2010, 2012). In the previous retrieval methods, the sea ice was assumed to be snow-free. Here, we investigate how a snow layer on top of the ice influences the brightness temperature above sea ice, and whether there is a relationship between the brightness temperature and the thickness of the snow layer.

Due to its high albedo, snow on sea ice is important for the surface energy balance. Additionally, snow affects the surface radiative properties of sea ice and thus modifies the remote sensing signal. Furthermore, information on snow thickness is required for the freeboard-based estimation of sea ice thickness from lidar and radar altimetry (Kwok and Cunningham, 2008; Giles et al., 2007). However, snow thickness observations over sea ice are scarce. The most comprehensive analysis for the Arctic Ocean is based on snow thickness and density measurements from Soviet drifting stations between 1954 and 1991 (Warren et al., 1999). Though, it is not clear how well this climatology represents present-day snow conditions (Kurtz and Farrell, 2011). First cross-basin surveys of snow thickness over Arctic sea ice have been provided by airborne radar measurements (Kwok et al., 2011). However, airborne remote sensing is spatially and temporally restricted to individual campaigns. A snow thickness retrieval method from passive microwave satellite measurements uses the spectral gradient ratio of the 19 and 37 GHz vertical polarisation channels (e.g. Markus and Cavalieri, 1998). However, surface roughness variations introduce uncertainties to this method (Stroeve et al., 2006), and the method is only applicable to dry snow conditions and

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only to Antarctic sea ice and first-year ice in the Arctic, but fails over multi-year ice (Comiso et al., 2003). Thus, a method to estimate snow thickness over thick Arctic multi-year ice from SMOS brightness temperatures would improve monitoring of sea ice conditions in the Arctic from space.

The maximum ice thickness that can be retrieved from L-band radiometry depends on the dielectric properties of sea ice, which can be described by ice temperature and salinity (Kaleschke et al., 2010). The maximum retrievable ice thickness determines what we consider here as thin ice. For example, for sea ice with a salinity of  $S_{\text{ice}} = 8 \text{ g kg}^{-1}$  and a bulk temperature of  $T_{\text{ice}} = -5^\circ\text{C}$ , the maximum retrievable ice thickness in L-band is about 50 cm and increases to higher values for colder and less saline conditions (Kaleschke et al., 2010). The one ice layer radiation model used for sea ice thickness retrieval from SMOS in previous studies (Kaleschke et al., 2010, 2012) neglects the potential presence of a snow cover on sea ice. Snow is almost transparent for microwave radiation at 1.4 GHz frequency (e.g. Hallikainen, 1989; Rott and Mätzler, 1987; Hall, 1996). However, the reflectivities at the ice-snow and the snow-air boundaries differ from the reflectivity at the ice-air boundary. Thus, a snow layer on the ice has an impact on the effective emissivity and accordingly on the brightness temperature of sea ice. Additionally, snow has a thermal insulation effect on ice and the bulk ice temperature of snow-covered sea ice is generally higher than the bulk ice temperature of bare sea ice. Because the ice temperature determines the dielectric properties of sea ice, snow thus also has an indirect effect on the brightness temperature of sea ice.

Here, we use a multiple layer model based on the radiation model presented in Burke et al. (1979) to examine the impact of a snow cover on brightness temperatures above sea ice and the implications for the ice thickness retrieval of snow-covered sea ice. In order to test the validity of our theoretical investigations, we simulate brightness temperatures for ice and snow thicknesses measured during the National Aeronautics and Space Administration (NASA) Operation IceBridge flight campaign in spring 2012 in the Arctic. We perform simulations that neglect and that include the snow layer. We then

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compare these simulated brightness temperatures with brightness temperatures measured by SMOS. Furthermore, we investigate whether there is a dependence between the brightness temperatures observed above snow-covered ice and the thickness of the snow cover on the ice. We evaluate whether SMOS has not only the potential for retrieving ice thickness over thin ice, but also the potential for estimating snow thickness over thick sea ice in the Arctic.

## 2 Data and methods

The analysis presented is based on three components: (1) an improved ice emission model to perform sensitivity studies and to simulate brightness temperatures for observed ice conditions, (2) SMOS brightness temperature measurements, and (3) measurements from the IceBridge campaign that provide model input and validation data. These three components are described in the following.

### 2.1 Emission model

In previous studies (Kaleschke et al., 2010, 2012), the ice brightness temperature was determined with a radiation model based on the approach described in Menashi et al. (1993). The approach is valid for a dielectric slab of ice that is bordered by the underlying water and the air above the slab of ice. However, here we need a radiation model that describes radiation in an ice layer that is covered by a snow layer. Thus, we choose the radiation model described in Burke et al. (1979). This model is based on the radiative transfer equation and was originally developed for soil moisture applications of X- and L-band radiometer measurements. The model describes the radiation emitted from a stratified bare soil with  $N$  layers. The dielectric properties are assumed to be constant across the layers. For our sea ice applications, we consider a semi-infinite layer of air on top, a layer of snow on top of a layer of ice, and a semi-infinite layer of sea water at the bottom. The emission model describes the brightness temperature above

snow-covered sea ice as a function of the air permittivity, of the water temperature and permittivity, and of the temperatures, permittivities and thicknesses of the snow and the ice layer.

The permittivity of air is assumed to be  $\epsilon_{\text{air}} = 1$ , which is the value for vacuum. For the permittivity of sea water, we here use the empirical relationship by Klein and Swift (1977) for a water salinity of  $S_{\text{water}} = 33 \text{ g kg}^{-1}$ . For the permittivity of sea ice, we use an empirical relationship that describes the ice permittivity  $\epsilon_{\text{ice}}$  as a function of brine volume fraction  $V_b$  within the ice (Vant et al., 1978):

$$\epsilon_{\text{ice}} = a_1 + a_2 V_b + i(a_3 + a_4 V_b), \quad (1)$$

where  $V_b$  is given in ‰, and  $a_1$ ,  $a_2$ ,  $a_3$ , and  $a_4$  are frequency-dependent coefficients. Here, we use the coefficients for a frequency of 1.4 GHz for multi-year ice as given in Kaleschke et al. (2010). The empirical relationship is valid for  $V_b < 70$  ‰. The brine volume fraction  $V_b$  can be expressed as a function of the bulk values for the ice salinity  $S_{\text{ice}}$ , the ice density  $\rho_{\text{ice}}$ , the density of the brine  $\rho_{\text{brine}}$ , and the ice temperature  $T_{\text{ice}}$  (Cox and Weeks, 1983). For the ice density  $\rho_{\text{ice}}$ , we use an expression that relates  $\rho_{\text{ice}}$  to ice temperature  $T_{\text{ice}}$  (Pounder, 1965):

$$\rho_{\text{ice}} = 0.917 - 1.403 \times 10^{-4} T_{\text{ice}}, \quad (2)$$

where  $T_{\text{ice}}$  is given in °C. For the brine density  $\rho_{\text{brine}}$ , we use an expression that depends on brine salinity  $S_{\text{brine}}$  (Cox and Weeks, 1983):

$$\rho_{\text{brine}} = 1 + 0.0008 S_{\text{brine}}, \quad (3)$$

where  $S_{\text{brine}}$  is inserted in ‰. We obtain the brine salinity  $S_{\text{brine}}$  from polynomial approximations for the dependency between brine salinity and ice temperature (Vant et al., 1978). For the snow permittivity, we use a polynomial fit obtained for snow permittivity measurements at microwave frequencies ranging between 840 MHz and 12.6 GHz (Tiuri et al., 1984). Based on these measurements, it is suggested that the permittivity

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of snow mainly depends on snow density and snow wetness and that the permittivity is practically independent of the structure of snow.

The water temperature is assumed to be at the freezing point of Arctic sea water ( $T_w = -1.8^\circ\text{C}$ ). In order to determine the bulk temperatures of the snow layer and the snow-covered ice layer, we here assume a balance of heat fluxes at the snow-ice-interface and account for the different thermal conductivities of ice and snow. We assume that the temperature at the bottom of the ice is at the freezing point of water (i.e.  $T_{\text{bottom}} = T_w = -1.8^\circ\text{C}$ ), and that the temperature gradients within ice and snow are linear. We assume that at the snow-ice-interface the ice temperature equals the snow temperature:  $T_{\text{ice}}(z = d_{\text{ice}}) = T_{\text{snow}}(z = d_{\text{ice}}) = T_{\text{si}}$  with  $T_{\text{si}}$  being the snow-ice-interface temperature, and  $d_{\text{ice}}$  being the ice thickness. Here,  $z$  denotes the vertical distance from the ice-water-interface. Thus,  $z$  is  $z = 0$  at the ice-water-interface,  $z = d_{\text{ice}}$  at the snow-ice-interface, and  $z = d_{\text{ice}} + d_{\text{snow}}$  at the snow surface. We assume that thermal conduction is continuous through the snow-ice-interface (Maykut and Untersteiner, 1971):

$$k_{\text{ice}}\gamma_{\text{ice}}(z = d_{\text{ice}}) = k_{\text{snow}}\gamma_{\text{snow}}(z = d_{\text{ice}}), \quad (4)$$

where

$$\gamma_{\text{ice}}(z^*) = \left. \frac{\partial T_{\text{ice}}(z)}{\partial z} \right|_{z=z^*} \quad (5)$$

$$\gamma_{\text{snow}}(z^*) = \left. \frac{\partial T_{\text{snow}}(z)}{\partial z} \right|_{z=z^*}, \quad (6)$$

and  $k_{\text{ice}}$  and  $k_{\text{snow}}$  are the thermal conductivities of ice and snow, respectively. Because we assume linear temperature gradients within the ice and the snow layer,  $\gamma_{\text{ice}}(z) = \gamma_{\text{ice}}$  and  $\gamma_{\text{snow}}(z) = \gamma_{\text{snow}}$  are constant values. The surface temperature  $T_{\text{surf}}$  and the snow-ice-interface temperature  $T_{\text{si}}$  are then described by

$$T_{\text{surf}} = T_{\text{si}} + \gamma_{\text{snow}}d_{\text{snow}} \quad (7)$$

$$T_{\text{si}} = T_w + \gamma_{\text{ice}}d_{\text{ice}}. \quad (8)$$

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If we know the surface temperature  $T_{\text{surf}}$ , we can solve this system of three Eqs. (4), (7), and (8) and three unknowns  $\gamma_{\text{ice}}$ ,  $\gamma_{\text{snow}}$ , and  $T_{\text{si}}$ . For the thermal conductivity of snow we use a constant climatological value of  $k_{\text{snow}} = 0.31 \text{ W m}^{-1} \text{ K}^{-1}$  (Yu and Rothrock, 1996), and for the thermal conductivity of ice we use a parameterisation accounting for ice temperature and salinity (Untersteiner, 1964):

$$k_{\text{ice}} = 2.034 \text{ W K}^{-1} \text{ m}^{-1} + 0.13 \text{ W kg}^{-1} \text{ m}^{-2} \frac{S_{\text{ice}}}{T_{\text{ice}} - 273}. \quad (9)$$

To simplify the calculations we use the mean temperature of the snow and ice column  $T_{\text{mean}} = 0.5(T_{\text{surf}} + T_{\text{w}})$  instead of the ice temperature  $T_{\text{ice}}$  in Eq. (9).

The bulk ice and snow temperatures  $T_{\text{ice}}$  and  $T_{\text{snow}}$  are then

$$T_{\text{ice}} = \frac{1}{2}(T_{\text{w}} + T_{\text{si}}) \quad (10)$$

$$= T_{\text{w}} + \frac{1}{2}K(T_{\text{surf}} - T_{\text{w}})k_{\text{snow}}d_{\text{ice}} \quad (11)$$

$$T_{\text{snow}} = \frac{1}{2}(T_{\text{si}} + T_{\text{surf}}) \quad (12)$$

$$= \frac{1}{2}(T_{\text{w}} + T_{\text{surf}} + K(T_{\text{surf}} - T_{\text{w}})k_{\text{snow}}d_{\text{ice}}) \quad (13)$$

$$\text{where } K = (k_{\text{ice}}d_{\text{snow}} + k_{\text{snow}}d_{\text{ice}})^{-1}.$$

In summary, the input parameters for the model are water temperature ( $T_{\text{w}} = -1.8^\circ\text{C} = \text{constant}$ ), water salinity ( $S_{\text{water}} = 33 \text{ g kg}^{-1} = \text{constant}$ ), surface temperature, ice salinity, snow density, and ice and snow thickness.

## 2.2 SMOS data

SMOS is an Earth Explorer mission of the European Space Agency (ESA). The SMOS satellite was launched in November 2009 and achieves a global coverage every three

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days. The SMOS payload is a passive microwave 2-D-interferometer: the Microwave Imaging Radiometer using Aperture Synthesis (MIRAS). MIRAS measures the microwave radiation emitted from the Earth's surface at a frequency of 1.4 GHz in the L-band (Kerr et al., 2001). The corresponding wavelength is 21 cm. Every 1.2 s a two-dimensional snapshot is obtained, which contains observations under various viewing angles between 0 and 65°. The field of view is a hexagon-like shaped area about 1000 km across (Kerr et al., 2001). The resolution at the centre of the field of view (i.e. at nadir view) is about 35 km and decreases to about 50 km at the edge of the field of view (i.e. at incidence angles of 65°). The radiometric accuracy of single measurements is 2.1 to 2.4 K (M. Martin-Neira, personal communication, 2013). MIRAS measures brightness temperatures at full polarisation. Thus, all parameters of the Stokes vector are provided.

In this study, we use SMOS Level 1C Version 505 data. The Level 1C product contains multi-angular brightness temperatures at the top of the atmosphere. The Level 1C data are geolocated in an equal-area Discrete Global Grid (DGG) system called ISEA 4H9 (Pinori et al., 2008). This grid is characterised by a uniform distance of 15 km between the cells. For every SMOS grid point, we first gather all Level 1C brightness temperatures and the auxiliary data provided during one day. The latitude, the longitude, and the grid point number of each measurement are stored together with the brightness temperatures at full polarisation, the incidence angles, the Faraday rotation angles, and the georotation angles. The Level 1C brightness temperatures are given in the antenna reference frame, and thus have to be transformed to the Earth reference frame if the horizontal and the vertical polarisation are considered. This correction for the geometrical rotation is combined with the correction for Faraday rotation (Zine et al., 2008). The MIRAS radiometer alternately measures one or two of the four Stokes vector components per snapshot. For the data processing of SMOS brightness temperatures in this study, we thus use the four Stokes vector components, as well as the Faraday rotation and the georotation angles from subsequent snapshots to calcu-

late the horizontally and vertically polarised brightness temperatures for the considered SMOS grid point.

SMOS measurements are affected by Radio Frequency Interference (RFI) that originates from radars, TV stations, and radio transmission (Oliva et al., 2012). Until now, it is not clear how to deal most effectively with the problems caused by RFI contamination. There are several algorithms under development (e.g. Camps et al., 2010; Oliva et al., 2012; Misra and Ruf, 2012). In this study, we use the following approach. We collect all horizontally and vertically polarised SMOS brightness temperatures measured within a SMOS grid point. These brightness temperatures are sorted according to their incidence angles. In a first step, we exclude all brightness temperatures that are not in the range of 50–280 K, because we do not expect natural radiation to take values outside of this range for sea ice applications. If, after this step, we have more than ten brightness temperatures for each polarisation and cover at least an incidence angle range of 10° per grid point, we use the remaining brightness temperatures to calculate a third order polynomial fit of the SMOS brightness temperatures as a function of incidence angle. The polynomial fit is calculated separately for the horizontal and the vertical polarisation. All brightness temperatures that deviate more than 15 K from this polynomial fit are excluded from the further analysis. Due to the transformation from the antenna reference frame to the Earth reference frame, the brightness temperatures at horizontal and at vertical polarisation are interdependent and in both exclusion steps both polarisations are excluded, even if only one of them does not fulfill the conditions.

The value of 15 K for the exclusion of brightness temperatures is determined after visual inspection of example cases and is only a rough estimation. Furthermore, fitting the brightness temperatures at horizontal and vertical polarisation as a function of incidence angle by two separate third order polynomials can be problematic, especially for low incidence angles, where horizontally and vertically polarised brightness temperatures should be nearly equal. However, we do not expect our RFI filter processing to introduce significant errors.

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## 2.3 The IceBridge flight campaign

For simulation of brightness temperatures, we use ice and snow thicknesses measured during the NASA Operation IceBridge mission (Kurtz et al., 2012). We use the data from the flight campaign that took place from 14 March to 2 April 2012 in the Arctic.

The NASA IceBridge Sea Ice Freeboard, Snow Depth, and Thickness (IDCSI2) data set contains derived geophysical data products including sea ice thickness retrieved from the Airborne Topographic Mapper (ATM) laser altimeter, and snow depth retrieved from the IceBridge snow radar. Simultaneously, the ice surface temperature was measured by a KT19.85 infrared pyrometer alongside the ATM instrument. The IceBridge flight tracks and the measured snow thicknesses are depicted in Fig. 11.

The ice thickness measurements with the ATM laser altimeter have a footprint size of about 1 m × 268 m. In spring 2009, airborne ATM laser altimeter measurements and temporally and spatially coincident in-situ sea ice thickness measurements were conducted on sea ice north of Greenland (GreenArc campaign). The mean sea ice thickness retrieved from laser altimeter data and the mean in-situ sea ice thickness agreed within 5 cm, while the modal values agreed within 10 cm. From this analysis, the uncertainty associated with IceBridge sea ice thickness estimates is assumed to be 40 cm (Farrell et al., 2012). The average ice thickness of all IceBridge ice thickness measurements obtained between 14 March and 2 April 2012 is about 4 m, and the flight tracks were mainly located over multi-year ice. Only a very small fraction of the measurements was carried out over ice thicknesses below 1 m ice thickness. Thus, the IceBridge measurements are not suitable for a validation of the ice thickness retrieval from SMOS brightness temperatures. However, we here use the IceBridge measurements to examine whether we can use our radiation model to realistically simulate L-band brightness temperatures over snow-covered thick sea ice.

The footprint size of the IceBridge snow radar is about 15 m × 16 m. The GreenArc campaign served also as a validation for the IceBridge snow thicknesses. The mean snow thickness retrieved from radar data and the mean in-situ snow thickness agreed

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within 1 cm, while the modal values agreed within 2 cm (Farrell et al., 2012). The correlation coefficient between the snow thicknesses was  $r = 0.7$ , i.e.  $r^2 = 0.49$  (Farrell et al., 2012). The agreement between airborne and in-situ measurements was very good over level ice, while the observed differences were larger over multi-year ice, particularly at pressure ridges. In these regions, snow thicknesses differed by up to 20–60 cm (Farrell et al., 2012). A comparison of the mean radar derived snow thicknesses with climatological snow thicknesses (Warren et al., 1999) revealed a difference of 0.3 cm over multi-year ice and of 16.5 cm over first-year ice (Kurtz and Farrell, 2011).

The footprint size of the KT19.85 infrared pyrometer is about 40 m (Kurtz, 2012). Here, we use the KT19.85 temperature data (Krabill, 2012) to estimate the surface temperature of (snow-covered) sea ice. The uncertainty of the surface temperature data is assumed to be around 0.5 K (Kurtz, 2012). Additionally, in case of clouds or fog underneath the aircraft, the instrument measures the cloud temperature instead of the ice surface temperature (Kurtz, 2012).

### 3 Model simulations and sensitivities

In this section, we first use our emission model to theoretically investigate the impact of a snow layer on brightness temperatures over thin sea ice and thus on the ice thickness retrieval from L-band measurements. For thin ice, the brightness temperature is mainly sensitive to the ice layer's thickness. Whether the brightness temperature of thick ice depends on the snow layer thickness is investigated in the second part of this section.

#### 3.1 Thermal insulation and dielectric properties of snow

We use our radiation model to simulate brightness temperatures over snow-covered sea ice for typical Arctic winter conditions: snow density is set to  $\rho_s = 260 \text{ kg m}^{-3}$ , and the snow is assumed to be dry (wetness  $W_v = 0\%$ ) and to have a thickness of 10% of the ice thickness (Doronin, 1971). The surface temperature is  $T_{\text{surf}} = -20^\circ\text{C}$ . Sim-



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temperatures are less sensitive to ice temperature at higher temperatures. For surface temperatures higher than about  $-10^{\circ}\text{C}$ , the thermal insulation even causes a decrease of brightness temperature (for the conditions considered here). At vertical polarisation, the brightness temperatures of warmer sea ice are thus very similar for snow-covered and snow-free sea ice. While at horizontal polarisation the overall effect of a snow layer is still an increase of the brightness temperature of snow-covered ice compared to snow-free ice, because the contribution of the dielectric properties is higher than the contribution by thermal insulation of snow.

Because the emission model by Burke et al. (1979) is based on the radiative transfer equation, the model does not converge to the correct solution for layer thicknesses approaching zero (Menashi et al., 1993). Thus, we see a jump in the brightness temperature from open water to a very thin ice layer, as well as from bare sea ice to sea ice that is covered by a very thin snow layer (Fig. 1). Moreover, in further studies we found that the emission model after Burke et al. (1979) is not suitable for considering multiple layers within ice, because the model neglects higher order reflection terms (Maaß, 2013). Therefore, the emission model after Burke et al. (1979) has been compared to a coherent model described in Ulaby et al. (1981) that is based on the Maxwell equations and accounts for higher order reflection terms (Maaß, 2013). Except for the first few cm of ice and snow layer thickness the brightness temperatures from these two models agreed well. Thus, we think that although the emission model after Burke et al. (1979) neglects higher order reflection terms and does not describe the transition from a non-existing layer to a very thin layer (a few cm), our emission model reproduces well the brightness temperature changes caused by a layer of snow on top of sea ice.

### 3.2 Snow thickness

After investigating the impact of a snow layer with a typical thickness on top of relatively thin ice in the previous part, we now investigate whether the brightness temperature over thick ice depends on the thickness of the snow layer. Therefore, we use our model to calculate the brightness temperature over 4 m thick ice as a function of snow thick-

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ness. In one simulation we account for the thermal insulation effect of snow, in the other simulation we consider only the dielectric properties of snow. As in the previous investigation, snow density is  $\rho_d = 260 \text{ kg m}^{-3}$ , and the snow is assumed to be dry. The simulations are performed for two surface temperatures ( $T_{\text{surf}} = -15^\circ\text{C}$  and  $T_{\text{surf}} = -30^\circ\text{C}$ ). Here, we consider horizontally polarised brightness temperatures at an incidence angle  $\theta = 45^\circ$ .

We find that, if we consider only the dielectric properties of snow, the thickness of the snow layer does not influence brightness temperatures in the considered range of snow thicknesses (Fig. 2). However, if we take into account the thermal insulation by snow, the brightness temperature increases with increasing snow thickness. A thicker snow layer has a higher insulation effect and thus the bulk ice temperature under a thick snow layer is higher than under a thin snow layer. In our investigations the brightness temperature thus increases by 5.4 K when the snow thickness increases from 0 to 50 cm for  $T_{\text{surf}} = -30^\circ\text{C}$ . For higher temperatures ( $T_{\text{surf}} = -15^\circ\text{C}$ ), brightness temperature is less sensitive to snow thickness, and the brightness temperature increases by 2.3 K, when the snow thickness increases to 50 cm.

For thin ice, the sensitivity of brightness temperature to ice thickness and to snow thickness are similar (not shown here). Thus, we cannot distinguish between an increasing ice and an increasing snow thickness. For thick ice (as compared to the maximum retrievable ice thickness), the sensitivity of brightness temperature to snow thickness is about an order of magnitude higher than the sensitivity to ice thickness. Thus, we here focus on the influence of snow thickness on brightness temperatures over relatively thick sea ice.

In order to test the validity of our theoretical considerations, we in the following section simulate brightness temperatures over snow-covered sea ice and compare these brightness temperature simulations with SMOS brightness temperature measurements. In the subsequent section, we investigate whether brightness temperatures as observed by SMOS over thick Arctic sea ice depend on snow thickness and whether there is a potential for retrieving snow thickness from SMOS observations.

## 4 Comparison of brightness temperature simulations and SMOS observations

We compare two different brightness temperature simulations with brightness temperatures measured by SMOS. For one simulation our emission model is applied to one ice layer and the ice thickness information from IceBridge measurements. In this simulation we neglect a potential snow cover. For the other simulation our emission model is applied to one layer of ice that is covered by one layer of snow; the ice and snow layer thicknesses are taken from the IceBridge measurements.

We simulate brightness temperatures for every SMOS grid cell that contains at least 50 single IceBridge measurements. For the simulations, we use the mean values of all IceBridge ice and snow thickness measurements that are located within the SMOS grid cell as model input for ice and snow thickness. Ice salinity is estimated from ice thickness using an empirical relationship between ice salinity and ice thickness in the Arctic (Cox and Weeks, 1974). This empirical relationship was determined for ice thicknesses up to 4 m. For thicker ice we use the value for 4 m ice thickness, which is  $S_{\text{ice}} = 1.52 \text{ g kg}^{-1}$ . Ice and snow temperatures are estimated from the KT19 ice surface temperatures measured during the IceBridge flight campaign. For the snow-free simulations, we use the mean value between the KT19 ice surface temperature and the water temperature as model input for the ice temperature. For the simulations that include a snow layer, we use equations (11) and (13) to calculate ice and snow temperatures for the model from KT19 ice surface temperatures. As model input for ice concentration, we use ice concentrations that have been retrieved from the 85 GHz channel of SSMIS using the ARTIST Sea Ice (ASI) algorithm (Kaleschke et al., 2001; Spreen et al., 2008). The ice concentration data are given on a polar stereographic grid with 12.5 km grid resolution. The data are 5 day median filtered in order to mitigate unrealistic short-term sea ice concentration variations due to weather effects. For snow density we assume a value of  $\rho_{\text{snow}} = 260 \text{ kg m}^{-3}$ , which is the mean snow density determined from in-situ measurements during the GreenArc campaign (Farrell et al., 2012). According to the IceBridge surface temperatures, the campaign was conducted

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under freezing conditions. Thus, we expect the ice to be covered by dry snow and assume a snow wetness of  $W = 0\%$  in the model.

Because ice concentration and ice temperature have a large influence on the modelled brightness temperatures, we in a second comparison constrain the pixels with respect to (a) ice concentration and (b) ice temperature. (a) The difference between the brightness temperature of thick ice (for our purposes:  $d_{\text{ice}} > 0.5$  m in the Arctic) and water is typically in the order of 130 K. For example, an error of only 5% in the ice concentration would thus cause an error in the brightness temperature of about 6.5 K. The uncertainty of ASI ice concentrations is higher for low ice concentrations than for high ice concentrations. The theoretically expected standard deviation of ice concentration is about 25% for ice concentrations around  $c_{\text{ice}} = 0\%$ , about 13% for  $c_{\text{ice}} = 50\%$ , and about 6% for  $c_{\text{ice}} = 100\%$  (Sprenn et al., 2008). A comparison with in-situ data and high-resolution satellite data revealed a standard deviation of almost 5% for ice concentrations  $c_{\text{ice}} > 90\%$  (Andersen et al., 2007). Thus, in a second comparison, we consider only cases with an almost closed ice cover and use only pixels with ice concentrations  $c_{\text{ice}} \geq 95\%$ , and set the ice concentration in the model to  $c_{\text{ice}} = 100\%$ . (b) Because we do not know how reliable the KT19 surface temperature information is, we also perform the simulations in the second comparison for a fixed surface temperature. Therefore, we calculate the average surface temperature from all KT19 measurements that are included in our analysis. As model input for the surface temperature we then use this average value and include only pixels with surface temperatures that are within one standard deviation of the average surface temperature. The average surface temperature for all pixels with ice concentrations  $c_{\text{ice}} \geq 95\%$  is  $\overline{T_{\text{surf}}} = -32.8^\circ\text{C}$ , and the standard deviation is  $\sigma_{T_{\text{surf}}} = 4.5^\circ\text{C}$ .

Thus, we here compare SMOS brightness temperatures with brightness temperature simulations that neglect and that include a snow layer

1. for all ice concentrations and surface temperatures; ice concentration and surface temperature are variable, and

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2. only for almost completely ice-covered cases ( $c_{\text{ice}} \geq 95\%$ ) and only for surface temperatures within one standard deviation of the average surface temperature ( $-37.3^\circ\text{C} < T_{\text{surf}} < -28.3^\circ\text{C}$ ); the ice concentration is set to  $c_{\text{ice}} = 100\%$ , the surface temperature is set to the average value  $T_{\text{surf}} = -32.8^\circ\text{C}$ .

5 For the comparison with simulated brightness temperatures, we use all SMOS brightness temperatures with incidence angles  $\theta$  between  $0$  and  $60^\circ$ . For each SMOS pixel we average the brightness temperatures measured at  $\theta = 0\text{--}10^\circ$ , for the remaining incidence angles we average the brightness temperatures over  $5^\circ$  incidence angle intervals, i.e. for  $10\text{--}15$ ,  $12.5\text{--}17.5$ ,  $15\text{--}20$ , ...,  $55\text{--}60^\circ$ . The simulations are calculated  
10 for the corresponding mean incidence angles  $\theta = 5$ ,  $12.5$ ,  $15$ ,  $17.5$ , ...,  $57.5^\circ$ .

### 4.1 Results for all ice concentrations and all ice surface temperatures

For this simulation scenario, the simulations that neglect and that include a snow layer differ considerably for horizontal polarisation (Fig. 3). At vertical polarisation, the impact of a snow layer is smaller (Fig. 4). For both polarisations, the modelled brightness  
15 temperatures increase, when a snow layer is added. Contemporaneously, the range of brightness temperatures decreases, when a snow layer is added. At horizontal polarisation, brightness temperatures at low incidence angles ( $\theta < 15^\circ$ ) increase by about  $13\text{K}$ , while brightness temperatures at high incidence angles ( $\theta > 50^\circ$ ) increase by about  $26\text{K}$ , when a snow layer is added. The range of brightness temperatures for different incidence angles decreases from about  $60\text{K}$ , when neglecting the snow cover, to about  $47\text{K}$ , when including the snow cover. At vertical polarisation, brightness temperatures at low incidence angles ( $\theta < 15^\circ$ ) increase by about  $10\text{K}$ , while brightness temperatures at high incidence angles ( $\theta > 50^\circ$ ) increase only by about  $3\text{K}$ , when a snow layer is added. The range of values decreases from about  $35\text{K}$ , when neglecting  
20 the snow cover, to about  $28\text{K}$ , when including the snow cover.

At horizontal polarisation, including the snow layer considerably reduces the deviations between simulated and measured brightness temperatures (Table 1). When ne-

glecting the snow layer, the simulations underestimate the measured brightness temperatures on average by 22.9 K, as compared to an average underestimation by 4.4 K, when the snow layer is included. The correlation coefficients are rather similar for the simulations without and with snow ( $r^2 = 0.44$  and  $r^2 = 0.38$ , respectively). Compared to the horizontal polarisation, at vertical polarisation the correlation coefficients between the simulated and the observed brightness temperatures are considerably smaller, as well as the differences between simulations that neglect and that include the snow layer (Table 1).

## 4.2 Results for the closed ice cover cases and a fixed surface temperature

Figures 5 and 6 show the results for comparing only the pixels that are almost completely ice-covered and that have a surface temperature within one standard deviation of the average surface temperature ( $-37.3^\circ\text{C} < T_{\text{surf}} < -28.3^\circ\text{C}$ ). Compared to scenario (1), the root mean square deviations decrease, and the correlation coefficients increase. At horizontal polarisation, the root mean square deviation between simulated and measured brightness temperatures is 20.0 K, when the snow layer is neglected, and decreases to 4.4 K, when the snow layer is included. The correlation coefficients are  $r^2 = 0.58$  without snow, and  $r^2 = 0.61$  with snow.

## 5 Potential for retrieval of snow thickness

The above comparison between measured and modelled brightness temperatures suggests that brightness temperatures observed by SMOS are influenced by the presence of a snow layer on top of the ice. According to the results from the previous sections, brightness temperatures over snow-covered sea ice are independent of snow layer thickness, when only the dielectric properties of the snow layer are considered. However, due to the thermal insulation effect of snow, there is an indirect effect of snow thickness on brightness temperatures.

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## 5.1 Brightness temperatures for different snow thicknesses

In order to investigate this indirect impact of snow layer thickness on observed brightness temperatures, we divide the IceBridge measurements into five snow thickness classes and consider the corresponding SMOS brightness temperatures. We choose the snow thickness classes such that each class is represented by approximately the same amount of observations. The snow thickness classes are (1)  $d_{\text{snow}} = 4\text{--}12$  cm, (2)  $d_{\text{snow}} = 12\text{--}20$  cm, (3)  $d_{\text{snow}} = 20\text{--}26$  cm, (4)  $d_{\text{snow}} = 26\text{--}32$  cm, and (5)  $d_{\text{snow}} = 32\text{--}44$  cm. Every snow thickness class contains between 128 and 282 SMOS grid cells. These grid cells contain at least 50 IceBridge snow thickness measurements and contemporaneous SMOS brightness temperatures. SMOS brightness temperatures are averaged over incidence angle ranges of  $5^\circ$  (except for the incidence angles averaged over  $\theta = 0\text{--}10^\circ$ ), as was done in the previous section.

For comparison, we not only consider the observed brightness temperatures, but also simulate brightness temperatures for the snow thickness classes 1 to 5. We use fixed values for the ice concentration, ice thickness, ice salinity, and ice surface temperature. We use only SMOS grid cells with  $c_{\text{ice}} \geq 95\%$  and set the ice concentration in the model to  $c_{\text{ice}} = 100\%$ . For ice thickness in the model, we use the average value of all IceBridge ice thickness measurements with  $c_{\text{ice}} \geq 95\%$ , which is  $\overline{d_{\text{ice}}} = 4$  m. Accordingly, the ice salinity in the model is set to  $S_{\text{ice}} = 1.5 \text{ g kg}^{-1}$  (Cox and Weeks, 1974). For the surface temperature, we use the average value of all KT19 values, i.e.  $T_{\text{surf}} = -32.8^\circ\text{C}$ .

The mean brightness temperatures, averaged over the whole incidence angle range, for the average snow thicknesses of the five snow thickness classes are shown in Fig. 7. In general, the observed brightness temperatures increase with increasing snow thickness for both horizontal and vertical polarisation. The mean brightness temperature observations at horizontal polarisation increase by 1.4 K from snow thickness class 1 to 2, by 2.7 K from snow thickness class 2 to 3, by additional 1.0 K for snow thickness class 4, and another 0.7 K for snow thickness class 5. At vertical polarisation, observed

mean brightness temperatures increase by 1.8 K from snow thickness class 1 to 2, and by 2.3 K from 2 to 3. For snow thickness classes 3 to 5, i.e. for snow thicknesses  $d_{\text{snow}} = 20\text{--}44$  cm, the vertically polarised brightness temperatures are very similar and differ by only 0.4 K.

5 Brightness temperatures increase more pronouncedly from snow thickness class 2 to 3, than from 1 to 2. This reflects the fact, that the average snow thicknesses of snow thickness classes 1, 2, and 3 are 10.2 cm, 15.3 cm, and 23.4 cm, respectively. Thus, snow thickness from 1 to 2 increases on average by 5.1 cm, while snow thickness from 2 to 3 increases on average by 8.1 cm. For the other cases, the increase of brightness  
10 temperatures with increasing snow thickness is smaller for higher snow thicknesses. When we compare the observed brightness temperatures with the modelled brightness temperatures, the mean deviations at horizontal polarisation are 2.1 and 1.7 K for the snow thickness classes 1 and 2, respectively; the mean deviations for the snow thick-  
15 ness classes 3 to 5 are between 0.0 and 0.3 K. The horizontally polarised brightness temperatures are thus on average slightly overestimated by the model, when compared to the observations. At vertical polarisation, we find that the model systematically over-  
20 estimates the observed brightness temperatures by 4.3 to 6.1 K for the five snow thickness classes. When we try different values for the ice temperature, snow density, and ice salinity (not shown here), we find that only the ice salinity impacts the modelled  
25 brightness temperatures thus that the deviations between the observations and the model are more evenly distributed for horizontal and for vertical polarisation, such that the horizontally polarised brightness temperatures are slightly underestimated and the vertically polarised brightness temperatures are slightly overestimated by the model. However, this is only the case for a very low ice salinity of  $S_{\text{ice}} = 1 \text{ g kg}^{-1}$ , because brightness temperatures are very sensitive to ice salinity for low ice salinities (Maaß, 2013). For ice salinities  $S_{\text{ice}} \geq 1.5 \text{ g kg}^{-1}$ , the brightness temperatures are only slightly influenced by ice salinity. Because we can expect an ice salinity of  $1 \text{ g kg}^{-1}$  to be a too low value for the average ice salinity of the sea ice in the IceBridge campaign area, we

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expect the results shown here to be representative for our sea ice radiation model in its current state.

The angular dependency of the observed and simulated brightness temperatures for the five snow thickness classes are shown in Fig. 8. For low incidence angles  $\theta < 20^\circ$ , modelled brightness temperatures are about 5–8 K higher than the observed brightness temperatures. This is in accordance with reports about problems with the SMOS brightness temperature processor that cause brightness temperatures for low incidence angles to be some K too low (M. Martin-Neira, personal communication, 2012).

## 5.2 Comparison of retrieved and measured snow thicknesses

The results from the previous sections suggest that brightness temperatures at vertical polarisation are less affected by the presence of a snow cover than the brightness temperatures at horizontal polarisation. The brightness temperature's sensitivity to snow thickness appears to be similar for horizontal and vertical polarisation. However, our radiation model reproduces the absolute values of the observed brightness temperatures of snow-covered thick sea ice considerably better for horizontal than for vertical polarisation. Thus, we here focus on investigating whether horizontally polarised brightness temperatures as observed by SMOS have the potential for the retrieval of snow thickness over thick Arctic sea ice. For comparison, we also consider the retrieval as obtained from vertically polarised brightness temperatures.

In order to investigate how successfully we can retrieve snow thickness over thick sea ice from SMOS, we use different simulation scenarios to simulate brightness temperatures at horizontal and at vertical polarisation over a range of incidence angles. In these simulation scenarios, we use fixed values for all model input parameters and perform the simulations for different snow thicknesses ( $d_{\text{snow}} = 0\text{--}70$  cm). These simulated brightness temperatures are then compared with observed SMOS brightness temperatures over a range of incidence angles. The retrieved snow thickness is the

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snow thickness that is related with the simulation that has the lowest root mean square deviation between the simulated and the observed brightness temperatures.

For the retrieval, we use only pixels that are almost completely ice-covered ( $c_{\text{ice}} \geq 95\%$ ) and have a surface temperature  $-37.3^\circ\text{C} < T_{\text{surf}} < -28.3^\circ\text{C}$ . Thus, we can assume an ice concentration  $c_{\text{ice}} = 100\%$  and use a constant surface temperature. For the remaining model input parameters, we assume different values and perform 15 different scenarios. Thus, we pretend not to have exact information on the ice conditions when we retrieve snow thickness from SMOS data. In Sect. 4, the ice surface temperature is assumed to be  $T_{\text{surf}} = -32.8^\circ\text{C}$ , the ice salinity is  $S_{\text{ice}} = 1.5\text{ g kg}^{-1}$ , the ice thickness is  $d_{\text{ice}} = 4\text{ m}$ , and the snow density is  $\rho_{\text{snow}} = 260\text{ kg m}^{-3}$ . In the scenarios here, the surface temperature takes values between  $-40.15$  and  $-33.15^\circ\text{C}$ , the ice salinity is  $1.5$  or  $2.5\text{ g kg}^{-1}$ , the ice thickness is between  $3$  and  $5\text{ m}$ , and the snow density takes values between  $200$  and  $320\text{ kg m}^{-3}$ . Additionally, we consider simulations over the incidence angle range  $15\text{--}50^\circ$  or  $15\text{--}60^\circ$ . The simulated brightness temperatures are then compared to SMOS brightness temperatures only using data from the day on which the corresponding IceBridge measurements took place, or additionally from the day before and after that day, i.e. we average the SMOS brightness temperatures over three days.

The root mean square deviations and the correlation coefficients between the IceBridge snow thicknesses and the snow thicknesses retrieved from SMOS brightness temperatures for the 15 simulation scenarios are shown in Fig. 9. The corresponding minimum, maximum, and mean values for the root mean square deviations and the correlation coefficients are given in Table 1.

The results from the previous sections suggest that the sensitivity of brightness temperature to snow thickness decreases with increasing snow thickness. Thus, the accuracy of a potential snow thickness retrieval from SMOS brightness temperatures is expected to be higher for lower snow thicknesses. Therefore, we here also compare only SMOS and IceBridge pixels for that the SMOS retrieval returns snow thicknesses  $d_{\text{snow}} < 35\text{ cm}$ . For these lower snow thicknesses, only the results for horizontal polar-

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isation are depicted in Fig. 9 and Table 1, because for vertical polarisation almost all retrieved snow thicknesses are below 35 cm, even if they are not explicitly constrained to these values.

At vertical polarisation, the root mean square deviations and the correlation coefficients between the IceBridge and the SMOS snow thicknesses show an approximately linear relationship (Fig. 9). Lower correlation coefficients coincide with higher root mean square deviations, and higher correlation coefficients coincide with lower root mean square deviations. For the 15 simulation scenarios for vertical polarisation, the correlation coefficients  $r^2$  take values between 0.10 and 0.26, and the root mean square deviations between the IceBridge and the SMOS snow thicknesses range between 17.1 and 23.3 cm. For the 15 simulation scenarios at horizontal polarisation, the correlation coefficients  $r^2$  take values between 0.44 and 0.60, on average we obtain a correlation coefficient  $r^2 = 0.53$ . The root mean square deviations between the IceBridge and the SMOS snow thicknesses range between 14.2 and 25.8 cm, the average value being 17.6 cm.

Thus, all correlation coefficients for horizontal polarisation are higher than for vertical polarisation. The range of correlation coefficients is smaller at horizontal than at vertical polarisation, while the range of root mean square deviations is broader at horizontal than at vertical polarisation. However, the minimum and the mean values of the root mean square deviations for horizontal polarisation are smaller than for vertical polarisation. If we consider only pixels for that the retrieval from horizontally polarised SMOS brightness temperatures gives snow thicknesses  $d_{\text{snow}} < 35$  cm, the correlation coefficients slightly decrease, compared to the case, when we consider all snow thicknesses. The average correlation coefficient is  $r^2 = 0.51$ . The minimum correlation is  $r^2 = 0.32$ , but except for the one simulation scenario that has this low value, all other correlation coefficients  $r^2$  take values between 0.45 and 0.56. The root mean square deviations are between 6.3 and 10.6 cm, the average value being 7.7 cm.

We choose the simulation scenario no. 6 to illustrate the comparison between IceBridge and SMOS snow thicknesses. For this simulation scenario, the root mean

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square deviation is lowest, when we consider only snow thicknesses retrieved to be  $d_{\text{snow}} < 35$  cm. In this simulation, the ice surface temperature is assumed to be  $T_{\text{surf}} = -38.15$  °C, the ice salinity is  $S_{\text{ice}} = 1.5$  g kg<sup>-1</sup>, the ice thickness is  $d_{\text{ice}} = 4$  m, and the snow density is  $\rho_{\text{snow}} = 260$  kg m<sup>-3</sup>. We consider simulations over the incidence angle range 15–50°, and we use SMOS brightness temperatures averaged over three days.

The comparison for all snow thicknesses shows a good average agreement for snow thicknesses up to about 30–35 cm and an overestimation of snow thicknesses, when the thickness retrieval returns higher values (Fig. 10). The minimum detectable snow thickness of the IceBridge snow radar is about 5 cm, thus there are no values below 5 cm for the IceBridge snow thickness. The average Ice Bridge snow thickness is  $\overline{d_{\text{snow}}} = 23.1$  cm, while the average snow thickness from the SMOS retrieval is  $\overline{d_{\text{snow}}} = 32.7$  cm. The correlation coefficient for the snow thicknesses of this simulation scenario is  $r^2 = 0.55$ , and the root mean square deviation is 17.9 cm.

If we consider only snow thicknesses retrieved to be  $d_{\text{snow}} < 35$  cm, the correlation coefficient for the snow thicknesses of this simulation scenario is  $r^2 = 0.55$ , and the root mean square deviation is 6.3 cm. The average IceBridge snow thickness is  $\overline{d_{\text{snow}}} = 18.9$  cm, and the average snow thickness from the SMOS retrieval is  $\overline{d_{\text{snow}}} = 18.2$  cm. Thus, the average snow thicknesses differ by only 0.7 cm.

We use the parameter settings of the above described simulation to produce a first snow thickness map for the Arctic and to compare the spatial distributions of the snow thicknesses as measured during the IceBridge campaign and as obtained from the SMOS snow thickness retrieval from brightness temperatures averaged over 14 to 31 March 2012 (Fig. 11). In accordance with the IceBridge measurements, this first SMOS snow thickness map reveals a thinner snow cover in the Canadian Arctic (about 130 to 180° W longitude), and thicker snow towards the coast of Greenland (about 0 to 120° W).

## 6 Summary and discussion

In this study, we used an emission model developed by Burke et al. (1979) and empirical relationships for the ice and snow permittivities to calculate L-band brightness temperatures of snow-covered sea ice. When we assume values for the ice surface temperature and the ice salinity, the emission model describes the brightness temperature of a slab of ice above a semi-infinite layer of water as a function of the ice thickness. Additionally, a snow layer with a certain density (and wetness) on top of the ice layer can be included in the model. When we added a snow layer on top of the sea ice, we obtained a brightness temperature increase because of the lower reflectivities of radiation at the air-snow and the snow-ice-boundaries as compared to the air-ice boundary. Thus, more of the radiation originating from within the ice is observable above a snow-covered ice layer than above a bare ice layer. The difference between the brightness temperatures of snow-covered sea ice and bare sea ice is larger at horizontal polarisation than at vertical polarisation. Because dry snow is almost transparent in L-band, only a very small fraction of the radiation from the underlying ice is attenuated on its way through the snow layer to the surface, and the thickness of the snow layer does not influence the dielectric properties of the snow layer. The brightness temperature above snow-covered sea ice depends only on the snow layer's thickness because the thickness of the snow layer influences the temperature of the underlying sea ice, which in turn influences the brightness temperature. Thus, there is an indirect dependence between snow thickness and brightness temperature.

In order to test the validity of our results from the theoretical investigations, we used snow and ice thickness measurements from the IceBridge flight campaign in spring 2012 in the Arctic to simulate brightness temperatures and to compare these simulated brightness temperatures with brightness temperatures measured by SMOS. In agreement with the findings from the model simulations in the previous section, we obtained two main findings. Firstly, the observed SMOS brightness temperatures were considerably underestimated when we neglected the snow layer in our model simula-

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tions. Secondly, the horizontally polarised brightness temperatures were more affected by the presence of a snow layer than the vertically polarised brightness temperatures.

The reasons for the deviations between the simulated and the observed brightness temperatures are mainly the remaining uncertainties for the ice and snow thicknesses, the ice concentration, the ice temperature, and the ice salinity. While we had information on the former ice parameters, the ice salinity was only roughly estimated from the ice thickness using an empirical relationship between ice salinity and thickness. The IceBridge measurements were mainly taken over thick sea ice. The average value was 4 m and there were only very few measurements over sea ice with a thickness lower than 1 m. At these high ice thicknesses, ice thickness itself does not have a large impact on brightness temperature. However, the salinity of thick sea ice is usually low, and studies on the brightness temperature's sensitivity to ice salinity variations have shown a very high sensitivity for low ice salinities (Maaß, 2013). Hence, knowledge on ice salinity is more crucial for thick multi-year ice with low salinities, as considered here, than for thin first-year ice with high salinities.

In accordance with our theoretical considerations, we found that observed SMOS brightness temperatures slightly increased with increasing snow thickness. From the snow thickness class with  $d_{\text{snow}} = 4\text{--}12$  cm to the snow thickness class with  $d_{\text{snow}} = 32\text{--}44$  cm, SMOS brightness temperatures, averaged over all incidence angles, increased by 5.8 K at horizontal polarisation, and by 4.3 K at vertical polarisation. The observed increases were somewhat higher than the modelled increases of 3.7 K at horizontal, and 3.8 K at vertical polarisation. For the attempt to retrieve snow thickness from SMOS brightness temperatures, we set up different simulations that assumed different constant values for the model input parameters surface temperature, ice salinity, ice thickness, and snow density. For the model input parameters we assumed values within a range that is likely to occur in nature. Thus, we were able to estimate how the retrieval would perform under the least suitable and under the most suitable assumptions for the above mentioned ice parameters, if we had no exact information on ice conditions. Ideally, we would have performed the simulations for all possible combinations of the

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ice parameters that were varied here. However, we think that the 15 selected scenarios representatively cover the range of conceivable settings and the corresponding results. The surface temperatures were varied only over a range of 7 °C here, because we can assume that for the retrieval of snow thickness from SMOS brightness temperatures, we would have at least some information on ice temperature, for example from air temperatures, which are available from observations or near real-time reanalysis data.

Here, we used a retrieval method based on discretized snow thickness values. More continuous results would have been obtained, if we used a gradient method, for which the deviation between simulated and observed brightness temperatures quickly approaches towards minimum values. Thus, we would not have to compare simulated and observed brightness temperatures for the entire snow thickness range and could resolve snow thicknesses on a finer scale. However, as the accuracy of the snow thickness retrieval is in the order of several cm, the results from a retrieval approach giving finer resolved snow thicknesses would not differ from our findings here. Though, such a gradient based retrieval approach could be more advisable for large-scale retrieval of snow thickness.

A conceivable reason for the observed dependence between snow thickness and brightness temperature is that brightness temperature actually depends on ice thickness (even if the ice is very thick). If this was the case, we would possibly observe a dependence between snow thickness and brightness temperature, because snow thickness is related to ice thickness, as assumed in the calculations of Doronin (1971), who estimated that snow thickness is on average 10% of ice thickness (for ice thicknesses  $d_{ice} > 20$  cm). In order to exclude this possibility, we tried to retrieve ice thickness with the same approach as for the snow thickness (not shown here). The results confirmed that brightness temperatures cannot be mainly attributed to ice thickness, because the correlation between ice thicknesses was negative with  $r^2 \approx 0.2$ , and root mean square deviations were almost 4 m.

Here, the retrieval exploited the SMOS measurement principle of observing brightness temperatures not only under one incidence angle but for a range of incidence

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angles. With the current sea ice radiation model the retrieval was not successful, when we considered only single SMOS measurements or SMOS brightness temperatures from certain incidence angles. One reason is that the SMOS brightness temperatures have a relatively high variability. Thus, we need to average over as many measurements as possible in order to extract any information from the brightness temperatures. A second reason is, that with our radiation model we succeeded to simulate brightness temperatures averaged over a range of incidence angles such that they agreed well with SMOS measurements. However, the SMOS brightness temperatures showed some oscillations, when considered as a function of incidence angle. Using our model, at the present state, we were not able to reproduce these oscillations with increasing incidence angle, but only the average brightness temperature over the incidence angle range. These observed oscillations could be related to roughness effects on the ice and snow surface, which are not taken into account by our model.

Our snow thickness map obtained from SMOS measurements showed a good spatial agreement with the IceBridge snow thicknesses. The areas outside the area covered by the IceBridge campaign should be interpreted more carefully, because we used constant values for the model input parameters ice temperature, ice salinity, and snow density, as we estimated them from the conditions found during the IceBridge flight campaign. The highest deviations between SMOS and IceBridge snow thicknesses were found east of Greenland around the longitude of  $0^\circ$ . In this area, surface air temperature data show highly variable temperatures with very warm conditions. When we assume too low temperatures in the retrieval, we obtain an overestimation of snow thickness. Additionally, due to the high variability of temperatures our approach of assuming constant values for the ice temperature may lead to higher uncertainties in the snow thickness retrieval than in areas with more constant temperature conditions. Furthermore, the SSMIS ice concentration data in this area show a considerably higher variability than in the remaining areas. Our approach of excluding SMOS measurements over low ice concentration areas is based on 5 day median filtered SSMIS data. Over the highly variable ice concentration field, this approach may fail and the consid-

ered SMOS pixels may contain water areas. In this case, we would observe considerably lower brightness temperatures and would underestimate snow thickness. A future improved snow thickness retrieval from SMOS brightness temperatures should thus include temperature information and more carefully take into account ice concentration variability.

## 7 Conclusions

According to our emission model, snow has a twofold impact on sea ice brightness temperatures. Firstly, the presence of a snow layer modifies the radiation observed above sea ice, because the reflectivities between the air-snow- and the snow-ice-boundaries are lower than the reflectivity at the air-ice-boundary. Secondly, the thermal insulation by snow modifies the ice temperature and thus the ice permittivity. The first effect causes brightness temperatures above snow-covered sea ice to be higher than above snow-free sea ice. At horizontal polarisation, this increase of brightness temperature increases with increasing incidence angle  $\theta$  and reaches almost 20 K at  $\theta = 45^\circ$  (for  $d_{\text{ice}} = 50$  cm). In contrast, at vertical polarisation, the brightness temperature increase due to the presence of a snow cover decreases with increasing incidence angle. At  $\theta = 45^\circ$ , vertically polarised brightness temperatures of snow-covered and snow-free sea ice are almost identical, if the second effect, the thermal insulation by snow, is neglected. The presence of a snow layer appears as a sudden increase of brightness temperature in our emission model, because the model fails to describe the transition from no snow to a very thin snow cover (of a few cm). Apart from this sudden increase, brightness temperatures in our emission model are nearly independent of the thickness of the snow layer, because snow is almost transparent in L-band. However, this holds only if we neglect the thermal insulation effect by snow. The influence of the snow's thermal insulation on sea ice brightness temperatures depends on the surface temperature conditions. For the relatively cold conditions considered here (ice surface

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warmer than ice covered by a thinner snow layer. This allows us to reasonably estimate the snow thickness over thick sea ice from SMOS brightness temperatures at horizontal polarisation.

*Acknowledgements.* The authors would like to thank the members of the ESA SMOSIce project for helpful discussions. This work was largely supported by the International Max Planck Research School on Earth System Modelling, and in parts through the Cluster of Excellence “CliSAP” (EXC177), University of Hamburg, funded through the German Science Foundation (DFG). SMOS data were provided by the ESA Support to Science Element programme under contract 4000101476. Ice concentration data retrieved from SSMIS measurements were provided by the Integrated Climate Data Center (ICDC), CliSAP/KlimaCampus, University of Hamburg, <http://icdc.zmaw.de>. IceBridge ice and snow thicknesses, and surface temperatures were provided by the NASA Distributed Active Archive Center at the National Snow and Ice Data Center, Boulder, Colorado USA. Digital media. <http://nsidc.org/data/idcsi2.html>.

## References

- Andersen, S., Tonboe, R., Kaleschke, L., Heygster, G., and Pedersen, L.: Intercomparison of passive microwave sea ice concentration retrievals over the high-concentration Arctic sea ice, *J. Geophys. Res.*, 112, C08004, doi:10.1029/2006JC003543, 2007. 3643
- Burke, W., Schmugge, T., and Paris, J.: Comparison of 2.8 and 21 cm microwave radiometer observations over soils with emission model calculations, *J. Geophys. Res.*, 84, 287–294, 1979. 3630, 3631, 3640, 3652
- Camps, A., Gourrion, J., Tarongi, J., Gutierrez, A., Barbosa, J., and Castro, R.: RFI analysis in SMOS imagery, in: *Proceedings IGARSS, 2007–2010*, 2010. 3636
- Comiso, J., Cavalieri, D., and Markus, T.: Sea ice concentration, ice temperature, and snow depth using AMSR-E data, *IEEE T. Geosci. Remote*, 41, 243–252, 2003. 3630
- Cox, G. and Weeks, W.: Salinity variations in sea ice, *J. Glaciol.*, 13, 109–120, 1974. 3642, 3646
- Cox, G. and Weeks, W.: Equations for determining the gas and brine volumes in sea ice samples, *J. Glaciol.*, 29, 306–316, 1983. 3632
- Doronin, Y.: *Thermal interaction of the atmosphere and the hydrosphere in the Arctic*, Coronet-Books, Philadelphia, 1971. 3638, 3654

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- Farrell, S., Kurtz, N., Connor, L., Elder, B., Leuschen, C., Markus, T., McAdoo, D., Panzer, B., Richter-Menge, J., and Sonntag, J.: A first assessment of IceBridge snow and ice thickness data over Arctic sea ice, *IEEE T. Geosci. Remote*, 50, 2098–2111, 2012. 3637, 3638, 3642
- Giles, K., Laxon, S., Wingham, D., Wallis, D., Krabill, W., Leuschen, C., McAdoo, D., Manizade, S., and Raney, R.: Combined airborne laser and radar altimeter measurements over the Fram Strait in May 2002, *Remote Sens. Environ.*, 111, 182–194, 2007. 3629
- Hall, D.: Remote sensing applications to hydrology: imaging radar, *Hydrolog. Sci.*, 41, 609–624, 1996. 3630
- Hallikainen, M.: Microwave radiometry of snow, *Adv. Space Res.*, 9, 267–275, 1989. 3630
- Kaleschke, L., Lüpkes, C., Vihma, T., Haarpaintner, J., Bochert, A., Hartmann, J., and Heygster, G.: SSM/I sea ice remote sensing for mesoscale ocean-atmosphere interaction analysis, *Can. J. Remote Sens.*, 27, 526–537, 2001. 3642
- Kaleschke, L., Maaß, N., Haas, C., Hendricks, S., Heygster, G., and Tonboe, R. T.: A sea-ice thickness retrieval model for 1.4 GHz radiometry and application to airborne measurements over low salinity sea-ice, *The Cryosphere*, 4, 583–592, doi:10.5194/tc-4-583-2010, 2010. 3629, 3630, 3631, 3632
- Kaleschke, L., Tian-Kunze, X., Maaß, N., Mäkynen, M., and Drusch, M.: Sea ice thickness retrieval from SMOS brightness temperatures during the Arctic freeze-up period, *Geophys. Res. Lett.*, 39, L05501, doi:10.129/2012GL050916, 2012. 3629, 3630, 3631
- Kerr, Y., Waldteufel, P., Wigneron, J., Martinuzzi, J., Font, J., and Berger, M.: Soil moisture retrieval from space: the Soil Moisture and Ocean Salinity (SMOS) mission, *IEEE T. Geosci. Remote*, 39, 1729–1735, 2001. 3635
- Klein, L. and Swift, C.: An improved model for the dielectric constant of sea water at microwave frequencies, *IEEE T. Antenn. Propag.*, 25, 104–111, 1977. 3632
- Krabill, W. B.: IceBridge KT19 IR Surface Temperature, online, NASA DAAC at NSIDC, Boulder, Colorado, USA, 2012. 3638
- Kurtz, N.: IceBridge Quick Look Sea Ice Freeboard, Snow Depth, and Thickness Product Manual, 2012. 3638
- Kurtz, N. and Farrell, S.: Large-scale surveys of snow depth on Arctic sea ice from operation IceBridge, *Geophys. Res. Lett.*, 38, L20505, doi:10.1029/2011GL049216, 2011. 3629, 3638
- Kurtz, N., Studinger, M., Harbeck, J., Onana, V., and Farrell, S.: IceBridge Sea Ice Freeboard, Snow Depth, and Thickness, online, NASA DAAC at NSIDC, Boulder, Colorado USA, 2012. 3637

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- Kwok, R. and Cunningham, G.: ICESat over Arctic sea ice: estimation of snow depth and ice thickness, *J. Geophys. Res.*, 113, C08010, doi:10.1029/2008JC004753, 2008. 3629
- Kwok, R., Panzer, B., Leuschen, C., Pang, S., Markus, T., Holt, B., and Gogineni, S.: Airborne surveys of snow depth over Arctic sea ice, *J. Geophys. Res.*, 116, C11018, doi:10.1029/2011JC007371, 2011. 3629
- Maaß, N.: Remote Sensing of Sea Ice Thickness Using SMOS Data, Reports on Earth System Science, available at: [www.mpimet.mpg.de/fileadmin/publikationen/Reports/WEB\\_BzE\\_131.pdf](http://www.mpimet.mpg.de/fileadmin/publikationen/Reports/WEB_BzE_131.pdf), 2013. 3640, 3647, 3653
- Markus, T. and Cavalieri, D. J.: Snow depth distribution over sea ice in the Southern Ocean from satellite passive microwave data, *Antarct. Res. Ser.*, 74, 19–39, 1998. 3629
- Maykut, G. and Untersteiner, N.: Some results from a time-dependent thermodynamic model of sea ice, *J. Geophys. Res.*, 76, 1550–1575, 1971. 3633
- Menashi, J., Germain, K., Swift, C., Comiso, J., and Lohanick, A.: Low-frequency passive-microwave observations of sea ice in the Weddell Sea, *J. Geophys. Res.*, 98, 22569–22577, 1993. 3631, 3640
- Misra, S. and Ruf, C.: Analysis of radio frequency interference detection algorithms in the angular domain for SMOS, *IEEE T. Geosci. Remote*, 50, 1448–1457, 2012. 3636
- Oliva, R., Daganzo, E., Kerr, Y., Mecklenburg, S., Nieto, S., Richaume, P., and Gruhier, C.: SMOS radio frequency interference scenario: status and actions taken to improve the RFI environment in the 1400–1427-MHz passive band, *IEEE T. Geosci. Remote*, 50, 1427–1439, 2012. 3636
- Pinori, S., Crapolicchio, R., and Mecklenburg, S.: Preparing the ESA-SMOS (soil moisture and ocean salinity) mission-overview of the user data products and data distribution strategy, in: *Microwave Radiometry and Remote Sensing of the Environment, MICRORAD*, IEEE, 1–4, 2008. 3635
- Pounder, E.: *The Physics of Ice*, Pergamon Press, the Commonwealth and International Library, Geophysics Division, Oxford, 1965. 3632
- Rott, H. and Mätzler, C.: Possibilities and limits of synthetic aperture radar for snow and glacier surveying, *Ann. Glaciol.*, 9, 195–199, 1987. 3630
- Spreen, G., Kaleschke, L., and Heygster, G.: Sea ice remote sensing using AMSR-E 89 GHz channels, *J. Geophys. Res.*, 113, C02S03, doi:10.1029/2005JC003384, 2008. 3642, 3643

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- Stroeve, J. C., Markus, T., Maslanik, J. A., Cavalieri, D. J., Gasiewski, A. J., Heinrichs, J. F., Holmgren, J., Perovich, D. K., and Sturm, M.: Impact of surface roughness on AMSR-E sea ice products, *IEEE T. Geosci. Remote S.*, 44, 3103–3117, 2006. 3629
- 5 Tiuri, M., Sihvola, A., Nyfors, E., and Hallikainen, M.: The complex dielectric constant of snow at microwave frequencies, *IEEE J. Oceanic Eng.*, 9, 377–382, 1984. 3632
- Ulaby, F., Moore, R., and Fung, A.: *Microwave Remote Sensing: Active and Passive*, vol. 1 – Microwave Remote Sensing Fundamentals and Radiometry, Addison-Wesley, London, UK, 1981. 3640
- 10 Untersteiner, N.: Calculations of temperature regime and heat budget of sea ice in the Central Arctic, *J. Geophys. Res.*, 69, 4755–4766, 1964. 3634
- Vant, M., Ramseier, R., and Makios, V.: The complex-dielectric constant of sea ice at frequencies in the range 0.1–40 GHz, *J. Appl. Phys.*, 49, 1264–1280, 1978. 3632
- Warren, S., Rigor, I., Untersteiner, N., Radionov, V., Bryazgin, N., Aleksandrov, Y., and Colony, R.: Snow depth on Arctic sea ice, *J. Climate*, 12, 1814–1829, 1999. 3629, 3638
- 15 Yu, Y. and Rothrock, D.: Thin ice thickness from satellite thermal imagery, *J. Geophys. Res.*, 101, 25753–25766, 1996. 3634
- Zine, S., Boutin, J., Font, J., Reul, N., Waldteufel, P., Gabarró, C., Tenerelli, J., Petitcolin, F., Vergely, J., Talone, M., and Delwart, S.: Overview of the SMOS sea surface salinity prototype processor, *IEEE T. Geosci. Remote*, 46, 621–645, 2008. 3635

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**Table 1.** Root mean square deviations RMSD, mean deviations MD, and correlation coefficients  $r^2$  for simulated brightness temperatures and brightness temperatures as measured by SMOS for horizontal and vertical polarisation.

			Scenario 1	Scenario 2
H-Pol	RMSD [K]	no snow	24.6	20.0
		snow	7.8	4.4
	MD [K]	no snow	22.9	18.5
		snow	4.4	−0.9
	$r^2$	no snow	0.44	0.58
		snow	0.38	0.61
V-Pol	RMSD [K]	no snow	8.6	5.2
		snow	5.6	7.4
	MD [K]	no snow	5.8	2.0
		snow	−2.1	−6.5
	$r^2$	no snow	0.25	0.39
		snow	0.19	0.39
Data points $N$			22 798	12 084

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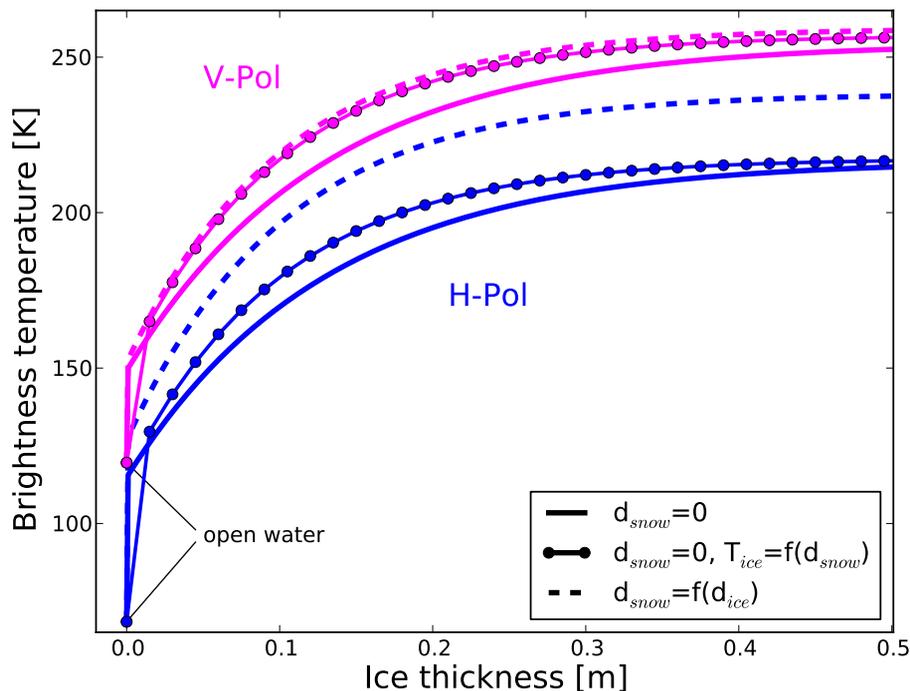
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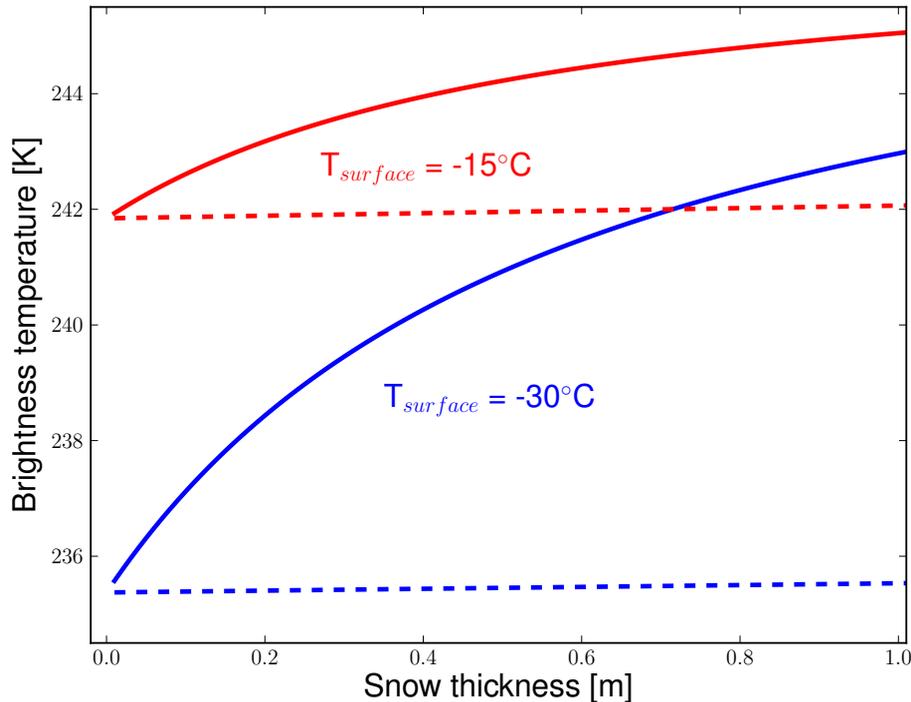
**Table 2.** Minimum, maximum, and mean root mean square deviations RMSD and correlation coefficients  $r^2$  for the IceBridge snow thicknesses and the snow thicknesses as retrieved from SMOS brightness temperatures for the 15 simulation scenarios. The values are given for the retrieval with horizontally and with vertically polarised brightness temperatures, as well as for horizontally polarised brightness temperatures, when only retrieved snow thicknesses  $d_{\text{snow}} < 35$  cm are considered.

		MIN	MAX	MEAN
RMSD [cm]	H-Pol	14.15	25.82	17.64
	V-Pol	17.11	23.25	20.29
	H-Pol ( $d_{\text{snow}} < 35$ cm)	6.33	10.64	7.66
$r^2$	H-Pol	0.44	0.60	0.53
	V-Pol	0.10	0.26	0.18
	H-Pol ( $d_{\text{snow}} < 35$ cm)	0.32	0.56	0.51



**Fig. 1.** Brightness temperature at horizontal (blue) and vertical (magenta) polarisation for an incidence angle  $\theta = 45^\circ$  as a function of ice thickness for three different scenarios: (1) the solid lines show brightness temperatures as modelled for snow-free sea ice, (2) the circles show brightness temperatures as modelled for snow-free sea ice with ice temperatures, as if the snow cover was present, and (3) the dashed lines show brightness temperatures as modelled for snow-covered sea ice.

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**Fig. 2.** Brightness temperature of 4 m thick snow-covered ice as a function of snow thickness at horizontal polarisation at an incidence angle  $\theta = 45^\circ$  (solid line) for an ice surface temperature of  $-30^\circ\text{C}$  (blue) and of  $-15^\circ\text{C}$  (red). The dashed line indicates the brightness temperature of snow-covered ice, when the thermal insulation by snow is neglected.

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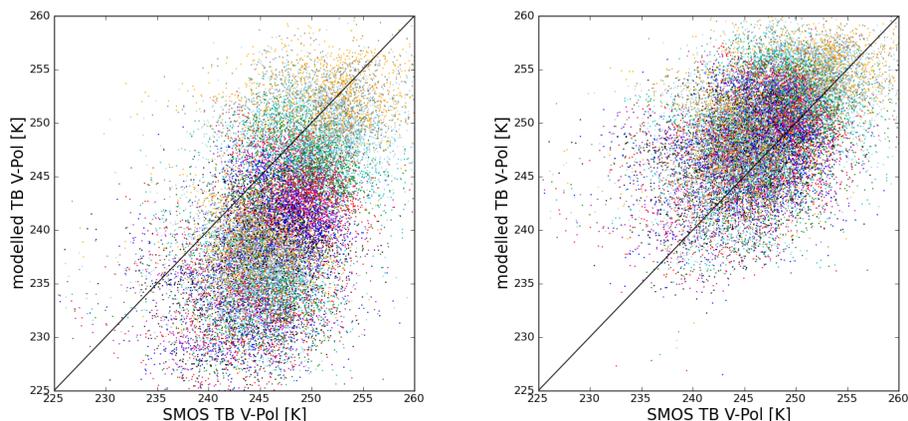
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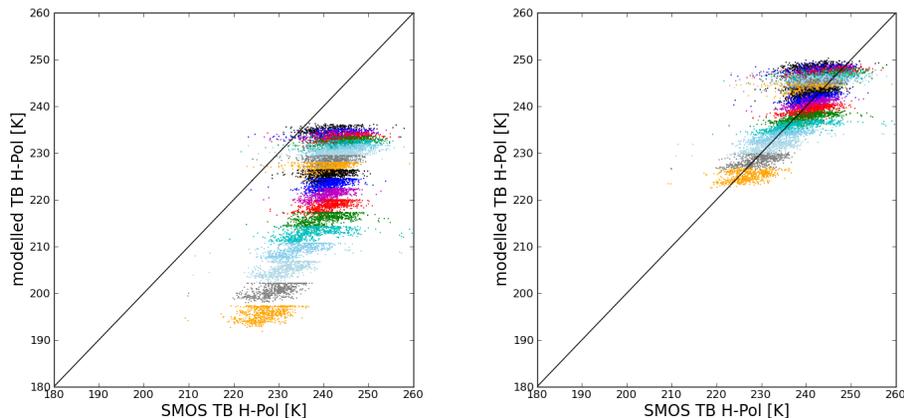
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**Fig. 4.** Vertically polarised brightness temperatures as measured by SMOS vs. brightness temperatures as modelled for IceBridge ice thicknesses (left figure) and as modelled for IceBridge ice and snow thicknesses (right figure) using the model for one ice layer and for one ice and one snow layer, respectively. The colors indicate the incidence angle increasing from  $0^\circ$  at the lower left corner to  $60^\circ$  at the upper right corner of the data cloud.

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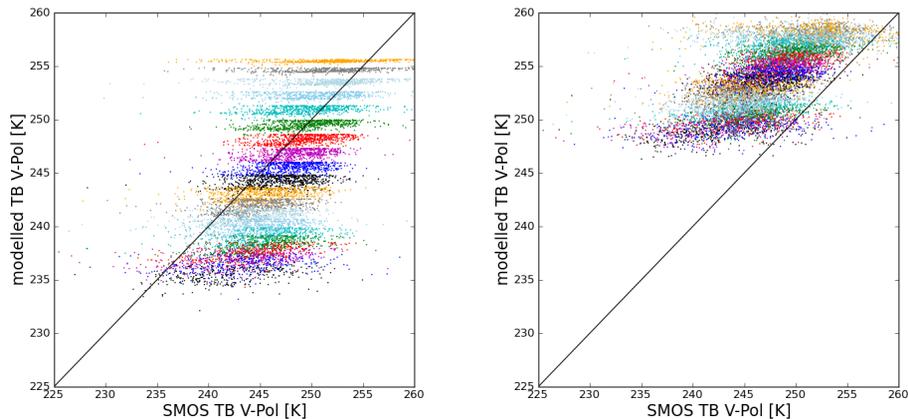
**Fig. 5.** Horizontally polarised brightness temperatures as measured by SMOS vs. brightness temperatures as modelled for IceBridge ice thicknesses (left figure) and as modelled for IceBridge ice and snow thicknesses (right figure) using the model for one ice layer and for one ice and one snow layer, respectively. The colors indicate the incidence angle increasing from  $0^\circ$  at the upper right corner to  $60^\circ$  at the lower left corner of the data cloud. Only data points with  $c_{ice} \geq 95\%$  and  $-37.3^\circ\text{C} < T_{surf} < -28.3^\circ\text{C}$  are included.

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**Fig. 6.** Vertically polarised brightness temperatures as measured by SMOS vs. brightness temperatures as modelled for IceBridge ice thicknesses (left figure) and as modelled for IceBridge ice and snow thicknesses (right figure) using the model for one ice layer and for one ice and one snow layer, respectively. The colors indicate the incidence angle increasing from  $0^\circ$  at the lower left corner to  $60^\circ$  at the upper right corner of the data cloud. Only data points with  $c_{\text{ice}} \geq 95\%$  and  $-37.3^\circ\text{C} < T_{\text{surf}} < -28.3^\circ\text{C}$  are included.

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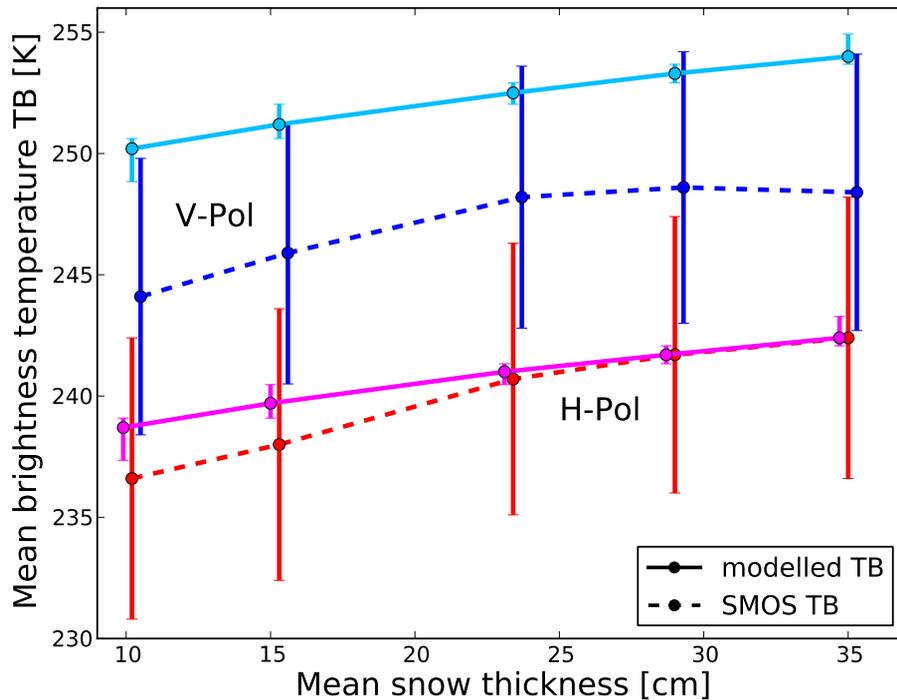
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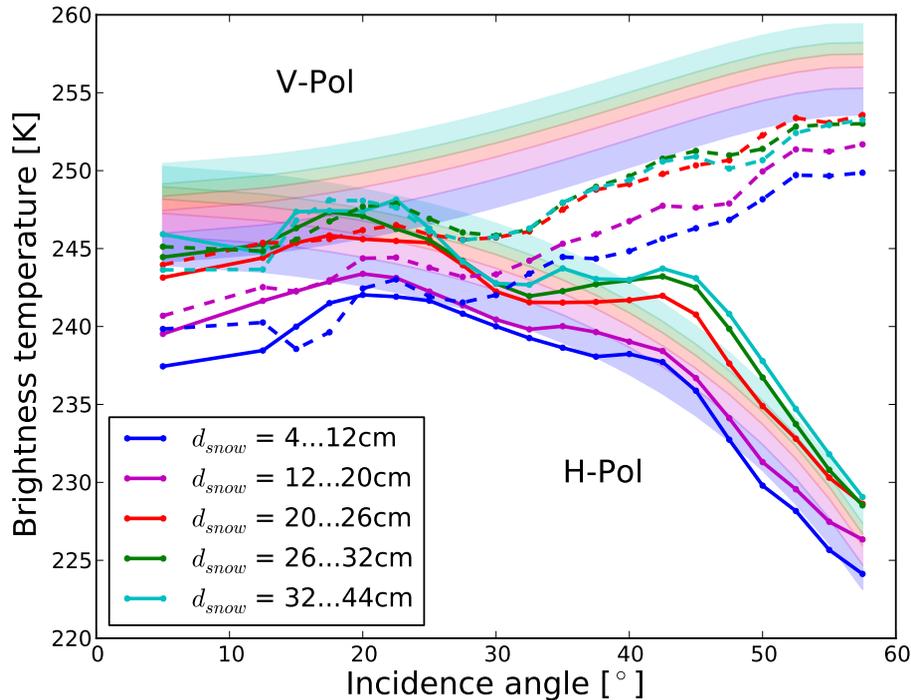
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**Fig. 7.** Brightness temperatures averaged over the incidence angle range  $0\text{--}60^\circ$  as simulated (solid line) and as observed by SMOS (dashed line) at horizontal (reddish colors) and at vertical (bluish colors) polarisation vs. the average snow thickness of the snow thickness classes 1–5. For the SMOS observations, the error bars indicate the average standard deviation of the different incidence angle ranges. For the simulations, the error bars indicate the modelled brightness temperature range for the range of snow thicknesses contained within each class.



**Fig. 8.** Brightness temperatures as simulated and as measured by SMOS for the snow thickness classes 1–5 as given in the figure legend. The shaded areas indicate the brightness temperature simulations, the lines indicate the averaged brightness temperature measurements. The solid lines indicate horizontal polarisation, the dashed lines indicate vertical polarisation.

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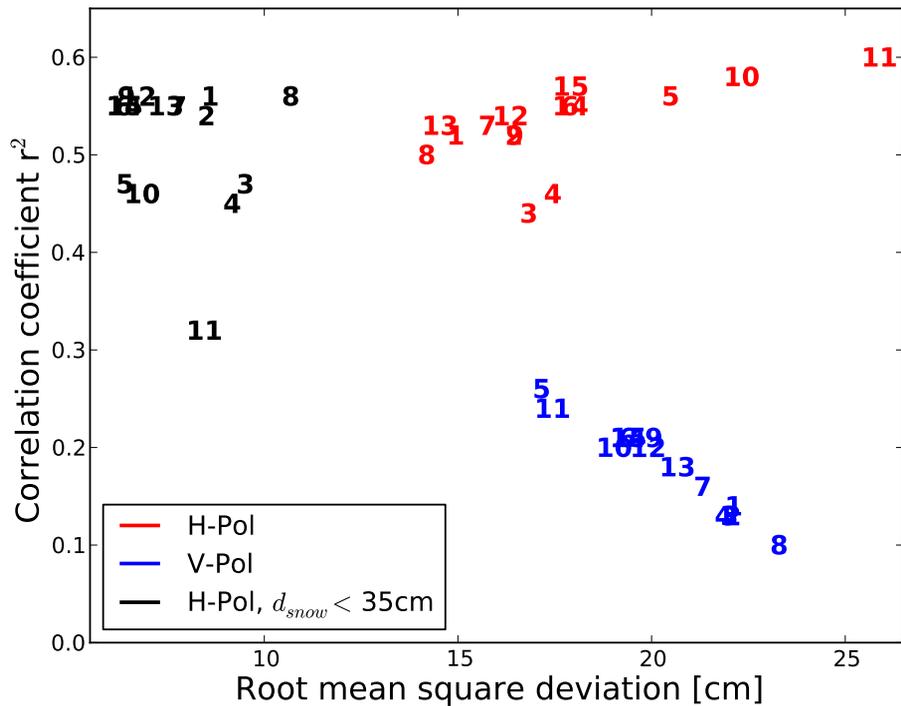
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**Fig. 9.** Correlation coefficients  $r^2$  and root mean square deviations for the IceBridge snow thicknesses and the snow thicknesses retrieved from SMOS brightness temperatures for 15 different simulation scenarios. The numbers give the scenario number. The red and blue numbers show the results for the consideration of all snow thicknesses. The red numbers indicate horizontal, the blue numbers vertical polarisation. The black numbers indicate the results for horizontal polarisation, when only snow thicknesses  $d_{snow} < 35\text{cm}$  (as retrieved from SMOS) are considered.

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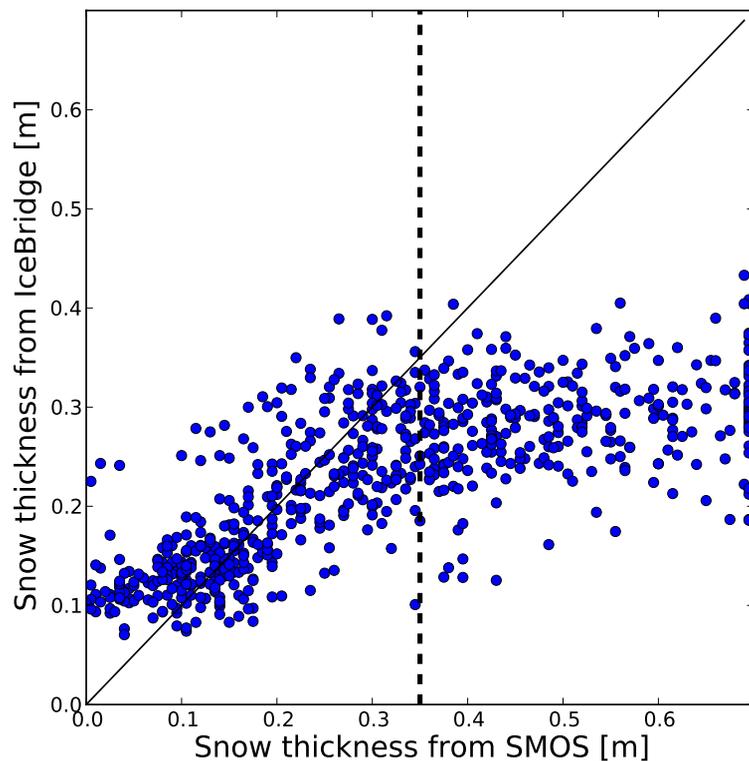
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**Fig. 10.** Snow thicknesses as retrieved from horizontally polarised SMOS brightness temperatures vs. coincident snow thicknesses as measured during the IceBridge campaign. Here, we show the results for the simulation with  $\theta = 15\text{--}50^\circ$ ,  $T_{\text{surf}} = -38.15^\circ\text{C}$ ,  $\rho_{\text{snow}} = 260\text{ kg m}^{-3}$ ,  $d_{\text{ice}} = 4\text{ m}$ , and  $S_{\text{ice}} = 1.5\text{ g kg}^{-1}$  (simulation scenario no. 6). The dashed line indicates the result, if we consider only snow thicknesses for that the retrieval from SMOS brightness temperatures gives snow thicknesses  $d_{\text{snow}} < 35\text{ cm}$ .

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