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Observationally constrained surface mass balance of Larsen C ice shelf, Antarctica

Peter Kuipers Munneke¹, Daniel McGrath²-³, Brooke Medley¹, Adrian Luckman⁵, Suzanne Bevan⁵, Bernd Kulessa⁵, Daniela Jansen⁶, Adam Booth⁷, Paul Smeets¹, Bryn Hubbard⁸, David Ashmore⁸, Michiel Van den Broeke¹, Heidi Sevestre⁶, Konrad Steffen¹⁰, Andrew Shepherd⁷, and Noel Gourmelen¹¹

1 Institute for Marine and Atmospheric research, Utrecht University, Utrecht, the Netherlands
2 Geosciences Department, Colorado State University, Fort Collins, CO, USA
3 U.S. Geological Survey, Alaska Science Center, Anchorage, AK, USA
4 Cryospheric Sciences Laboratory, NASA Goddard Space Flight Center, Greenbelt, MD, USA
5 Geography Department, College of Science, Swansea University, Swansea, UK
6 Alfred Wegener Institute Helmholtz-Centre for Polar and Marine Research, Bremerhaven, Germany
7 School of Earth and Environment, University of Leeds, Leeds, UK
8 Centre for Glaciology, Department of Geography and Earth Sciences, Aberystwyth University, Aberystwyth, UK
9 Department of Geography and Sustainable Development, University of St Andrews, St Andrews, UK
10 Swiss Federal Research Institute WSL, Birmensdorf, Switzerland
11 School of Geosciences, University of Edinburgh, Edinburgh, UK

Correspondence to: Peter Kuipers Munneke (p.kuipersmunneke@uu.nl)

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Abstract. The surface mass balance (SMB) of the Larsen C ice shelf (LCIS), Antarctica, is poorly constrained due to a dearth of in situ observations. Combining several geophysical techniques, we reconstruct spatial and temporal patterns of SMB over the LCIS. Continuous time series of snow height (2.5–6 years) at five locations allow for multi-year estimates of seasonal and annual SMB over the LCIS. There is high interannual variability in SMB as well as spatial variability: in the north, SMB is 0.40 ± 0.06 to 0.41 ± 0.04 m w.e. year⁻¹, while farther south, SMB is up to 0.50 ± 0.05 m w.e. year⁻¹. This difference between north and south is corroborated by winter snow accumulation derived from an airborne radar survey from 2009, which showed an average snow thickness of 0.34 m w.e. north of 66° S, and 0.40 m w.e. south of 68° S. Analysis of ground-penetrating radar from several field campaigns allows for a longer-term perspective of spatial variations in SMB: a particularly strong and coherent reflection horizon below 25–44 m of water-equivalent ice and firn is observed in radargrams collected across the shelf. We propose that this horizon was formed synchronously across the ice shelf. Combining snow height observations, ground and airborne radar, and SMB output from a regional climate model yields a gridded estimate of SMB over the LCIS. It confirms that SMB increases from north to south, overprinted by a gradient of increasing SMB to the west, modulated in the west by föhn-induced sublimation. Previous observations show a strong decrease in firn air content toward the west, which we attribute to spatial patterns of melt, refreezing, and densification rather than SMB.

1 Introduction

About 74 % of the grounded ice sheet of Antarctica drains into the Southern Ocean through floating ice shelves (Bindschadler et al., 2011). By buttressing the grounded ice sheet (e.g. Dupont and Alley, 2005; Gagliardini et al., 2010), ice shelves strongly modulate the flux of ice into the ocean, thereby exerting an important control over the contribution of mass variations of the Antarctic Ice Sheet to global sea level.
In recent decades, the collapse of ice shelves along the Antarctic Peninsula (Cook and Vaughan, 2010) was immediately followed by a sustained velocity increase in the glaciers previously feeding these ice shelves, by a factor 3–4 in documented cases (De Angelis and Skvarča, 2003; Scambos et al., 2004; Berthier et al., 2012). It is likely that the current (and growing) mass imbalance of the Antarctic Peninsula (e.g. Shepherd et al., 2012; Harig and Simons, 2015) is due in part to these ice-dynamical adjustments to the loss of ice shelves. It is believed that these ice-shelf collapses along the Antarctic Peninsula have been attributed, at least in part, to warming of the near-surface atmosphere (Morris and Vaughan, 2003). Warming has been faster in this region than the global average since the 1950s (Marshall et al., 2006; Turner et al., 2014), in part caused by a very large regional decadal variability (Turner et al., 2016). It has been hypothesized that enhanced meltwater production at the ice-shelf surface can lead to hydrofracturing, whereby meltwater-filled crevasses open up under the pressure exerted at the crevasse tip by the column of standing meltwater (Scambos et al., 2003; Van der Veen, 2007), and/or where drainage of meltwater lakes induces fracture by strong flexural stresses (MacAyeal and Sergienko, 2013). Kuipers Munneke et al. (2014b) suggested that the conditions for ponding and hydrofracturing depend on the local accumulation and melt fluxes, and their effect on the vertical structure of the firn layer. Hence, there is a need to describe the surface mass balance (SMB) of these ice shelves accurately.

The Larsen C ice shelf (LCIS) is the largest ice shelf of the Antarctic Peninsula, and is located to the east of the north–south-oriented Antarctic Peninsula mountain range (Fig. 1). As the dominant upper-air wind direction is westerly, the LCIS is in the climatological leeside of the mountains. This position gives rise to particular patterns in surface melt, with more melt and the occurrence of meltwater ponding in the inlets directly east of the mountains, and gradually less melt away from the mountains towards the calving front in the Weddell Sea to the east (Trusel et al., 2013; Barrand et al., 2013). The advection of warm, dry air masses over the ice shelf during föhn winds is the likely cause of this surface melt distribution (Luckman et al., 2014). A notable expression of this surface melt gradient is seen in the composition of the firn layer over the LCIS: the smallest amounts of firn air are found in the inlets in the western part of the LCIS (Holland et al., 2011; Ashmore et al., 2017). In Cainen Inlet (Fig. 1), the firn contains a massive subsurface ice layer (Hubbard et al., 2016), influencing ice-shelf temperatures, density, and potentially, flow properties. The occurrence of föhn winds has increased in recent decades (Cape et al., 2015), contributing to the destabilization of Larsen B ice shelf through increased meltwater production.

However, although important, surface melt is not part of the SMB as we define it in Eq. (1). Spatially varying patterns of snowfall and other SMB components can amplify or counteract the effect of surface melt on the observed gradients of firn air. Moreover, along with the ocean-driven basal mass balance, iceberg calving, and glacier inflow, SMB is an essential component of the overall LCIS mass balance. In addition, SMB controls LCIS firn properties such as density and temperature. Considering the importance of SMB, the dearth of published SMB over the LCIS is problematic. In the most comprehensive compilation of Antarctic Peninsula SMB measurements to date (Turner et al., 2002), none were available on the ice shelves themselves. However, four observational records from sites on the grounded ice adjacent to the LCIS are available. Two firn cores were collected on Dollameleon Island (70.58° S, 60.92° W, 398 m a.s.l.), yielding mean SMB values of 0.390 m w.e. year\(^{-1}\) for 1962–1982 (Mulvaney and Wolff, 1993), and 0.404 m w.e. year\(^{-1}\) (Peel and Clausen, 1982). A firn core on Gipps Ice Rise (68.77° S, 60.93° W, 290 m a.s.l.) gives 0.349 m w.e. year\(^{-1}\) (Peel and Clausen, 1982). Stake observations by Rott et al. (1998) revealed a mean SMB of 0.36 m w.e. year\(^{-1}\) on Jason Peninsula (66.25° S, 61.00° W), which forms the northern boundary of the LCIS. Based on an interpolation between these records, Turner et al. (2002) estimated annual SMB at < 0.50 m w.e. year\(^{-1}\) over the LCIS. However, given the prominence of orographic gradients in this region, one must be wary of extrapolating data from these elevated sites: high-resolution climate modelling as well as radar observations show that the SMB field around such features can deviate by up to ±50% relative to that of the flat surrounding terrain (Lenaerts et al., 2014). Finally, a reanalysis for the period 1979–1993 by Turner et al. (1998a) indicates that solid precipitation at a central point on the LCIS was 0.49 m w.e. year\(^{-1}\).

Accumulation is much lower on the eastern side of the Antarctic Peninsula than on its western side. The latter is dominated by slow-moving low-pressure systems over the Bellingshausen Sea, leading to a mostly north-westerly flow of humid and relatively warm air (Turner et al., 2002). Subsequently, precipitation is orographically driven by the Antarctic Peninsula mountains. The largest precipitation events are associated with advection of moist air from midlatitudes at times when a strong low-pressure system develops over the Bellingshausen Sea. In contrast, a continental climate exists on the eastern side. Barrier flow along the orography of the Antarctic Peninsula, resulting from a climatological low-pressure area over the Weddell Sea, leads to predominantly southerly flow over the eastern part of the peninsula (Parish, 1983). Precipitation events over the LCIS are frequent and generally small (Turner et al., 1998a), and associated with (1) lee cyclogenesis over the LCIS and the Weddell Sea immediately east of the LCIS, and (2) active fronts arriving from the north-east or east (Turner et al., 1998b).

Over the past decade, several field campaigns have been undertaken to collect data on firn, SMB, meteorology, and climate over the LCIS. During various field campaigns in 2008, 2009, 2010, 2011, 2014, and 2015, data have been collected using a number of geophysical techniques, including...
ground-penetrating radar, shallow firn coring, snow pits, and continuous snow height observations. Moreover, airborne radar data have been collected over the LCIS in late 2009 and 2010 as part of the NASA Operation IceBridge campaign.

The aim of this paper is to bring together this suite of data sets in order to provide a coherent picture of SMB over the LCIS, focusing on the intra-annual, annual, and decadal timescales. These data can be used to evaluate the performance of atmospheric models over the LCIS (King et al., 2015), such as RACMO2 (van Wessem et al., 2016), AMPS (Powers et al., 2012), or CESM (Lenaerts et al., 2016). Models assessing ice-shelf stability require an estimate of SMB (Powers et al., 2012), or CESM (Lenaerts et al., 2016). Models assessing ice-shelf stability require an estimate of SMB as a boundary condition (e.g. DeConto and Pollard, 2016), and their performance is subject to a correct representation of current melt and SMB over the ice shelf.

2 Data and methods

2.1 Surface mass balance

The specific surface mass balance (in metre water equivalent per year, m w.e. year\(^{-1}\)) is defined as

\[
\text{SMB} = \int \left( \text{PR} - \text{SU}_s - \text{RU} - \text{ER}_d - \text{SU}_d \right) \, d \text{year},
\]

where PR is precipitation, SU\(_s\) is surface sublimation, RU is meltwater run-off, and ER\(_d\) and SU\(_d\) represent erosion and sublimation of drifting snow particles, respectively. Run-off depends on the full liquid water balance, defined as

\[
\text{RU} = \text{RA} + \text{ME} + \text{CO} - \text{RF} - \text{RT},
\]

i.e. the difference between surface melt (ME), rainfall (RA) and condensation (CO) on one hand, and internal refreezing within the snow and firm, RF, and retention by capillary forces, RT, on the other. Note that SMB, as used in this study refers to the mass balance of the entire firn column, not only of the surface.

Following Turner et al. (2002), we add the definition of mass in an annual layer (MAL) of the snow cover:

\[
\text{MAL} = \text{SMB} - m.
\]

Here, \(m\) is the loss from an annual layer to lower or higher layers due to vapour transport or meltwater percolation. If \(m\) is assumed to be small, MAL can be used as an estimate for SMB.

2.2 Sonic height rangers

Since 2009, sonic height rangers at various automatic weather stations (AWS) have been measuring snow surface height (see Table 1 and Fig. 1), yielding records of 3 to 7 years in length. A correction to account for the dependence of sound velocity on temperature is applied using concurrent observations of air temperature. Sometimes, sonic pulses are reflected from the AWS mast, from the datalogger box attached to the mast, or from suspended snow particles during snowfall or snow drift, giving erroneously low snow height readings. Such observations are filtered out, and the resulting data gaps are filled by linear interpolation. Snow height observations are corroborated by manual measurements of the distance between the sonic ranger and the snow surface upon each annual maintenance visit. We assume an uncertainty in the surface height ranger readings of 0.10 m, which mainly represents noise due to small-scale surface roughness, as the accuracy of individual measurements is of the order of 0.01 m.

At all locations, sonic height rangers were attached to an AWS mast. Between 2009 and 2011, additional height rangers were placed at AWS 14 and 15 (see map in Fig. 1), mounted at \(\sim 2\) m above the surface on a separate triaxial construction consisting of three lightweight aluminium poles drilled into the snow at an angle of about 30\(^\circ\) to a depth of 2.5–3.0 m below the surface. After 2011, we make use of the sonic height rangers attached to the AWS itself. We present a continuous snow surface time series that accounts for the occasional raising of the sonic height ranger to avoid burial. The sonic height rangers themselves were always located between 1 and 4 m above the surface.

All snow height data series exhibit data gaps, but unambiguous values for summer, winter, and annual surface height change could be derived for all years and all stations, except for one: the LAR1 time series was interrupted from 16 October to 9 November 2013. We have filled this gap with the mean elevation change of the other four sonic rangers, allowing us to compute a summer and winter surface height change at LAR1 for 2013.

2.3 Snow pits and firn cores

During various field campaigns between 2008 and 2015, snow pits were dug to a depth of \(\sim 2\) m below the surface (Fig. 1). In these snow pits, detailed snow stratigraphy was logged, and vertical profiles of density were recorded using a variety of tools and methods. In total, 22 shallow snow pits – all collected before melt onset – were used for the analysis in this paper. From these 22 snow pits, we collected vertical profiles of density, resampled into 5 cm depth bins.

Table 1. Overview of sonic height rangers operated over the LCIS and used in this study. Locations are shown on the map in Fig. 1.

<table>
<thead>
<tr>
<th>Name</th>
<th>Lat ((^\circ)S)</th>
<th>Lon ((^\circ)W)</th>
<th>Start date</th>
<th>End date</th>
</tr>
</thead>
<tbody>
<tr>
<td>AWS 14</td>
<td>67.01</td>
<td>61.48</td>
<td>Jan 2009</td>
<td>still operational</td>
</tr>
<tr>
<td>AWS 15</td>
<td>67.57</td>
<td>62.13</td>
<td>Jan 2009</td>
<td>Jun 2014</td>
</tr>
<tr>
<td>LAR1</td>
<td>68.14</td>
<td>63.95</td>
<td>Dec 2008</td>
<td>still operational</td>
</tr>
<tr>
<td>LAR2</td>
<td>67.58</td>
<td>63.26</td>
<td>Dec 2008</td>
<td>Nov 2011</td>
</tr>
<tr>
<td>LAR3</td>
<td>67.03</td>
<td>62.65</td>
<td>Aug 2009</td>
<td>Nov 2011</td>
</tr>
</tbody>
</table>
sign an uncertainty to the density observations of 20 kg m\(^{-3}\), which is based on observations presented below.

In addition, three vertical profiles of density up to 11 m depth were collected using a neutron-scattering probe (Morriss, 2008) in 2009, at locations J1, J2, and J4, indicated on the map in Fig. 1. In 2015, a 90 m-long borehole was drilled with hot water and surveyed with an optical televiewer (Hubbard et al., 2008) at the site of AWS14 (referred to as CI-120 in Ashmore et al., 2017). Using the vertical profile of luminosity as a proxy for density (Hubbard et al., 2013), an estimate of firn density is available for the entire length of the borehole.

All of these snow pits and firn cores provide an estimate of the density of the uppermost layers of firn to convert from radar two-way travel time to thickness and from actual layer thickness to SMB in water-equivalent thickness.

2.4 Ground-penetrating radar

Between 2008 and 2015, five ground-based radar surveys were carried out in different locations, covering northern and central portions of the LCIS (Fig. 1) (Luckman et al., 2012; McGrath et al., 2012; Jansen et al., 2013; Kulessa et al., 2014; McGrath et al., 2014). Surveys were carried out using antennas of different frequencies (25, 100, or 200 MHz) depending on the field campaign. Despite the different frequencies, a distinct, spatially continuous reflector was identified in a large portion of the ground surveys, at a median depth of 41 m below the surface. Several assumptions are necessary in order to convert the measured two-way travel time to this reflector (\(\tau\)) to snow accumulation rates.

First, we use an empirical relation between firn density and its dielectric constant (\(\epsilon\); Kovacs et al., 1995), which determines the effective wave velocity through the firn. Thus, depth (\(d\)) is calculated as follows:

\[
d = 0.5c \tau \epsilon^{-0.5},
\]

where \(c\) is the speed of light in a vacuum. As \(\epsilon\) depends on the firn density profile, and the accumulation rate (proportional to \(d\)) affects the density profile, an iterative technique is required to solve for \(d\) in a consistent manner. The firn density profile can exhibit strong spatial variations due to differing surface melt rates and internal refreezing, snow accumulation rates, and air temperatures; however, field observations on Larsen C are sparse. Indirect measurements do exist, however, such as the map of firn air content (FAC) over the LCIS derived by Holland et al. (2011), which provides insight into the spatial variations in firn density. This gridded product represents the firn air content as a column thickness, which requires a firn densification model in order to derive a density profile. Thus, we have modified a scheme developed by Medley et al. (2015) to solve iteratively for a density profile that is consistent with the radar-derived accumulation rates and the FAC derived by Holland et al. (2011). The method relies on the Herron and Langway Jr. (1980) semi-empirical densification model to produce steady-state depth-density profiles. Starting with an initial accumulation rate estimate, a constant initial density, and long-term average tem-
perature from the atmospheric model RACMO2 (Sect. 2.6 below), we model the depth–density profile using the relationship from Herron and Langway Jr. (1980). Accumulation rates are then derived from the radar horizons using that initial density profile, yielding a new long-term accumulation value. This new accumulation rate then replaces our initial estimate and the process repeats until convergence. This results in radar-derived accumulation rates that are consistent with the prescribed density data (i.e. the density profile is dependent on the accumulation rate, so it is necessary to ensure they are mutually consistent).

Because we have additional independent information in the form of the FAC, we use an additional iterative scheme to ensure the modelled FAC matches the gridded FAC from Holland et al. (2011). To accomplish this fitting, we iterate the assumed initial density to reach the desired FAC. We are effectively lumping all changes in the FAC into the surface density, which do not necessarily represent real surface processes. In reality, surface melt is seasonal, and meltwater infiltration and refreezing do not occur right at the surface, resulting in complex ice structures that can be interspersed with firm pockets throughout the vertical firn column (Ashmore et al., 2017). Therefore, in areas of high melt, the scheme will yield unrealistically high surface density estimates in order to accommodate for refreezing within the firn column that is not accounted for in our dry snow densification model. A modelling effort to include the complexity of meltwater infiltration, retention, and refreezing is beyond the scope of this work and, while our scheme is necessarily simplified, it provides reasonable bounds on the snow accumulation over the LCIS sufficient to meet the needs of this work. Using our method, we generate radar-derived accumulation rates that are physically related to the modelled density profile, which, in turn, is consistent with the FAC data of Holland et al. (2011).

To convert the estimated reflector depth to mass, we assume that the reflector is below the pore close-off depth (estimated at 3–9 m within 120 km from the grounding line (Ashmore et al., 2017) and likely around 10–15 m further downstream), and then subtract the firm air thickness of Hol-land et al. (2011) from the reflector depth, yielding an ice-equivalent thickness, which is converted to mass using an ice density of 910 kg m$^{-3}$. Column mass is corrected for differences in acquisition dates of the various radar lines, assuming a mean accumulation of 0.45 m w.e. year$^{-1}$ in recent years. Finally, we assume that the change in reflector depth due to dynamical stretching is negligible.

### 2.5 Airborne snow radar and radar picking

Beginning in 2009, the NASA Operation IceBridge (OIB) campaign has annually surveyed both the Greenland and Antarctic ice sheets using a variety of instruments designed to map the geometry and internal structure of the ice. Two frequency-modulated continuous-wave (FMCW) radar sys-
Figure 2. Sample radar echogram obtained from CReSIS Ku-band radar (14–16 GHz) on board the 16 November 2009 Operation IceBridge flight. Automated surface and subsurface picks are overlaid. Darkest colours indicate air above the snow surface. On the vertical axis, we show the vertical direction relative to an arbitrary level. Nearly all echograms look similar, with very strong surface and a strong to very strong subsurface reflection horizon. The subsurface horizon is interpreted as the end of the last melt season, and it physically represents the transition from dry snow (above) and snow that has strong melt features (ice lenses).

Figure 3. Complete, filtered time series of cumulative surface height change (m) recorded with sonic height rangers at five locations. LAR1, 2, and 3 were equipped with two sonic rangers, plotted in the same colour. Grey bars indicate the summer season used in this paper (1 November–31 March).

3 Results

3.1 SMB estimates from sonic height rangers

We present the complete time series of snow height for the five sonic rangers on the LCIS (Fig. 3). At all locations, the surface height increased quite gradually, suggesting that precipitation occurs in frequent small-magnitude events, rather than in a small number of large accumulation events per year. In the summer months, surface lowering is observed at all stations, consistent with expected melting of the surface snow, and refreezing in the firn below, and with enhanced firn compaction at elevated snow temperatures.

The multi-year records from sonic height rangers can be used to establish in situ estimates of annual and seasonal SMB, provided that a reliable estimate of snow and firn density is available to convert surface elevation increase to mass accumulation. First, we estimate the density of the winter accumulation (i.e. before the start of the melt season) from the 22 snow pits investigated (Sect. 2.3 above). We excluded the part of the density profile below melt layers that likely
is seen at AWS14 at 0.23 ± mostly within error bounds. The highest mean winter SMB increases temperature and summer melt (Cape et al., 2015). SAM is not a good indicator for Föhn is enhanced during negative SAM, which also in-

Combining this value with wintertime (April–October) sonic height ranger observations, we can construct annual time series of wintertime SMB. Figure 5 shows that interannual variability of wintertime SMB is large (standard deviation at 19–44% of the mean for stations with time series longer than 3 years), but all stations show similar year-to-year variations. For example, the winters of 2009 and 2013 stand out as high-SMB winters at all stations. In this time span, we found no discernable temporal or spatial variability in the density of the winter accumulation. We therefore computed one mean vertical profile, shown along with the individual density profiles in Fig. 4a. The vertically averaged (±1 standard deviation) density of the winter accumulation is 360 ± 20 kg m⁻³. Combining this value with wintertime (April–October) sonic height ranger observations, we can construct annual time series of wintertime SMB. Figure 5 shows that interannual variability of wintertime SMB is large (standard deviation at 19–44% of the mean for stations with time series longer than 3 years), but all stations show similar year-to-year variations. For example, the winters of 2009 and 2013 stand out as high-SMB winters at all stations. In this time span, we found no link between SMB and the southern annular mode (SAM, shown as grey bars in Fig. 5), with values of $R^2$ lower than 0.3. Föhn is enhanced during negative SAM, which also increases temperature and summer melt (Cape et al., 2015). However, our data show that SAM is not a good indicator for the occurrence of precipitation on the LCIS.

The multi-year mean winter SMB at each site is summarized in Table 2, with confidence intervals reflecting measurement accuracy, not interannual variability shown in Fig. 5. The differences between the stations are small and mostly within error bounds. The highest mean winter SMB is seen at AWS14 at 0.23 ± 0.02 m w.e. year⁻¹ for 2009–2015, and the lowest at LAR3 at 0.17 ± 0.02 m w.e. year⁻¹ for 2009–2011. The winter SMB is somewhat higher near the ice-shelf margin than farther inland.

A lack of detailed density profiles at the end of the melt season precludes the construction of an annually resolved record of summer SMB (November–March) for each site. However, a multi-year mean annual SMB can be computed by using the cumulative height signal over multiple years and combining that with firn-core-derived density profiles. The mean summer SMB at each site can then be calculated by subtracting the mean winter SMB from the mean annual SMB. This assumes there is no summer run-off and that all surface melt water refreezes within the annual layer of snowfall (which is typically 0.70–1.40 m thick). In Eqs. (1) and (3), this corresponds to assuming that $RU = 0$ and $m = 0$. As an estimate of the density profile required to convert height change to SMB, we take the mean density profile of four available deep density profiles: three firn cores at LAR1, 2, and 3, and an OPTV density log at AWS14. This mean density profile is shown in Fig. 4b.

Annual SMB (see Table 2) is highest for LAR1 (0.50 ± 0.05 m w.e. year⁻¹) and lowest for LAR3 (0.40 ± 0.06 m w.e. year⁻¹). Only the annual SMB value from LAR1 is substantially higher than at the four other locations. If we consider the common period only (2009–2011), LAR1 stands out even more at 0.56 ± 0.07 m w.e. year⁻¹. For summer SMB (1 November–31 March), we see that LAR1 shows the largest values by far at 0.37 ± 0.06 m w.e. year⁻¹, which is ~40% more summer mass gain than the other locations. At the other locations, the summer SMB values are almost identical at 0.19–0.22 m w.e. year⁻¹.

### 3.2 Estimating long-term SMB using ground-penetrating radar

In all ground-based radar surveys of the LCIS, we find a particularly strong reflection horizon at 35–45 m (median 41 m) depth below the surface (Fig. 6). Radar reflection horizons in firn and ice are related to strong contrasts in dielectric proper-
Table 2. Estimates of mean winter (April–October), summer (November–March) and annual SMB (in m w.e. year\(^{-1}\)) for five sites with continuous sonic height ranger observations. The rightmost column shows the period on which the estimated SMB is based. The bottom five rows show results for the longest common period of all records (April 2009–November 2011). The confidence interval reflects the measurement accuracy, not the interannual variability. Winter and annual SMB are derived from sonic height ranger and density observations; summer SMB is derived as the difference between them.

<table>
<thead>
<tr>
<th>Site</th>
<th>Winter SMB</th>
<th>Summer SMB</th>
<th>Annual SMB</th>
<th>Period</th>
</tr>
</thead>
<tbody>
<tr>
<td>AWS14</td>
<td>0.23 ± 0.02</td>
<td>0.19 ± 0.04</td>
<td>0.41 ± 0.04</td>
<td>Apr 2009–Oct 2015</td>
</tr>
<tr>
<td>AWS15</td>
<td>0.21 ± 0.02</td>
<td>0.23 ± 0.04</td>
<td>0.45 ± 0.04</td>
<td>Apr 2009–Mar 2014</td>
</tr>
<tr>
<td>LAR1</td>
<td>0.21 ± 0.02</td>
<td>0.29 ± 0.05</td>
<td>0.50 ± 0.05</td>
<td>Apr 2009–Oct 2014</td>
</tr>
<tr>
<td>AWS14</td>
<td>0.24 ± 0.03</td>
<td>0.19 ± 0.05</td>
<td>0.44 ± 0.07</td>
<td>Apr 2009–Nov 2011</td>
</tr>
<tr>
<td>AWS15</td>
<td>0.21 ± 0.03</td>
<td>0.22 ± 0.06</td>
<td>0.43 ± 0.06</td>
<td>Apr 2009–Nov 2011</td>
</tr>
<tr>
<td>LAR1</td>
<td>0.19 ± 0.02</td>
<td>0.37 ± 0.06</td>
<td>0.56 ± 0.07</td>
<td>Apr 2009–Nov 2011</td>
</tr>
<tr>
<td>LAR2</td>
<td>0.20 ± 0.02</td>
<td>0.22 ± 0.06</td>
<td>0.42 ± 0.06</td>
<td>Apr 2009–Nov 2011</td>
</tr>
<tr>
<td>LAR3</td>
<td>0.17 ± 0.02</td>
<td>0.22 ± 0.06</td>
<td>0.40 ± 0.06</td>
<td>Apr 2009–Nov 2011</td>
</tr>
</tbody>
</table>

Figure 6. Example radargram from ground-penetrating radar, with the reflection horizon at 35–45 m below the surface clearly visible, partly indicated with a red line. The lowermost reflections are from the ice–ocean interface at about 350 m depth.

ties of the firm, originating, for example, from changes in the firm properties, such as density, fabric, grain size, and chemical constituents. Based on the distinctiveness and spatial continuity of this layer, and lack of east–west depth gradient that would suggest progressive burial with seaward ice advection, we interpret it as having formed synchronously over the ice shelf, within a single melt season. The strong reflection is at the same depth and of similar signature in radargrams at cross-over points. Unfortunately, the reflection horizon is undated, which precludes the conversion from accumulated mass to SMB.

Total accumulated mass is shown in Fig. 7 for all available radar lines. The lowest values (26–28 m w.e.) are found in the northern part of the LCIS along the MIDAS 2014 radar lines. The highest values (40–45 m w.e.) are concentrated near the southern end of the McGrath radar survey, and to the southwestern in the SOLIS 2009 survey. The multi-decadal SMB estimates from the radar surveys confirm the observations made by the sonic height rangers, in that the SMB is lower in the north than in the south.

3.3 Airborne radar and spatial wintertime snowfall

Airborne radar is another, independent method capable of mapping spatial variability in winter SMB across the ice shelf. The OIB Ku-band shows a spatially persistent, single reflection horizon approximately 0.70 to 1.40 m below the surface, across the entire ice shelf (Fig. 8). The presence of one single strong reflector is anomalous compared to data collected over other parts of Antarctica, where multiple reflectors provide information on the vertical layering in the top few metres of the firm (e.g. Medley et al., 2015). Unlike low-frequency GPR, the OIB radar data over the LCIS only show
a single strong subsurface horizon. This is likely related to differing scattering characteristics of the firm including melt features at the much higher OIB radar frequency.

To test our earlier assumption that this horizon represents the top of the melt layer formed during the previous melt season (Sect. 2.5), we compared the OIB reflector depth with the four sonic height rangers for which data are available in 2009. Specifically, we extracted the thickness of the snow layer that had accumulated since the last significant melt event of the melt season preceding the OIB flights. The melt-season termination dates were established using QuikSCAT and ASCAT microwave data (Trusel et al., 2013). Uncertainty in the sonic height ranger data is computed based on instrument measurement error and uncertainty in the melt termination date derived from the microwave satellite data. In most cases, this amounts to a few centimetres, depending on the snow height variability around the melt season termination date. At AWS15 in 2009, however, the approximated end-of-melt date coincides with a major snowfall event, raising the uncertainty in the sonic height ranger data to 0.17 m. Since we assume that the single OIB reflector represents the melt horizon of the last melt season, a snow density of 360 kg m$^{-3}$, representative of the winter accumulation and derived from in situ snow pit observations (Sect. 2.4), is used to convert the two-way travel time to the subsurface reflection to depth. Again, we use Eq. (4) after Kovacs et al. (1995) to relate firm density to its dielectric constant $\epsilon$, which determines the effective wave velocity through the firm. We calculate the depth uncertainty by combining the dominating uncertainty due to the picking scheme ($\pm 2$ range bins) with that of the density assumption ($\pm 20$ kg m$^{-3}$). Crossover analysis of the OIB radar-derived depths showed close agreement with an RMSE just under 2 range bins, which provides the basis for our uncertainty estimate due to layer picking. The comparison between sonic height ranger and OIB data, compiled in Table 3, shows that, at three locations, there is agreement within 0.02 m between the observed snow accumulation since the previous melt season and the depth of the OIB-observed reflection horizon. There is a discrepancy between OIB-estimated and sonic height ranger depth at LAR1 that exceeds the uncertainty estimates. The LAR1 sonic height ranger appears out of pattern, but we could not establish the cause for this discrepancy.

In Fig. 9, we compare OIB-derived accumulation for the period February–October 2009 with RACMO2-simulated SMB for the same period (assuming again a density of 360 kg m$^{-3}$). At a total of 415 equally spaced locations along the OIB flight lines, the mean bias is 0.10 m (RACMO2: 0.97 m, and OIB: 1.07 m). RACMO2 seems to underestimate at low-accumulation locations and overestimate at high-accumulation locations, leading to a slope that is $>1$ but not significantly so. Nonetheless, RACMO2 seems to capture the magnitude and the spatial pattern of accumulation for this winter season.

Having established the OIB reflection horizon over the LCIS to approximate snowfall since the last snowmelt event of the previous melt season, we are able to map the spatial variability of snow accumulation for the austral winter of 2009 at locations where OIB data are available (Fig. 8). We see diminished winter accumulation in the northern part of
the ice shelf, with smallest values in the north-western inlets of the LCIS. Higher accumulation is found in the southern part, with particularly high values around LAR1 in the south-western part of the ice shelf. In the far south of the ice shelf (south of the Kenyon Peninsula), we again see much lower amounts of accumulation.

### 3.4 A map of SMB and its origin

At all timescales examined in this study – from seasonal to multidecadal – we find greater SMB values in the middle and southern sectors of the LCIS, and lower SMB values in the north. In order to expand our coverage to unsurveyed areas of the LCIS, we combined the regional climate model RACMO2 (Sect. 2.6) and observations from GPR and the sonic height rangers. The average SMB from RACMO2 over the period 1979–2014 guided the extrapolation of the GPR data. Subsequently, RACMO2 SMB was adjusted to match the observed annual SMB from the sonic height rangers. This was done as follows: the 1979–2014 average SMB from RACMO2 was normalized with respect to its spatial mean and so were the GPR data. Next, we determined a linear regression of the normalized RACMO2 SMB values to the normalized GPR data. We used this regression to adjust the RACMO2 SMB to maximize its match to the GPR data while conserving the spatial mean SMB. The result, shown in Fig. 10c, is a RACMO2-guided extrapolation of the GPR over the unsurveyed portions of the LCIS, in which the spatial pattern of RACMO2 SMB is adjusted to the spatial pattern of the GPR observations.

The next step was to adjust the absolute values of RACMO2 SMB to available sonic height ranger observations. We converted RACMO2 SMB back from normalized to absolute values, again using the spatial mean SMB. We determined a weighted mean bias between RACMO2 SMB and all available sonic height ranger observations, selecting the periods for which both were available. We used the length of
the height ranger observation period as a weight for the averaging, reflecting how short-term variability plays a smaller role in longer time series. Compared to the sonic height rangers, RACMO2 underestimated SMB by 14 ± 10 %. Applying a bias adjustment leads to the gridded SMB shown in Fig. 10a. An estimate of SMB uncertainty was based on (1) the fit between normalized GPR and RACMO2 SMB, and (2) the 10 % uncertainty of the RACMO2 bias. The resulting uncertainty is typically 15 % of the SMB value, shown in Fig. 10b.

The underestimation of RACMO2 snowfall over the LCIS was noted by Kuipers Munneke et al. (2014a) and may be the result of the representation of snow formation in clouds or with underestimated evaporation in the Weddell Sea, the most important source region for moisture precipitated over the LCIS. The underestimation of RACMO2 snowfall is also apparent in the comparison with Operation IceBridge radar data and amounts to −13± 10 % (Fig. 9), reinforcing the robustness of our bias estimate.

The SMB pattern in Fig. 10a provides a broader context to the various data sets presented above. In the area of GPR observations, the RACMO2-guided interpolation suggests that the SMB gradient is not strictly north–south but tilted in the north-east to south-west direction, with the lowest values near Bawden Ice Rise in the north-east and highest values of SMB in the inlets in the west to south-west part of the ice shelf.

We used RACMO2 to study the origin of this spatial distribution of SMB. In the absence of notable run-off, SMB is dictated by snowfall, and by sublimation. Figure 11a illustrates the spatial coherence of snowfall across the LCIS, by showing the fraction of snowfall occurring simultaneously with snowfall events exceeding 5 mm w.e. day⁻¹ in the southern area of the LCIS (results are relatively insensitive to the exact location on the LCIS). The pattern shows that snowfall is coherent across the shelf and that only a small fraction of snowfall on the western side of the Peninsula occurs when there is snowfall on its eastern side. Figure 11b shows the mean circulation pattern during these snowfall events (wind speed and direction at the 850 hPa pressure level, along with temperature at 850 hPa in the background). Snowfall on the LCIS is thus strongly associated with low pressure centered near its northern end over the Weddell Sea. These low-pressure systems source water vapour from the Weddell Sea and from more northerly regions to produce snowfall on the LCIS. This circulation pattern can explain the relatively high snowfall rates in the south and south-west, where snowfall is orographically enhanced. As a peculiar smaller-scale expression of this orographic effect, the south-eastern side of the Kenyon Peninsula receives more snowfall than its north-western side, situated in the lee of the flow associated with snowfall. The same pattern can be seen around the promontories clockwise around the ice shelf (Cole and Churchill peninsulas, Veier Head, and Argo Point), where reduction of snowfall is seen at the obstacle’s lee side. The snowfall minimum in the north can be explained by the fact that the low-pressure centre will fluctuate around the location illustrated in Fig. 11b. If the low-pressure system is located farther to the south, the north-eastern part of the LCIS will experience an offshore wind, in the lee of the Jason Peninsula and its promontories, extending to the south (Churchill Peninsula, Veier Head, and Argo Point). Such a southerly position of the low-pressure system would only exert a notable influence on snowfall in the north-east sector of the ice shelf. Thus, snowfall in the north-east of the LCIS is more restricted during local offshore wind than in other places.

Sublimation exerts a secondary control over SMB. Over the LCIS, föhn winds are frequent, and the combination of high wind speed and dry air increases the sublimation rate. During föhn, a pattern of alternating higher and lower wind speed emerges in the western part of the LCIS, where low-elevation inlets are separated by higher-elevation promontories that protrude from the Antarctic Peninsula mountains (Luckman et al., 2014; Elvidge et al., 2014). An estimate of annual mean sublimation rate from RACMO2 is shown in Fig. 12. While sublimation rates from RACMO2 are poorly evaluated over the LCIS, an estimate of sublimation from in situ AWS observations reveals that it amounts to ~25–30 mm w.e. year⁻¹ at AWS 14 (see map in Fig. 1) and between 17 and 64 mm w.e. year⁻¹ at the site of a newly installed AWS in Cabinet Inlet. Comparing the sublimation flux (Fig. 12) with the total SMB (Fig. 10) shows that sub-
limation spatially modulates SMB in the western part of the LCIS, likely by föhn. According to RACMO2, annual mean sublimation typically removes 5–15% of the annual snowfall over the LCIS. This fraction could be larger if RACMO2 underestimates the sublimation flux. It is conceivable that the SMB in certain inlets (Cabinet, Mill, Whirlwind, Mobiloil) has decreased in recent decades following intensification of föhn (Cape et al., 2015) due to enhanced sublimation.

4 Conclusions

We have combined several geophysical techniques along with a regional climate model to constrain spatial and temporal patterns of SMB over the Larsen C ice shelf. Results have been integrated to show that SMB is larger towards the south of the ice shelf, overwritten by an increase in SMB toward the west. Assuming that run-off is negligibly small over the LCIS (van Wessem et al., 2016), the spatial pattern of SMB is dominated by spatial differences in snowfall and sublimation. Thus, our results indicate that snowfall is larger in the south than in the north of the LCIS.

Previous studies have indicated a strong gradient in firn air content from west to east across the LCIS (Holland et al., 2011), with the lowest values in the west. It has been suggested that this reflects enhanced melt and subsequent refreezing, directly at the foot of the Antarctic Peninsula mountains in the western part of the LCIS (Trusel et al., 2013) caused by föhn winds descending from the mountains (Luck-

Figure 10. (a) Map of the Larsen C ice shelf showing a reconstruction of annual SMB (in mm w.e. year$^{-1}$), obtained by adjusting 1979–2014 RACMO2 SMB to the spatial pattern of ground-penetrating radar and observations of SMB from sonic height rangers. (b) Estimated uncertainty of the SMB values in panel (a) (in mm w.e. year$^{-1}$). (c) Coloured dots indicate depth of the reflection horizon, detected by GPR, normalized with respect to the mean depth (unitless). The background shows a map of relative SMB (unitless) from RACMO2.
man et al., 2014). Using observations of SMB, this study shows an east-to-west SMB gradient that, in the absence of melt, would lead to the highest, rather than the lowest, values of firn air in the west. We therefore conclude that the gradient in firn air content is caused by melt and refreezing, rather than by spatial patterns of snowfall or SMB.

We interpret a strong, shallow reflection horizon in the OIB radar data as the top of the melt layer formed during the last melt season. The presence of sufficiently thick melt layers, like on the LCIS, precludes airborne observations of multi-year firn stratigraphy like in the dry snow areas of Antarctica and Greenland. Still, we demonstrate that OIB radar data can be used to track the shallowest melt layer and derive winter SMB immediately prior to the radar survey. This opens up the possibility of acquiring winter SMB estimates over other Antarctic ice shelves and over parts of the Greenland Ice Sheet that experience small to moderate amounts of melt, as is typical for percolation zones.
Much recent work has focused on the stability of the LCIS in a warming climate, with hydrofracturing suggested as a potential mechanism for ice-shelf collapse (Scambos et al., 2000; Kuipers Munneke et al., 2014b; DeConto and Pollard, 2016). Using models to test hypotheses that link atmospheric change to ice-shelf stability is challenging, given the complexity of terrain and climate in this region. A substantial portion of total melt is governed by the occurrence of small-scale föhn winds, which are captured reliably only in models with kilometre-scale horizontal resolution (Elvidge et al., 2014). Summertime melt is also observed frequently on other days (Kuipers Munneke et al., 2012), and its representation in models depends on subtle changes in albedo (Kuipers Munneke et al., 2011) and on the correct simulation of all components of the radiation balance at the surface (King et al., 2015).

Precipitation depends on the ability of existing weather fronts to cross the Antarctic Peninsula mountain range but also on local lee cyclogenesis and low-pressure systems crossing the Weddell Sea. The moisture content of the latter depends on both sea ice extent and the presence of polynyas. Thus, the LCIS is an important and challenging test bed for regional, global, and Earth system models. The performance of these models in predicting ice-shelf collapse is subject to a correct representation of observed climate. Estimates of SMB presented in this study can guide model evaluation and development with the aim of improving our capacity to predict the stability of the LCIS.

**Data availability.** The gridded SMB data are available from Kuipers Munneke et al. (2017) (https://doi.org/10.15784/601056). Underlying GPR and part of the sonic height ranger data are available from January 2018 at the NERC Polar Data Centre hosted at https://www.bas.ac.uk/data/uk-pdc. Operation IceBridge data sets are available at ftp://data.cresis.ku.edu/data/kuband/2009_Antarctica_DC8/CSARP_qlook/.

**Author contributions.** PKM, DM and BM conceived this study, and performed the analysis and synthesis of the data sets. PKM led the writing of the manuscript. DM, BM, SB, BK, DJ, AB, PS, DA, AS, and NG processed and provided observational data sets. AL, SB, BK, AB, BH, DA, and HS collected data in two MIDAS fieldwork campaigns. All authors contributed to discussions on writing this manuscript.

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

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