Simulation of the satellite radar altimeter sea ice thickness retrieval uncertainty

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Abstract

Although it is well known that radar waves penetrate into snow and sea ice, the exact mechanisms for radar-altimeter scattering and its link to the depth of the effective scattering surface from sea ice are still unknown. Previously proposed mechanisms linked the snow ice interface, i.e. the dominating scattering horizon, directly with the depth of the effective scattering surface. However, simulations using a multilayer radar scattering model show that the effective scattering surface is affected by snow-cover and ice properties. With the coming Cryosat-2 (planned launch 2009) satellite radar altimeter it is proposed that sea ice thickness can be derived by measuring its freeboard. In this study we evaluate the radar altimeter sea ice thickness retrieval uncertainty in terms of floe buoyancy, radar penetration and ice type distribution using both a scattering model and “Archimedes’ principle”. The effect of the snow cover on the floe buoyancy and the radar penetration and on the ice cover spatial and temporal variability is assessed from field campaign measurements in the Arctic and Antarctic. In addition to these well known uncertainties we use high resolution RADARSAT SAR data to simulate errors due to the variability of the effective scattering surface as a result of the sub-footprint spatial backscatter and elevation distribution sometimes called preferential sampling. In particular in areas where ridges represent a significant part of the ice volume (e.g. the Lincoln Sea) the simulated altimeter thickness estimate is lower than the real average footprint thickness. This means that the errors are large, yet manageable if the relevant quantities are known a priori. A discussion of the radar altimeter ice thickness retrieval uncertainties concludes the paper.

1 Introduction

Variation in sea ice thickness is a significant indicator for climate change (Wadhams, 1990; Rothrock et al., 2003), but its inter-annual, seasonal and spatial variability is poorly resolved (McLaren et al., 1992). Therefore, much interest is being paid to alter-
native methods for monitoring sea ice thickness for climate monitoring such as satellite radar altimetry on the upcoming CryoSat-2, planned for launch in 2009 (Laxon et al., 2003; Wingham, 1999; Wingham et al., 2006), and laser altimetry using ICESat (Kwok et al., 2006; Kwok and Cunningham, 2008). The ice thickness is derived from altimeters by multiplying the measured freeboard height by an effective snow/ice density factor (the K-factor). It is commonly assumed that radar altimeter signals operating at an electromagnetic frequency of about 13 GHz penetrate to the snow/ice interface. However, for pulse limited space-borne radar altimetry, backscatter modelling indicates that snow depth and density as well as snow and sea ice surface roughness influence the radar penetration into the snow and ice. As a result, the effective scattering surface depth, which is the horizon where the freeboard is measured, can vary as a function of these snow and ice properties (Tonboe et al., 2006a). In addition, snow depth and density and ice density critically affect the floe buoyancy and the chances for estimating sea ice thickness by measuring its freeboard (Rothrock, 1986; Giles et al., 2007). The freeboard height is multiplied by the effective density to estimate the ice thickness for a floe in hydrostatic equilibrium. Actually, the ice floe may not be in hydrostatic equilibrium on a point-by-point basis (Doronin and Kheisin, 1977), and this has consequences for the height measurements using radar as demonstrated in Sects. 4.5 and 4.6 below. However, on a floe to floe basis hydrostatic equilibrium logically is a valid assumption. Several ice thickness point measurements are needed to characterise the ice thickness distribution representative of a particular ice-covered region (Rothrock, 1986; Haas, 2003). The mode of the ice thickness distribution represents the dominating thermodynamically grown thickness of level ice. However, the distribution has a tail towards thicker ice, i.e. deformed ice, and the average may be significantly different from the mode (Haas, 2003). Typical ice thickness distributions are shown in Fig. 1.

As the ice freeboard has to be multiplied by about 10 to obtain the thickness, even small errors in the freeboard retrieval lead to large ice thickness estimation errors (Rothrock, 1986). This multiplication factor derived from the effective density is not constant and may vary between different thickness categories, i.e. new-ice and multiyear
ice, as well as seasonally and regionally (Haas et al., 2006a; Wadhams et al., 1992). Tonboe et al. (2006b) pointed out that the parameters affecting the sea ice freeboard and the radar penetration and ice type distribution are not mapped by any routine field campaign. The error-bars on the retrieved ice thickness estimates are needed when the data are assimilated into numerical models or when they are compared to other ice thickness estimates such as those from laser altimeters, submarine sonars, drilling, and electromagnetic induction instruments. It is further important to identify the largest and most important error sources. Rothrock (1986) stated that the uncertainties involved in deriving the ice thickness from its freeboard were too large. However with the advent of modern space borne altimeters the issue has been revisited. Recent error estimates of the ice thickness retrieval uncertainty for both laser (total error 0.76 m) and radar (total error 0.46 m) altimeters by Giles et al. (2007) included errors sources related to the floe buoyancy: i.e. the snow depth, freeboard estimation uncertainty, and the snow, ice and water density. The snow depth estimation error resulting in an ice thickness estimation error of 0.1 m in Giles et al. (2007) for the radar altimeter was the most important of the error sources. The error due to radar penetration was assumed negligible in their budget, and the error due to systematic height and radar backscatter variability within the footprint was not considered. The importance of these two error sources is simulated here using snow and ice measurements and a radar scattering model.

The radar scattering model is a multilayer one-dimensional radiative transfer model where surface scattering is computed at horizontal interfaces (snow surface, icy layers and ice surface), as described in Tonboe et al. (2006a). Propagation speed, attenuation and scattering are computed for each layer. The simulated echo delay due to freeboard variations and the time dependent backscatter intensity which is recorded onboard the satellite are integrated afterwards in a waveform model suitable for pulse limited space borne altimeters to compute the 1/2-power time also called the track point. The 1/2-power time is the point in time mid between the onboard satellite received backscatter noise floor and the maximum signal power. The effective scattering
surface is the level detected by the 1/2-power time. On ice sheets, in regions where surface scattering dominates, the 1/2-power time “retracking” threshold gives a good representation of the mean surface elevation (Davis, 1997). We use the 1/2-power time since surface scattering mechanisms dominate sea ice backscatter. It is a robust measure of the distance to the effective scattering surface: simulations using seasonal output from a thermodynamic model (snow cover parameters but not surface roughness or ice parameters) as input to the backscatter model show that the scattering surface follows the ice surface within about 5 cm during winter (Tonboe et al., 2006b). The model concept is different from single layer scattering models developed for ice sheet backscatter (e.g., Ridley and Partington, 1988) since surface scattering dominates in sea ice i.e. scattering from the snow and ice surfaces and possibly from layers within the snow.

The specific aim of this study is to evaluate the radar altimeter sea ice thickness retrieval uncertainties in terms of both floe buoyancy and radar surface penetration combining a radar scattering model with “Archimedes’ principle” (Archimedes, 287–212 BC). The primary sea ice thickness retrieval uncertainties are identified and discussed in relation to the natural variability from field measurements. Further, the altimeter footprint is not a point measurement, and thus the altimeter elevation measurement as a function of sub-footprint ice elevation and spatial backscatter intensity distribution is simulated using high resolution (50 m) SAR data.

2 Snow and ice properties

In situ data of snow and ice properties in the Central Arctic have always been sparse, but to overcome this problem there has been a long history of expeditions. From 1937 to 1991, the Soviet Union operated the series of North Pole drifting stations on multi-year ice floes (Frolov et al., 2006). In addition to the year round drifting stations the Sever Project collected snow and ice data at on-ice aircraft landing sites from 1928 to 1989. The Sever data were collected primarily during spring and not during sum-
mer melt, i.e. at the end of winter and therefore representing maximum thickness. The measurements were distributed geographically across the Arctic Ocean, but with higher frequency in the Eastern Arctic. The National Snow and Ice Data Center (NSIDC) received a subset of the Sever data also including data from the drifting stations (NSIDC, 2004). The data are described in Warren et al. (1999) and are used here to assess the all-Arctic snow and ice variability. Furthermore, an extensive field programme directed towards ice thickness monitoring was carried out in the GREENICE project in the Fram Strait in April 2003 and in the Lincoln Sea in May 2004 (Haas et al., 2006b). These GREENICE activities were almost coincident and overlapping with the two SAR scenes used in this study; however, ice drift makes direct comparison difficult. The ice thickness and snow thickness data obtained from both Fram Strait and the Lincoln Sea are representative of their respective regions during late winter and spring, and the geographical distribution of these datasets is shown in Fig. 2. These measurements give the total ice thickness and the bulk density. However, layering and vertical variability of the snow cover properties is an inherent part of natural snow packs. The layering, in natural snow packs is formed by individual precipitation events where density is a function of wind speed and temperature during deposition. After deposition temperature gradient metamorphosis increases grain sizes and compaction and temporary melt may form icy layers. Three measured and detailed snow profiles from Antarctica are used to simulate the scattering in natural snow packs (Haas et al., 2008).

2.1 Use of SAR imagery

The standard mode RADARSAT SAR data classified into the four surface types (in Table 1) is used to prescribe realistic input ice type distributions. The SAR data classification algorithm is based on fuzzy-logic principles. The classification is done by letting an experienced observer identify selected regions visually as belonging to one of the four surface types, and the fuzzy-logic algorithm uses this information to automatically classify the remaining data. The algorithm and method are further described in Gill and Tonboe (2006).
3 Modelling the depth of the effective scattering surface

The forward model uses a set of snow and ice microphysical parameters for each layer: temperature, layer thickness, density, correlation length (a measure of the snow grain size or the ice inclusion size), interface roughness, salinity, and snow wetness to compute the effective scattering surface. The effective scattering surface is the level detected by the 1/2-power time. The permittivity of dry snow is primarily a function of snow density, and the permittivity of sea ice is primarily a function of salinity and temperature. The permittivity of both materials is computed using the mixing formulae for rounded ice spheres (Mätzler, 1998):

\[
\varepsilon_{\text{eff}} = \frac{2\varepsilon_1 - 2\nu (\varepsilon_2 - \varepsilon_1) + \sqrt{(2\varepsilon_1 - 2\nu (\varepsilon_2 - \varepsilon_1))^2 + 8\varepsilon_1\varepsilon_2}}{4} 
\]

where \( \nu \) is the fraction of volume occupied by inclusions, \( \varepsilon_1 \) is the host permittivity of the material surrounding the inclusions and \( \varepsilon_2 \) is the permittivity of the inclusions. For snow \( \varepsilon_1 \) is the permittivity of air (\( \varepsilon_{\text{air}} = 1 \)), and for saline ice, \( \varepsilon_1 \) is the permittivity of pure ice given in Mätzler et al. (2006). For snow the inclusions are pure ice, and for saline first-year ice the inclusions are brine pockets. The permittivity and also the volume of brine are given in Ulaby et al. (1986). For multiyear ice the host material is saline ice and the inclusions are air bubbles.

When the snow reaches the melting point the absorption by liquid water increases the dielectric loss dramatically. The permittivity of wet snow is given in Ulaby et al. (1986). The same formulation is used for saline snow since the permittivity of brine and fresh liquid water is nearly the same at 13 GHz. The permittivity of snow and ice using the equations above is given in Tables 2–5.

Surface scattering is the scattering at dielectric interfaces such as the air-snow and snow-ice interface. The nadir-looking surface backscatter is a function of the nadir
reflection coefficient $|R(0)|$ and the flat-patch area $F$ (Fetterer et al., 1992), i.e.

$$\sigma_{\text{surf}} = 0.9F |R(0)|^2 \frac{H}{ut},$$

(2)

where $H$ is the satellite height, $u$ the pulse propagation speed (speed of light in air, snow and ice, respectively) and $\tau$ the pulse length. $F$ is the flat-patch area, which is inversely related to roughness (i.e., smooth surface have high $F$). This model assumes that the signal is dominated by reflection processes from relatively small plane areas (flat-patches) normal to the incident signal within the footprint. In the review of different surface scattering models in Fetterer et al. (1992) the approach in Eq. (2) is believed to be “more realistic” than other models. The geometrical optics model, which is an alternative to Eq. (2), makes very similar predictions. The basic concept for all surface scattering models is that the backscatter is a function of reflection coefficient and surface roughness; i.e., when the surface is smooth the backscatter is high, and when the surface is rough then the backscatter is smaller. All models described in Fetterer et al. (1992), including Eq. (2), make that prediction.

The improved Born approximation, suitable for microwave scattering in a dense medium such as snow, is used to compute the volume scattering coefficient (Mätzler, 1998; Mätzler and Wiesmann, 1999). Volume scattering is scattering from particles or inclusions within layers, i.e. snow grains within the snow layers, and air bubbles and brine pockets within the ice layers.

The improved Born approximation for spherical inclusions is (Mätzler, 1998)

$$\sigma_{\text{vol}} \approx \frac{3p_{\text{ec}}^3 k^4}{32} \nu(1 - \nu) \left\{ \frac{(\varepsilon_2 - \varepsilon_1)(2\varepsilon_{\text{eff}} + \varepsilon_1)}{2\varepsilon_{\text{eff}} + \varepsilon_2} \right\}^2,$$

(3)

where $p_{\text{ec}}$ is the correlation length, $k$ the wavenumber, $\nu$ the volume fraction of scatterers, and $\varepsilon_1$, $\varepsilon_2$, $\varepsilon_{\text{eff}}$ are the permittivity of the background, the scatterers, and the layer, respectively. Volume scattering is an important backscatter mechanism for scatterometers operating at 13 GHz and about 50° incidence such as QuikScat SeaWinds.
However, the total altimeter backscatter is dominated by surface (or interface) scattering, and in our altimeter simulations volume scattering is insignificant as a backscatter source. This is in agreement with laboratory experiments showing that at nadir incidence, volume scattering is insignificant as a backscatter source for snow-covered sea ice (Beaven et al., 1995). Though volume scattering is not a backscatter source, it does increase extinction and to some extent the distribution of backscatter between the snow and the ice surface. This distribution and the snow depth do affect the depth of the effective scattering surface (Tonboe et al., 2006a).

No specific correction is applied for antenna gain or pulse modulation in the characterisation of the emitted pulse. We use a geometric description of the footprint area in each layer $i$ as a function of time $t$ from Chelton et al. (2001) for a pulse-limited altimeter,

$$A_i(t) = \frac{\pi u_i t H}{1 + H/R_e} - \frac{\pi u_i (t - \tau) H}{1 + H/R_e}, \quad (4)$$

where the second term is 0 when $t<\tau$. $R_e$ is Earth’s radius (6371 km), $u_i$ is the speed of light in the layer and $H$ is the satellite height (800 km).

The waveform model integrates the time-dependent backscatter from each scattering horizon. The pulse propagation speed, signal extinction and backscatter are computed as the pulse penetrates the profile, and each individual contribution is summed with appropriate time delay. The backscattered energy, $E$, measured at the satellite for each model time-step ($1 \times 10^{-11}$ s), is the sum of the footprint area, $A_i$, multiplied by the layer backscatter coefficient, $\sigma_i$, i.e.

$$E_t = \sum_{i=1}^{n} A_i \sigma_i \quad (5)$$

The layer backscatter coefficient includes volume backscattering though its magnitude is negligible. The backscatter coefficient from each layer is adjusted for extinction using
the radiative transfer approach, i.e.

$$
\sigma_{\text{total}} = \left( \sigma_{i}^{\text{surf}} + T_{i}^{2} \sigma_{i}^{\text{vol}} \right) \prod_{i=1}^{n} \frac{1}{L_{i-1}^{2} T_{i-1}^{2}}.
$$

(6)

$L$ is the loss and $T$ the transmission coefficient where $L_0 = T_0 = 1$ for the first layer and $\sigma^{\text{vol}}$ is the negligible volume backscatter coefficient.

4 Simulation results

4.1 Combining the depth of the effective scattering surface with ice buoyancy

Since both the height of the scattering surface and the floe buoyancy are affected by snow depth and snow density, the scattering model is used together with “Archimedes’ principle” to compute the sensitivity of both simultaneously. The surface roughness affects the height of the scattering surface and the ice density affects primarily the floe buoyancy. Snow measurements are input to the model in order to translate the natural snow variability to simulated range variability. The measurements are from the Russian Sever project (NSIDC, 2004) and the EU GREENICE project (Haas et al., 2006b). The locations of the measurements are shown in Fig. 2.

Natural snow and ice profiles are more complicated than what is indicated by Table 2. Layering is an inherent part of natural snow packs (Wiesmann et al., 1998), and melt-water and brine may be included in the snow. Also, snow grain sizes and sea ice inclusion size may deviate significantly from the values given in Table 2. In order to illustrate the sensitivity to liquid water, icy layers, and snow grain sizes we use a selection of measured profiles from the Weddell Sea (Haas et al., 2008).

The waveform model is used separately to investigate the sensitivity of the elevation measurement to sub-footprint spatial backscatter and elevation variability. Fixed surface elevations and backscatter coefficients for four different surface types given in
Table 1 are used as a look-up table in the simulations. The waveform model is using a $10^{-11} \text{s}$ time-step.

4.2 1-D sensitivity study

Table 2 is a reference for the sensitivity simulation study shown in Fig. 3. Each parameter (ice density, ice surface roughness, snow density and snow depth) is evaluated separately, and both buoyancy and radar penetration effects are included. The profile in Table 2 has a snow freeboard of 0.6 m (water density is 1035 kg/m$^3$). The range over which the parameters are varied is assumed to provide realistic upper and lower bounds. The ranges of surface roughness and ice density values are discussed in Sect. 4.3, and the sensitivity to snow depth and snow density using measurements as input is shown in Sect. 4.4.

4.3 The sea ice density and the surface roughness

Density of multiyear ice varies between 720 and 910 kg/m$^3$, and of first-year ice between 900 and 940 kg/m$^3$, and densities of the submerged part varies between 900 and 940 kg/m$^3$ for both ice types. Typical variability of the ice density is between 5 and 10 kg/m$^3$ (Wadhams et al., 1992). Sea ice density is related to its salinity, temperature and air bubble volume (Timco and Frederking, 1996; Cox and Weeks, 1983). Increasing ice density makes the ice floe sink thus extending the range to the snow surface, as well as the apparent ice surface and the scattering surface (Fig. 3a). Decreasing ice density raises the snow, ice and scattering surface thus shortening range.

Surface roughness is a central model parameter (Dierking et al., 1997). Using Eq. (2) Fetterer et al. (1992) estimates that for realistic backscatter values between 20 and 40 dB the flat-patch-area is between 0.2% and 16%. Some ice types such as multiyear ice (10–20 dB) and deformed ice (10 dB) have even lower backscatter values. Following the reasoning from Fetterer et al. (1992) this means that the flat-patch-area could be as low as 0.02%. The snow surface roughness (flat-patch area) shown in Fig. 3b
does not affect floe buoyancy but instead the vertical distribution of backscattering between the snow and ice surface and thus the effective scattering surface. There is more backscatter from a smoother snow surface (see Eq. 2), and thus the effective scattering surface is lifted leading to shortening the range. It is not known how the snow-ice interface roughness varies in time and space. Therefore it is difficult to assess the importance of this parameter for the effective scattering surface. However, surface roughness is the primary factor affecting the sub-footprint spatial backscatter variability which will be discussed in Sect. 4.5.

4.4 Effect of variation of snow depth and density

The climatology of snow cover on sea ice measured at Russian drift stations is described in Warren et al. (1999): the mean Arctic Ocean snow depth increases during the cold season from September to June from zero to about 34 cm. The maximum snow depth (46 cm in June) is in the Lincoln Sea and there are local minima north of Siberia and Alaska. The mean snow density increases from 250 kg/m$^3$ in September to 320 kg/m$^3$ in May. Regional snow density variations are small (about 25 kg/m$^3$ in May). Maximum snow thickness on 17 mass balance buoys distributed in the Arctic Ocean for the years 1993-2005 (Richter-Menge et al., 2006) averaged 33.6 cm (standard deviation 12.6 cm). The deepest snow was 60 cm. Spatially, the snow depth variability is large on sea ice (Warren et al., 1999).

Deeper snow on sea ice (Fig. 3c) suppresses the ice surface and raises the snow surface. The scattering surface is not suppressed as much as the ice surface. The result is a range extension for deeper snow. The snow depth distributions from the GREENICE experiments and the Sever project are shown in Fig. 4. The snow depth records are input to the model replacing the snow thickness value in Table 2. In the computations this comprises both radar penetration and profile buoyancy for the range of measured values. The snow thickness distribution is shown in Fig. 4, and the resulting range variability for this particular distribution is shown in Fig. 5. The differences in simulated range variability in Fig. 5 between the multiyear-all-Arctic Sever data and...
the regional GREENICE datasets are small. The range variability using the measured data input is about 6 cm.

Snow density affects the buoyancy (dashed line in Fig. 3d) as well as the reflection coefficient at both the snow and ice surface, hence also the distribution of backscatter (see Eq. 2). The scattering surface is therefore lifted for greater snow density (the dash-dot line in Fig. 3d). The result is a slightly shorter range for densities above 300 kg/m$^3$. Measured snow densities from the Sever project are then used as input with the other values from Table 2 in the model, computing both radar penetration and profile buoyancy. The range of measured snow densities is shown in Fig. 6 with a median of 330 kg/m$^3$ and values between 170 kg/m$^3$ and 460 kg/m$^3$. The simulated range variability is shown in Fig. 7 for two different snow thickness values: 0.2 m, as in Table 2, and 0.3 m. The simulated range variability using the measured data input is about 6 to 8 cm.

### 4.5 Effect of two ice types within the footprint

High backscatter from limited smooth areas within the footprint can dominate the total altimeter backscatter coefficient as well the height of the effective scattering surface because the backscatter is nonlinear function of the surface roughness (Fetterer et al., 1992).

In a simulation experiment, shown in Fig. 8, multiyear ice, 3 m thick, is mixed with new ice, 0.1 m thick (see Table 1). The footprint is totally ice covered with these two types. It is assumed that when only one ice type covers the entire footprint its elevation and thickness can be estimated perfectly with the altimeter 1/2-power time. The backscatter coefficient for the multiyear ice is 15 dB and for the new ice is 10 times higher: 25 dB (Fetterer et al., 1992). For a 99% multiyear ice-cover with 1% new ice the simulated altimeter ice thickness is underestimated by 26 cm compared to the average ice thickness. The simulated backscatter coefficient of this mixture is 15.4 dB compared to 15 dB for multiyear ice and 25 dB for new ice.
4.6 Simulation of the sub-footprint ice type distribution using SAR data

SAR scenes were analysed in the two different ice regimes in the Fram Strait and in the Lincoln Sea (Gill and Tonboe, 2006). A map is shown in Fig. 2. Fig. 9a and b are typical scenes for each of the regions. In situ observations during the GREENICE field activities within the frame of both scenes in 2003 and 2004 at the time of acquisition indicate that the classification is realistic. Figure 9a shows the distribution of ice types in the Lincoln Sea in May 2004. The ice-cover is complete and dominated by multiyear ice (84.4%) with fractions of new-ice (1.3%), first-year ice (2.9%) and pressure ridges (11.4%). Figure 9b shows the distribution of ice types in Fram Strait in April 2003. Also here the ice-cover is complete and dominated by multiyear ice (63.9%) with fractions of new-ice (1.5%), first-year ice (33.9%) and pressure ridges (0.7%).

The SAR images are divided into 250 m × 7000 m altimeter footprints (this size corresponds to CryoSat’s sea ice mode), and these are then used as input to the waveform model. Each of the four ice types is assigned a fixed backscatter coefficient, using values from Fetterer et al. (1992) given in Table 1. The ice thickness for each of the four types roughly corresponds to the modes of the ice thickness distribution measured within the 2003 SAR frame (Haas et al., 2006b). The ice thickness distribution with modes of around 0.1 m, 1 m and 3 m is shown in Fig. 1.

The backscatter and ice thickness look-up-table values are summarised in Table 1. It is assumed that when only one ice type covers the entire footprint its elevation and thickness can be estimated perfectly with the altimeter as in the example in Sect. 4.4 (the “retracking” threshold is tuned to the ice surface and its density is known). Fewer than 0.5% of the “footprints” in the SAR scenes are covered by just one ice type. Table 7 and 8 show the percentage of footprints where only a small fraction (1%, 5% and 10%) of the footprint is covered by other ice types. In the Fram Strait (Table 8) there are footprints covered almost entirely by first-year ice or multiyear ice while in the Lincoln Sea (Table 9) there are only multiyear ice footprints. Footprints entirely covered by new-ice and ridges are not found in the two scenes. The average thickness is computed

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within each footprint, and the simulated ice thickness is estimated using the waveform model, Table 1, and the footprint ice type distribution in the SAR image. The simulated thickness distributions from each scene are shown in Figs. 10 and 11. The effects of snow are not included. The measured ice thickness distributions are shown from both the Lincoln Sea and the Fram Strait in Fig. 1.

Figure 12a and b shows that in the Lincoln Sea, where the fraction of ridges largely determines the average footprint ice thickness in our simulation, the simulated altimeter ice thickness is much less than the true ice thickness in areas with ridges.

In the Fram Strait, shown in Fig. 12c and d, the simulated altimeter ice thickness is lower than the average footprint thickness in particular areas with mixed ice types, i.e. multiyear ice, first-year ice and new-ice.

It is possible that sub-SAR-resolution surface roughness patterns on multiyear ice can have similar effect on the elevation estimate. While melt ponds typically have smooth surfaces, hummocks have rough surfaces (Onstott, 1992). Hummocks have higher freeboard than refrozen melt ponds and though this pattern is not resolvable in our SAR data it represents the same height and backscatter pattern as thin and thick ice does. Therefore a multiyear ice floe with deeper melt ponds may appear thinner than a floe with not so deep melt ponds due to the preferential sampling of the melt ponds.

4.7 Measured snow and ice profiles from the Weddell Sea

Snow depths on Antarctic sea ice are large compared to the Arctic, and temporary melt may occur even in winter, which means that the snow packs can have large vertical diversity. However, liquid water and layering in the snow-pack are also relevant for Arctic sea ice, especially during spring. Three measured snow profiles in the Weddell Sea were selected due to their different backscatter intensity in QuikScat SeaWinds satellite scatterometer data. The profiles are examples of highly metamorphous snow packs (Haas et al., 2008). It turns out that the three profiles belong to three different snow and ice thickness categories. The ice thickness is 0.8 m, 1.25 m and 2.2 m for the
three profiles, respectively. The snow correlation length ($p_{sc}$) is converted from snow grain size and snow type measurements according to Fig. 2, in Wiesmann et al. (1998, p. 275). The surface roughness was not measured but has been set constant to 0.5% for all layer surfaces. Other estimated parameters in the profiles are marked with (*). In particular the icy layers have been assigned the density 800 kg/m$^3$ and the ice density is set to 900 kg/m$^3$. The permittivity and the partial backscatcer from each layer are computed using the model. As stated in Eq. (2) the backscatter is a function of surface roughness and the reflection coefficient. Roughness is constant here and the snow reflection coefficient is primarily a function of snow density.

Profile 1 in Table 3 has 0.21 m snow on top of 0.8 m ice. The upper snow layers are wet. Profile 2 in Table 4 has 0.42 m snow on 1.25 m ice. Profile 3 in Table 5 has 0.53 m snow on 2.2 m ice. The penetration depth at 13 GHz in all three profiles is less than the snow depth. The penetration depth, where the unit power beneath the surface has decreased to 1/e (0.37), was computed using the radar model. Penetration depth is 0.06 m in profile 1, 0.33 m in profile 2, and 0.46 m in profile 3. The absolute freeboard was not measured in situ but the estimated ice thickness differences are compared in Table 6. Two effective density factors ($K=5$ and $K=6$) were tested. For the given density profiles and radar penetration $K=5$ gives the smallest differences. It is noted that the three thickness categories are distinguishable, i.e. the range from the satellite to the effective scattering surface given by the 1/2-power time is longer for profile 1 than for profile 2 and 3. The simulated waveform for each profile is shown in Fig. 13. All three profiles have icy layers and liquid water in the snow pack and most of the simulated backscatter from each layer comes from the top of the snow pack. In particular, icy layers and attenuation by liquid water are important for the vertical distribution of backscatter.
5 Discussion and conclusions

It is important to identify the most significant error sources so that these can be monitored and the ice thickness estimates corrected accordingly. Even though there are some differences in recent estimates of the variability of the snow and ice parameters affecting the floe buoyancy and the ice thickness retrieval uncertainty estimates, it is clear that four parameters are important: snow depth, ice density, freeboard estimation error and snow density (Giles et al., 2007; Kwok and Cunningham, 2008). The simulations in this study show that the radar penetration variability and the preferential sampling error are as important error sources as those affecting the floe buoyancy.

5.1 Buoyancy and penetration

Our simulations including both the radar penetration and the floe buoyancy show that when snow depth and snow density increase and the ice freeboard (snow/ice interface) is lowered, then the effective scattering surface is raised compared to the reference case (Fig. 3c and d). This means that the errors due to buoyancy may partly be compensated by radar penetration and the total error is therefore moderate. Nevertheless, for these two snow parameters it will be important to correct for both buoyancy and radar penetration. Further, using snow and ice climatology for correcting the ice thickness estimate may, on average, reduce the total error; it introduces, however, an ice thickness estimate bias for any snow cover or ice density deviating from the climatology. Therefore, using climatology it will be difficult to distinguish climate or interannual snow thickness and ice density variability on the one hand from ice thickness anomalies on the other hand.

5.2 Spatial variability and preferential sampling

Figure 8 (average thickness) shows that the high backscatter magnitude from the thinnest ice within the footprint largely determines the elevation of the effective scat-
tering surface. Because of this preferential sampling of the thinner ice types it will be important to measure just one ice type at a time. In fact it may be possible to identify echoes from surfaces which are a mix of ice types (Giles et al., 2007). However, nearly all footprints in the two SAR scenes are a mix of different ice types. Because of the limited spatial extent of both new ice and ridges it is not possible to find footprints which are more than 50% covered by these two types within the two SAR scenes. This may be acceptable for sampling new ice because of its high backscatter, but the ridge freeboard will be significantly underestimated. A significant part of the ice volume is found in ridges (Haas, 2003). In the SAR image from the Lincoln Sea ridges occupy 11% of the ice-covered area, which is 31% of the volume assuming thickness values from Table 1. The simulations using the SAR data show that although the ridges are significant for the average ice thickness, they have low backscatter intensity and are therefore underestimated in the simulated altimeter thickness estimate by the preferential sampling of thinner ice. In the Fram Strait, primarily a mixture between level multiyear ice and first-year ice, the average footprint thickness distribution is a mirror image of the simulated altimeter thickness estimate. It thus seems from the simulations that this error can be minimized in regions where one ice type dominates and where ridges are insignificant for the ice volume.

5.3 Snow and ice parameters relevant for the ice thickness retrieval uncertainty

Clearly, systematic monitoring is needed for key parameters such as for snow thickness and density, ice density, surface roughness and ice type distribution, in order to distinguish altimeter ice thickness anomalies from the noise introduced by these other parameters.

The sub-footprint spatial variability of backscatter intensity and surface elevation creates a bias towards the smooth parts of the ice, i.e., new and first-year ice, along with refrozen melt ponds on multiyear ice. This error is dependent on the mixing of surface types within the footprint. Modes in Figs. 10 and 11 are separated by up to 2 m. However errors will be largest in regions where ridges represent a significant part of the ice
volume.
Few studies have investigated the sea ice density variability (Timco and Frederking, 1996). Typical ice density variability of 5 to 10 kg/m$^3$ is mentioned in Wadhams et al. (1992) while Timco and Frederking (1996) state that the density can vary between 900 and 940 kg/m$^3$. The density varies with salinity, temperature and gas volume, i.e. ice type, which indicates that there can be a systematic variation from region to region. Ice density variability of 5, 10 or 20 kg/m$^3$ results in errors in the estimated ice thickness of 0.17 m, 0.34 m or 0.68 m, respectively.

The snow cover affects the buoyancy and the effective scattering surface depth. Using snow cover measurements as input to the model, the range variability is about 6 cm for both snow depth and density. Multiplying by an effective density and radar penetration factor for the snow and ice system (K-factor) of 5 this gives a 0.3 m ice thickness variability for both parameters. The effective scattering surface depth is thus affected by the snow and ice surface roughness. However, we were not able to find any measurements of surface roughness to account for both ice and snow surface. The simulations for a range of values indicate that the range variability is about 8 cm, giving an ice thickness error of 0.4 m with $K = 5$.

Today these parameters are not all monitored systematically. A systematic effort is needed to evaluate the changes in the Arctic Ocean sea ice observed by current and planned satellite altimeters.

Acknowledgements. This work was supported by the EU GREENICE (EVK2-2001-00280) and the EU DAMOCLES (018509) projects.

References


Simulation of the satellite radar altimeter sea ice

R. T. Tonboe et al.


### Table 1

Elevation and backscatter values from Fetterer et al. (1992), the values are used as a look-up-table for the four ice types identified in the SAR scenes: new ice, first-year ice, multiyear ice and ridges.

<table>
<thead>
<tr>
<th></th>
<th>New ice</th>
<th>First-year ice</th>
<th>Multiyear ice</th>
<th>Ridges</th>
</tr>
</thead>
<tbody>
<tr>
<td>Thickness [m]</td>
<td>0.1</td>
<td>1.0</td>
<td>3.0</td>
<td>10.0</td>
</tr>
<tr>
<td>Freeboard [m]</td>
<td>0.01</td>
<td>0.1</td>
<td>0.3</td>
<td>1.0</td>
</tr>
<tr>
<td>Backscatter [dB]</td>
<td>25</td>
<td>20</td>
<td>15</td>
<td>10</td>
</tr>
</tbody>
</table>
Table 2. The physical properties of the reference profile. $d$ is the layer thickness, $T$ is the temperature, $D$ is the density, $F$ is the flat-patch area (inversely related to roughness), $p_{ec}$ is the correlation length, $S$ is the salinity, $\varepsilon$ is the permittivity computed with the model.

<table>
<thead>
<tr>
<th>Layer number</th>
<th>$d$ [m]</th>
<th>$T$ [$^\circ$C]</th>
<th>$D$ [kg/m$^3$]</th>
<th>$F$ [%]</th>
<th>$p_{ec}$ [mm]</th>
<th>$S$ [ppt]</th>
<th>$\varepsilon$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0.20</td>
<td>−10</td>
<td>300</td>
<td>0.5</td>
<td>0.1</td>
<td>0</td>
<td>1.5+0.0001i</td>
</tr>
<tr>
<td>2</td>
<td>3.50</td>
<td>−5</td>
<td>900</td>
<td>0.5</td>
<td>0.2</td>
<td>3</td>
<td>3.5+0.06i</td>
</tr>
</tbody>
</table>
Table 3. Profile 1. The snow cover depth is 0.21 m and the ice thickness is 0.8 m. The numbers marked with * are not measured. The total backscatter coefficient from profile 1 is 18.9 dB and the partial backscatter from each layer is given in the right column in percent. Columns marked with S are simulated with the model using the measured input. Layer 4 is an icy layer.

<table>
<thead>
<tr>
<th>Layer No.</th>
<th>$T$ [$^\circ$C]</th>
<th>Dens [kg/m$^3$]</th>
<th>$d$ [cm]</th>
<th>$p_{ec}$ [mm]</th>
<th>$S$ [ppt]</th>
<th>$W_c$ [vol%]</th>
<th>Permittivity $S$</th>
<th>Backscatter [%] $\sigma S$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0.0</td>
<td>274</td>
<td>4</td>
<td>0.1</td>
<td>0</td>
<td>1.74</td>
<td>1.58+0.0587i</td>
<td>70.2</td>
</tr>
<tr>
<td>2</td>
<td>0.0</td>
<td>378</td>
<td>3</td>
<td>0.1</td>
<td>0</td>
<td>3.07</td>
<td>1.82+0.1034i</td>
<td>1.9</td>
</tr>
<tr>
<td>3</td>
<td>0.0</td>
<td>280</td>
<td>3</td>
<td>0.1</td>
<td>0</td>
<td>1.27</td>
<td>1.57+0.0428i</td>
<td>0.5</td>
</tr>
<tr>
<td>4</td>
<td>−1.0</td>
<td>800*</td>
<td>1</td>
<td>0.05*</td>
<td>0</td>
<td>0</td>
<td>2.81+0.0005i</td>
<td>4.8</td>
</tr>
<tr>
<td>5</td>
<td>−1.0</td>
<td>134</td>
<td>3</td>
<td>0.12</td>
<td>0</td>
<td>1.29</td>
<td>1.30+0.0435i</td>
<td>5.9</td>
</tr>
<tr>
<td>6</td>
<td>−1.3</td>
<td>800</td>
<td>1</td>
<td>0.05</td>
<td>0</td>
<td>0</td>
<td>2.81+0.0001i</td>
<td>4.4</td>
</tr>
<tr>
<td>7</td>
<td>−1.9</td>
<td>200</td>
<td>4</td>
<td>0.18</td>
<td>0</td>
<td>0</td>
<td>1.32+0.0001i</td>
<td>2.7</td>
</tr>
<tr>
<td>8</td>
<td>−1.9</td>
<td>200</td>
<td>2</td>
<td>0.15</td>
<td>0</td>
<td>0</td>
<td>1.32+0.0001i</td>
<td>0.0</td>
</tr>
<tr>
<td>9</td>
<td>−3.0*</td>
<td>900*</td>
<td>80</td>
<td>0.15*</td>
<td>7*</td>
<td>0</td>
<td>4.47+0.3858i</td>
<td>9.6</td>
</tr>
</tbody>
</table>
Table 4. Profile 2. The snow cover depth is 0.42 m and the ice thickness is 1.25 m. The numbers marked with * are not measured. The total backscatter coefficient from profile 2 is 25.1 dB and the partial backscatter from each layer is given in the right column in percent. Columns marked with S are simulated with the model using the measured input. Layers 3, 6, and 9 are icy layers.

<table>
<thead>
<tr>
<th>Layer No.</th>
<th>$T$ [°C]</th>
<th>Dens [kg/m³]</th>
<th>$d$ [cm]</th>
<th>$\rho_{ec}$ [mm]</th>
<th>$S$ [ppt]</th>
<th>$W_c$ [vol%]</th>
<th>Permittivity S</th>
<th>Backscatter [%] $\sigma S$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>−0.2</td>
<td>341</td>
<td>2</td>
<td>0.1</td>
<td>0</td>
<td>0</td>
<td>1.61+0.0001i</td>
<td>18.2</td>
</tr>
<tr>
<td>2</td>
<td>−1.3</td>
<td>321</td>
<td>2</td>
<td>0.1</td>
<td>0</td>
<td>0</td>
<td>1.56+0.0001i</td>
<td>0.0</td>
</tr>
<tr>
<td>3</td>
<td>−1.3</td>
<td>800*</td>
<td>1</td>
<td>0.05*</td>
<td>0</td>
<td>0</td>
<td>2.81+0.0005i</td>
<td>21.8</td>
</tr>
<tr>
<td>4</td>
<td>−1.9</td>
<td>379</td>
<td>4</td>
<td>0.1</td>
<td>0</td>
<td>0</td>
<td>1.69+0.0002i</td>
<td>11.9</td>
</tr>
<tr>
<td>5</td>
<td>−2.1</td>
<td>299</td>
<td>5</td>
<td>0.1</td>
<td>0</td>
<td>0</td>
<td>1.52+0.0001i</td>
<td>0.1</td>
</tr>
<tr>
<td>6</td>
<td>−2.1</td>
<td>800*</td>
<td>1</td>
<td>0.05*</td>
<td>0</td>
<td>0</td>
<td>2.81+0.0005i</td>
<td>23.3</td>
</tr>
<tr>
<td>7</td>
<td>−2.1</td>
<td>383</td>
<td>3</td>
<td>0.1</td>
<td>0</td>
<td>0.1</td>
<td>1.71+0.0034i</td>
<td>0.1</td>
</tr>
<tr>
<td>8</td>
<td>−2.0</td>
<td>401</td>
<td>4</td>
<td>0.1</td>
<td>0</td>
<td>0.4</td>
<td>1.75+0.0135i</td>
<td>0.0</td>
</tr>
<tr>
<td>9</td>
<td>−2.0</td>
<td>800*</td>
<td>3</td>
<td>0.05*</td>
<td>0</td>
<td>0.1</td>
<td>2.47+0.0034i</td>
<td>5.0</td>
</tr>
<tr>
<td>10</td>
<td>−2.0</td>
<td>334</td>
<td>2</td>
<td>0.1</td>
<td>0</td>
<td>0.1</td>
<td>1.62+0.0034i</td>
<td>6.1</td>
</tr>
<tr>
<td>11</td>
<td>−2.0</td>
<td>371</td>
<td>4</td>
<td>0.1</td>
<td>0</td>
<td>0.1</td>
<td>1.68+0.0034i</td>
<td>0.0</td>
</tr>
<tr>
<td>12</td>
<td>−1.9</td>
<td>371</td>
<td>4</td>
<td>0.1</td>
<td>1.2</td>
<td>9.4</td>
<td>2.08+0.3167i</td>
<td>1.7</td>
</tr>
<tr>
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<td>371</td>
<td>4</td>
<td>0.1</td>
<td>13.5</td>
<td>9.4</td>
<td>2.08+0.3167i</td>
<td>0.0</td>
</tr>
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<td>371</td>
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<td>0.1</td>
<td>16.9</td>
<td>14.0</td>
<td>2.28+0.4717i</td>
<td>0.0</td>
</tr>
<tr>
<td>15</td>
<td>−3.0*</td>
<td>900*</td>
<td>125</td>
<td>0.15*</td>
<td>7*</td>
<td>0.0*</td>
<td>4.47+0.3858i</td>
<td>0.0</td>
</tr>
</tbody>
</table>
Table 5. Profile 3. The snow cover depth is 0.53 m and the ice thickness is 2.2 m. The numbers marked with * are not measured. The total backscatter coefficient from profile 3 is 25.1 dB and the partial backscatter from each layer is given in the right column in percent. Columns marked with S are simulated with the model using the measured input. Layer 5 is an icy layer.

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>−1.2</td>
<td>200</td>
<td>2</td>
<td>0.07</td>
<td>0</td>
<td>0</td>
<td>1.32+0.0001i</td>
<td>12.7</td>
</tr>
<tr>
<td>2</td>
<td>−1.4</td>
<td>348</td>
<td>5</td>
<td>0.1</td>
<td>0</td>
<td>0</td>
<td>1.62+0.0001i</td>
<td>6.0</td>
</tr>
<tr>
<td>3</td>
<td>−1.8</td>
<td>311</td>
<td>3</td>
<td>0.07</td>
<td>0</td>
<td>0</td>
<td>1.54+0.0001i</td>
<td>0.3</td>
</tr>
<tr>
<td>4</td>
<td>−1.8</td>
<td>295</td>
<td>3</td>
<td>0.07</td>
<td>0</td>
<td>0</td>
<td>1.51+0.0001i</td>
<td>0.3</td>
</tr>
<tr>
<td>5</td>
<td>−1.8</td>
<td>800*</td>
<td>1</td>
<td>0.05*</td>
<td>0</td>
<td>0</td>
<td>2.81+0.0005i</td>
<td>50.1</td>
</tr>
<tr>
<td>6</td>
<td>−2.1</td>
<td>326</td>
<td>5</td>
<td>0.12</td>
<td>0</td>
<td>0</td>
<td>1.60+0.0057i</td>
<td>29.2</td>
</tr>
<tr>
<td>7</td>
<td>−2.1</td>
<td>315</td>
<td>5</td>
<td>0.14</td>
<td>0</td>
<td>0</td>
<td>1.60+0.0192i</td>
<td>0.0</td>
</tr>
<tr>
<td>8</td>
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<td>238</td>
<td>5</td>
<td>0.16</td>
<td>0</td>
<td>0</td>
<td>1.44+0.0047i</td>
<td>0.7</td>
</tr>
<tr>
<td>9</td>
<td>−2.2</td>
<td>227</td>
<td>5</td>
<td>0.18</td>
<td>0</td>
<td>0</td>
<td>1.37+0.0001i</td>
<td>0.2</td>
</tr>
<tr>
<td>10</td>
<td>−2.8</td>
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<td>0.2</td>
<td>0</td>
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<td>0.0</td>
</tr>
<tr>
<td>11</td>
<td>−3.0</td>
<td>240</td>
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<td>0.2</td>
<td>0</td>
<td>0</td>
<td>1.47+0.0222i</td>
<td>0.1</td>
</tr>
<tr>
<td>12</td>
<td>−3.0</td>
<td>235</td>
<td>5</td>
<td>0.2</td>
<td>0</td>
<td>0</td>
<td>1.56+0.1038i</td>
<td>0.2</td>
</tr>
<tr>
<td>13</td>
<td>−3.3</td>
<td>258</td>
<td>4</td>
<td>0.2</td>
<td>0.7</td>
<td>3.08</td>
<td>1.60+0.1038i</td>
<td>0.0</td>
</tr>
<tr>
<td>14</td>
<td>−3.3</td>
<td>900*</td>
<td>220</td>
<td>0.15*</td>
<td>7*</td>
<td>0.0*</td>
<td>4.33+0.3309i</td>
<td>0.5</td>
</tr>
</tbody>
</table>
Table 6. The simulated effective scattering surface height differences are used to compute the apparent ice thickness differences using the effective density factor $K=5$ and $K=6$.

<table>
<thead>
<tr>
<th>Difference between profile number</th>
<th>Real ice thickness difference</th>
<th>Simulated ice thickness difference $K=5$</th>
<th>Simulated ice thickness difference $K=6$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Profile 1–2</td>
<td>0.45 m</td>
<td>0.35 m</td>
<td>0.41 m</td>
</tr>
<tr>
<td>Profile 2–3</td>
<td>0.95 m</td>
<td>1.08 m</td>
<td>1.30 m</td>
</tr>
<tr>
<td>Profile 3–1</td>
<td>1.4 m</td>
<td>1.43 m</td>
<td>1.7 m</td>
</tr>
</tbody>
</table>
Table 7. Footprints (250 m × 7000 m) within the Fram Strait SAR scene (Fig. 9b) dominated by just one surface type either: open water, new-ice, first-year ice, multiyear ice and ridges. The table shows the percentage [%] of homogeneous footprints which are mixed by less than 1, 5 and 10% other surface types, respectively.

<table>
<thead>
<tr>
<th>Fram Strait</th>
<th>Open water</th>
<th>New-ice</th>
<th>First-year ice</th>
<th>Multiyear ice</th>
<th>Ridges</th>
</tr>
</thead>
<tbody>
<tr>
<td>Less than 1% other</td>
<td>0</td>
<td>0</td>
<td>1.4</td>
<td>1.5</td>
<td>0</td>
</tr>
<tr>
<td>Less than 5% other</td>
<td>0</td>
<td>0</td>
<td>3.1</td>
<td>13.8</td>
<td>0</td>
</tr>
<tr>
<td>Less than 10% other</td>
<td>0</td>
<td>0</td>
<td>5.4</td>
<td>23.7</td>
<td>0</td>
</tr>
</tbody>
</table>
Table 8. Footprints (250 m × 7000 m) within the Lincoln Sea SAR scene (Fig. 9a) dominated by just one surface type either: open water, new-ice, first-year ice, multiyear ice and ridges. The table shows the percentage [%] of homogeneous footprints which are mixed by less than 1, 5 and 10% of other surface types, respectively.

<table>
<thead>
<tr>
<th>Lincoln Sea</th>
<th>Open water</th>
<th>New ice</th>
<th>First-year ice</th>
<th>Multiyear ice</th>
<th>Ridges</th>
</tr>
</thead>
<tbody>
<tr>
<td>Less than 1% other</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0.6</td>
<td>0</td>
</tr>
<tr>
<td>Less than 5% other</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>7.5</td>
<td>0</td>
</tr>
<tr>
<td>Less than 10% other</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>25.8</td>
<td>0</td>
</tr>
</tbody>
</table>
Fig. 1. Typical ice thickness distributions measured with a helicopter-borne electromagnetic induction device: Fram Strait, 13 April 2003 (full-line) and Lincoln Sea, 12 May 2004 (dashed line).
Fig. 2. Map of the Arctic Ocean showing the locations of SAR scenes marked with rectangles in Fram Strait and the Lincoln Sea. The yellow crosses are Sever snow depth measurements and red diamonds are Sever snow density measurement locations concentrated in the Kara Sea. The purple lines near the SAR frames show the GREENICE ice thickness survey lines.
Fig. 3. Sensitivity of simulated altimeter range to variability of four ice and snow parameters. The reference profile is given in Table 2. The ice density variability does not affect the depth of the scattering surface and the surface roughness does not affect the floe buoyancy (measured from the height of the snow surface). The snow density and the snow depth affect both buoyancy and scattering surface depth. The scattering surface depth is shown with the dashed line and the buoyancy is shown with the dotted line. The resulting combined buoyancy and penetration range change is shown with the full line. The small scale oscillations are due to numerical rounding errors in the model. (A) shows the simulated range sensitivity to the snow density. (B) shows the simulated range sensitivity to the surface roughness represented by the flat-patch-area in percent [%]. (C) shows the sensitivity to snow depth. (D) shows the sensitivity to snow density.
Fig. 4. Various measured snow thickness distributions. The full line show Sever measurements between 1959 and 1988 primarily March, April and May. The dashed line and the dashed-dotted line shows the GREENICE measurements in April 2003 and May 2004. The measurement locations are shown in Fig. 2.
Fig. 5. The simulated range variability using measured snow depth of Fig. 4 and the other values from Table 2 as input to the model. The median range is subtracted. The full line shows the simulated range using Sever measurements. The dashed line and the dash-dot line shows the simulated range using the GREENICE measurements. Deeper snow gives longer range, see Fig. 3c.
Fig. 6. The snow density distribution from Sever measurements between 1959 and 1971 primarily March, April and May.
Fig. 7. The simulated range variability using Sever snow density measurements shown in Fig. 6 and the other values from Table 2 as input to the model. The full line shows the simulated range with 0.2 m snow as in Table 2 and the dotted line is for 0.3 m snow.
Fig. 8. The average thickness (full line) and the simulated altimeter ice thickness using the waveform model (dashed line) for different fractions of new ice and multiyear ice. The footprint is totally ice covered.
Fig. 9a. A classification of ice types 15 May 2004 in Lincoln Sea based on a Radarsat SAR (standard mode: 100 km × 100 km, 50 m spatial resolution). The position of the scene is marked in Fig. 2. The ice cover is complete and yellow is multiyear ice, white is ridges, red is first-year ice, and green is new ice.
Fig. 9b. A classification of ice types 11 April 2003 in Fram Strait based on a Radarsat SAR (standard mode: 100 km × 100 km, 50 m spatial resolution). The position of the scene is marked in Fig. 2. The ice cover is complete and yellow is multiyear ice, white is ridges, red is first-year ice, and green is new ice.
Fig. 10. The distributions based on the SAR data shown in Fig. 9a in the Lincoln Sea of average footprint ice thickness (dashed line) and simulated altimeter ice thickness estimate (full line).
Fig. 11. The distributions based on the SAR data shown in Fig. 9b in Fram Strait of average footprint ice thickness (dashed line) and simulated altimeter ice thickness estimate (full line).
Fig. 12a. Simulated altimeter ice thickness in meters, Lincoln Sea 15 May 2004. The thickness is computed using the waveform model and the look-up-table values in Table 1. The footprint size is 250 m × 7000 m.
Fig. 12b. Average ice thickness in meters for each altimeter footprint (250 m × 7000 m) Lincoln Sea 15 May 2004. The average thickness is computed from the look-up-table values in Table 1 and from Fig. 9a.
**Fig. 12c.** Simulated altimeter ice thickness in meters, Fram Strait 11 April 2003. The thickness is computed using the waveform model and the look-up-table values in Table 1. The footprint size is 250 m × 7000 m.
Fig. 12d. Average ice thickness in meters for each altimeter footprint Fram Strait 11 April 2003. The average thickness is computed from the look-up-table values in Table 1 and Fig. 9b.

Colorbar for Fig. 12, the number show the thickness in m.
Fig. 13. The simulated pulse form for (1), the thin snow and ice profile, (2), the medium thick snow and ice profile, and (3), the deep snow and thick ice profile. The numbers 1, 2, 3 are placed at the 1/2-power time for each curve. The curves are shifted in time according to the estimated snow freeboard for each profile.