Dear Dr. Mölg,

The two final corrections have been accepted, i.e. (1) the removal of the phrase “the first time that” in the abstract, per your suggestion, and (2) the addition of acknowledgments, per the reviewer’s suggestion.

Thank you very much for your help and guidance throughout this process.

Sincerely,

R. Tri Datta

---

Melting over the **Northeast Antarctic Peninsula (1999-2009):**
evaluation of a high-resolution regional climate model

Rajashree T. Datta¹, Marco Tedesco², Cecile Agosta³, Xavier Fettweis³, Peter Kuipers Munneke⁴, Michel R. van den Broeke.⁴

¹The Graduate Center, City University of New York, NY 10016, USA
²Lamont-Doherty Earth Observatory of Columbia University, Palisades, New York, NY 10964, USA
³Department of Geography, Université de Liège, Liège, Belgium
⁴Institute for Marine and Atmospheric Research, Utrecht University, Utrecht, The Netherlands

Correspondence to: Rajashree Tri Datta (Tri.Datta@gmail.com)

Abstract. Surface melting over the Antarctic Peninsula (AP) may impact the stability of ice shelves and therefore the rate at which grounded ice is discharged into the ocean. Energy and mass balance models are needed to understand how climatic change and atmospheric circulation variability drive current and future melting. In this study, we evaluate the regional climate model **MAR** over the AP at a 10 km spatial resolution between 1999 and 2009, a period when active microwave data from the QuikSCAT mission is available. This model has been validated extensively over Greenland, has been applied here to the AP at a high resolution and for a relatively long time period (full outputs are available to 2014). We find that melting in the northeastern AP, the focus area of this study, can be initiated both by sporadic westerly föhn flow over the AP mountains, and by northerly winds advecting warm air from lower latitudes.

A comparison of **MAR** with satellite and automatic weather station (AWS) data reveals that satellite estimates show greater melt frequency, a larger melt extent, and a quicker expansion to peak melt extent than MAR in the center and east of the Larsen C ice shelf. These differences are reduced in the north and west of the ice shelf, where the comparison with satellite data suggests that MAR is accurately capturing melt produced by warm westerly winds.

---

Deleted: East

Deleted: plays a crucial role for

Deleted: dynamics of

Deleted:

Deleted: ; hence modulating the mass balance in a region of the world which is particularly sensitive to increasing surface temperatures. Understanding the processes that drive melting using surface energy and mass balance models is fundamental to improving estimates of current and future surface melting and associated sea level rise through ice-shelf collapse. This is even more important in view of the specific challenges presented by how circulation patterns over the topographically-complex Antarctic Peninsula, especially föhn winds, impact surface melt.

Deleted: Modèle Atmosphérique Régionale (MAR)

Deleted: Antarctic Peninsula (AP)

Deleted: horizontal

Deleted: which coincides with the availability of

Deleted: is the first time that this

Deleted: , which

Deleted: been

Deleted: Antarctic Peninsula

Deleted: Our primary regional focus is the

Deleted: em East

Deleted: Antarctic Peninsula

Deleted: (East AP)

Deleted: where we define smaller sub-regions according to divergent melt occurrence biases. Melting in the East AP

Deleted: To assess MAR’s ability to simulate these physical processes, this study takes a unique approach, examining model biases for melt occurrence on the Larsen Ice Shelf, as evaluated by satellite estimates from passive and active microwave data, with concurrent temperature biases associated with wind direction biases as evaluated by three automatic weather stations (AWS). Our results indicate

Deleted: s
INTRODUCTION

Increased meltwater production over the Antarctic Peninsula (AP) in the latter half of the 20th century has been linked to a warming atmosphere, with potential implications for future sea-level rise (Barrand et al., 2013; Turner et al., 2005; Vaughan, 2006). Surface melting has been implicated in the weakening and eventual collapse of ice shelves as well as the subsequent acceleration of contributing glaciers, with the Larsen A (1995) and Larsen B (2002) on the eastern AP as the most notable examples (Vaughan et al., 1996; Rott et al., 1998; Scambos, 2004). In July 2017, a rift on the Larsen C Ice Shelf, which had been expanding for several years, resulted in the calving of the 5800 km² iceberg A68 (Hogg and Godmundsson, 2017).

Surface melting influences ice shelf stability through the stress produced by meltwater ponding as well as meltwater percolation through firm. One proposed mechanism for the disintegration of ice shelves hypothesizes that surface meltwater infills and deepens pre-existing crevasses, through a process called hydrofracture (Scambos et al., 2000; Weertman, 1973; van der Veen et al., 1997). In addition, a complementary mechanism proposes that when supraglacial lakes drain (becoming dolines), an upward flexure is induced which can weaken an ice shelf, both at the surface and at the base (MacAyeal and Sergienko, 2013). Large open-rift systems were observed over the Larsen B ice shelf in the summer of 2002 which are consistent with substantial melt initiating both mechanisms and leading to ice shelf disintegration (Glasser et al., 2008; MacAyeal and Sergienko, 2013). Alternatively, meltwater can affect ice shelf dynamics by percolating into firm and increasing its density, resulting in the calving of the 5800 km² iceberg A68 in 2017.

Meltwater forming over blue ice and flowing downstream to collect in subsurface layers (the ice-albedo feedback) has recently been shown to be substantial in parts of East Antarctica (Lenaerts et al., 2016). Recent work has also shown the lateral flow of meltwater (supraglacial runoff) on the Larsen A Ice Shelf in 1979 (Kingslake et al., 2017), which imply prolonged periods of lowered albedo. These surface rivers could become much more prevalent across Antarctica in future warming scenarios than previously expected, and may provide a means of stabilizing ice shelves by routing meltwater away (Bell et al., 2017).
Since the collapse of Larsen A and Larsen B ice shelves, ice velocities of several of their feeding glaciers have increased, and seasonal variations in flow have suggested that both summer meltwater percolation (Zwally, 2002) and the removal of backstress played a role in the acceleration (Scambos, 2004; Rott et al., 2002). The remaining Larsen C ice shelf to the south could prove to be similarly vulnerable to collapse due to atmospheric warming (Morris and Vaughan, 2013). Radar analysis over a 15-year period has shown that the surface of Larsen C has been lowering from both firm air depletion (due to either limited accumulation or high surface melt) and basal ice loss, although the latter term is thought to be more substantial (Holland, 2015). While most regional climate models (RCMs) do not account for englacial flow or surface rivers, accurate modelling of surface meltwater production is a crucial step in assessing the potential effects on the ice sheet, especially in the case of the Larsen C ice shelf.

The eastern AP, where the Larsen C ice shelf is located, is on average 3-5°C cooler than the western AP at the same latitude (Morris and Vaughan, 2013). When strong westerly winds force air across the bisecting mountain range of the AP (Fig. 1), the resulting föhn winds can produce pulses of warming on the eastern AP ice shelves (Marshall, 2007). Föhn is a warm, dry air flow on the lee slopes of a mountain range (Beran 1967) that can contribute to melt and sublimation (Figure 1). Multiple studies have focused on the use of high-resolution non-hydrostatic models over the eastern AP to determine the frequency of föhn occurrence over relatively short periods (Elvidge et al., 2015; Grosvenor et al., 2014; Elvidge et al., 2016; King et al., 2017). King et al. (2017) found that over a single season, föhn flow occurred 20% of the time. This study showed substantial melt occurrence observed by satellites without föhn flow, suggesting that surface melt was influenced by other factors as well. A recent study by Sutton et al. (2017), using a non-hydrostatic model, compared modelled flow characteristics during two föhn events and found that a 1.5 km version of the model was able to capture the eastward propagation of melt-inducing winds, whereas a 5km version could not, according to a comparison with AWS stations. However, Bozkurt et al. (2018) demonstrate that a 2km version of the same model was still unable to resolve high temperatures associated with the initiation of föhn flow during a short period. We note that because these modelling studies use a non-hydrostatic model, they are limited to short periods due to the prohibitive computational cost.

Models are limited by the parameterization of physics and our incomplete understanding of the physical processes driving the observed changes. Regional climate models (RCMs) such as the Modèle Atmosphérique Régionale (MAR), evaluated here, can be used for simulating the coupled atmosphere/surface system at a continental and decadal scale (Gallée and Schayes, 1994). The trade-off, in this case, is that RCMs might not be able to capture physical processes with the required accuracy and must be thoroughly evaluated with in-situ and remotely sensed observations. Several studies have used passive microwave estimates for melt occurrence alongside in-situ temperature data (Li et al., 2003; Ridley, 1993; Tedesco et al., 2007; Tedesco and Monaghan, 2009), reporting an increase of surface melting over the AP over the 1980-1999 period (Trompse et al., 2003). However, other studies have suggested that the findings may have been impacted by a change in the acquisition hours of the satellite and that changes in melt over the 1979-2010 period were insignificant (Kuipers Munneke et al., 2012).

Melt occurrence over the AP has also been investigated using the QuikSCAT satellite product at a ~2.25km resolution (Long and Hicks, 2010) in combination with model outputs from the RCM RACMO2 (Regional Atmospheric Climate Model) and in-situ temperature estimates (Barrand et al., 2013). Raw backscatter values from QuikSCAT have also

---

Moved insertion [1]

**Deleted:** Observation-based studies on the formation of melt ponds in the Cabinet Inlet portion of the Larsen C Ice Shelf have focused on the response to föhn winds (Luckman et al., 2014) and the formation of subsurface ice (Hubbard et al., 2016). These last studies taken together discuss both the atmospheric drivers for melt as well as the effects on the ice shelf within our region of interest, but are necessarily limited to a small region where observations are available. By contrast, spaceborne satellites allow us to estimate surface melt occurrence and meltwater production over the entire AP, complementing in-situ data. The combination of satellite-based and in-situ data provide an excellent toolset for model validation.

**Moved insertion [3]:** Liu et al., 2003; Ridley, 1993; Tedesco et al., 2007; Tedesco et al., 2009; Tedesco and Monaghan, 2009
been used to estimate melt flux over the AP (Trusel et al., 2013; Trusel et al., 2012). A recent study using 5.5km horizontal resolution run of RACMO 2.3 over the AP suggested that a further increase in resolution would be required to properly resolve föhn wind propagation, which would imply the removal of the hydrostatic assumption (Van Wessem et al., 2015a; Van Wessem et al., 2015b). However, Wiesenecker et al. (2018) show that föhn events observed by an AWS close to the AP mountain range were well captured by a later version of the same RCM, enabling a reconstruction back to 1979. Where the hydrostatic assumption is preserved (such as with MAR), higher resolutions may inhibit flow in the model, resulting in limited eastward föhn flow in the eastern AP (Hubert Gallée, personal communication). Despite these drawbacks, the current class of hydrostatic RCMs which include relatively complete representations of the snow physics are useful tools to simulate the effect of surface melt on the snowpack over long timescales. Additionally, these high-resolution runs can easily be compared to, and potentially nested into, continental-scale runs of the same model.

Here, we assess the MAR model at a 10 km horizontal spatial resolution over the AP, where outputs are available over a relatively long time period (1999-2014, i.e. 15 years), using both satellite and in-situ data, aggregating meltwater production to drainage systems (basins) as described by Zwally (2002). While previous studies have evaluated how surface melt is modelled using satellite data, or evaluated the representation of the near-surface atmosphere with automatic weather station (AWS) data, we use both sources in conjunction to understand MAR’s ability to simulate specific physical processes, i.e. to assess melt and temperature biases by wind direction. We first report total meltwater production from MAR at the basin scale and compare mean annual meltwater production with outputs from RACMO2.3p2 (Van Wessem et al. 2018), another hydrostatic RCM run at a 5.5km resolution (Sect. 3.1). We evaluate surface melt occurrence from MAR at the sub-basin scale using satellite estimates and link melt occurrence biases to temperature and wind biases at a point scale using AWS data. We compare meltwater occurrence derived from two satellite sources, passive microwave (PMW) and QuikSCAT active microwave, with MAR outputs over the AP (Section 3.2). We focus primarily on the NE basin in the East AP as it contains the former Larsen A, Larsen B and current Larsen C Ice Shelf, where we define sub-regions based on high and low melt occurrence estimated by PMW algorithms (Tedesco, 2009). We then compare climatologies of melt extent, as well as inter-annual trends, from both passive and active microwave data with those computed from MAR outputs (Section 3.3). Because melt on the Larsen C Ice Shelf can potentially be initiated by northwesterly föhn flow sourced from over the AP or southwesterly flow through gaps in the mountain range (even at sub-zero temperatures), we compare melt occurrence reported by satellite estimates vs MAR (coinciding with the 2000-2009 QuikSCAT period) partitioned by temperature differences and wind direction at the location of the Larsen Ice Shelf AWS. Two additional stations (AWS14 and AWS15) are used to examine the persistence and spatial distribution of wind biases from 2009 to 2014. (Section 4). Because all three stations are located on the eastern side of the Larsen C Ice Shelf, this comparison can assess the impact of limited eastward flow on temperature and melt occurrence. In light of the model biases found in this analysis and the potential to correct them with an enhanced resolution model in the future, the discussion (Section 5) includes a sensitivity test with MAR at multiple resolutions. This is performed to specifically assess the effects of increased resolution on eastward flow and resultant surface melt. Table 1 lists abbreviations used throughout the text along with sections in which the terms are introduced.
2. Data and Methods

This study takes a combined observational and modelling approach. The primary tool used to understand the coupled atmosphere and snowpack is the MAR RCM. We employ in-situ data collected from 3 AWS stations to evaluate the near-surface atmosphere biases in MAR as well as to assess inter-annual trends. While in-situ observations of 2m air temperature are frequently treated as a proxy for melt (Braithwaite 1981), this method is most effective when the energy budget is dominated by the turbulent sensible heat flux and incoming longwave radiation and does not capture melt which can occur due to shortwave radiative forcing when air temperatures are below 0°C (Hock, 2005; Kuipers Munneke et al., 2012). We also use observations from the QuikSCAT (QS) and SSM/I (Special Sensor Microwave Imager, 1987 - to date) satellites to evaluate both melt occurrence and intensity in MAR.

2.1 Regional climate model outputs

The MAR RCM is a modular atmospheric model coupled to the Soil Ice Snow Vegetation Atmosphere Transfer scheme (SISVAT) surface model (De Ridder and Gallée, 1998), which includes the multi-layer snow model Crocus (Brun et al., 1999). MAR was originally implemented to simulate energy and mass balance processes over Terra Nova Bay, Antarctica (Gallée and Schayes, 1994). Within SISVAT, meltwater is calculated at the surface when the surface reaches the melting point in combination with a surplus of energy (a deficit results in refreezing). The presence of meltwater alters the snow characteristics (for example, the type and size of snowgrains) and percolation through the snowpack is determined through a tipping bucket method based on snow density. A diagram and description of the sequence of these specific processes in MAR is provided in Figure S1.

The model configuration primarily used in this study is MAR version 3.5.2, with 23 sigma layers from 200 hPa to the surface. This version has been used in multiple studies over Greenland; the specific updates to the physics from the original version of MAR as well as multiple uses of this model are described in detail in Fettweis et al. (2016). The fresh snow density scheme used here is a new MAR implementation specific to Antarctica which has been tested from the original version of MAR as well as multiple uses of this model are described in that study. Here, fresh snow density (ρ) is computed as a function of 10m wind speed (WS, m s⁻¹) and surface temperature (Ts, K) such that:

\[ \rho = 149.2 + 6.84 \times WS + 0.48 \times Ts \]  

(1)

This parameterization was tuned such that the density of the first 50cm of snow fits observations collected over the Antarctic ice sheet, although we note that no reliable measurements were available over the AP. The subsequent compaction of snow layers uses the formulation from Brun (1989). There are 30 snow/ice layers of variable thickness from the surface to a 20 m depth (below which ice is assumed present). Topography is interpolated from 1 km Bedmap2 (Fretwell et al., 2013; Green et al., 2016) to the MAR grid. The snowpack is initialized at 300 kg m⁻³ at the surface and 600 kg m⁻³ at depth. Following 2 years of spinup, MAR results are independent of the initial conditions; for these results, 5 years of spinup were run.

MAR outputs are generated at a horizontal spatial resolution of 10 km for the years between 1999 and 2014. The model domain includes the AP region between ~79.5° and ~56.9° latitude and ~94.9° and ~39.7° longitude. Lateral boundary conditions are specified from the European Centre for Medium-Range Weather Forecasts (ECMWF), using
the ERA-Interim reanalysis (Dee et al., 2011), which is also used for a direct comparison with AWS wind speed/direction. This is a single model domain with no nesting. We note that the ice (vs sea) mask used does not include the Larsen A or Larsen B Ice Shelf in order to preserve consistency for comparison between years (most of which post-date the collapse of these ice shelves). For the analysis of the effects of resolution on surface melt estimates presented in Section 5, we use three version of MAR v. 3.9. Relative to version 3.5.2, which is primarily used in this study as well as in Fettweis et al. (2017), the computational efficiency of MAR v3.9 has been improved such that increased resolution runs are potentially viable. The improvements in the physics include an increase in the lifetime of clouds, partly correcting for the underestimation of downward longwave radiation and the overestimation of inland precipitation found in Fettweis et al. (2017). MAR v3.9 setups include a version at a 10km horizontal resolution similar to the model used for the main analysis, one where the horizontal resolution is reduced to 5km and one where the vertical discretization is increased to 32 sigma layers (at a 10km resolution).

We consider two conditions for identifying melting based on previous work comparing MAR outputs (version 3.2) and satellite microwave melt estimates that found that passive microwave estimates were sensitive to a meltwater content of 0.4% (or mm w.e.) in the first meter of the snowpack (Tedesco et al., 2007). The first condition \( \text{LWC}_{\text{crit}} \) determines melt occurrence in MAR when the daily-averaged integrated liquid water content (LWC) in the first meter of the snowpack exceeds 0.4% for at least three consecutive days. The second condition \( \text{MF}_{\text{crit}} \) determines melting when total meltwater production over the day exceeds 0.4 mm w.e., and is intended to capture both sporadic melt (which may refreeze) and melt which has percolated into the snowpack column below 1m, i.e. equivalent satellite-based estimates could have potentially shown melt occurrence during some portion of the day. A sensitivity test was conducted with multiple thresholds, finding that the differences between a threshold of 0.1 and 1 mm w.e. (suggested by Franco et al., 2013 as a melt threshold for Greenland) was negligible overall, but more substantial at the northern Larsen C Ice Shelf, where the 4 mm w.e. threshold proved insufficient to capture melt occurrence (Fig. S2). Similarly, we performed a comparison of melt occurrence computed from 2000-2009 at the Larsen Ice Shelf AWS for all satellite-based algorithms as well as AWS-based melt-occurrence criteria, i.e. where \( \text{MaxT2m} > 0 ^\circ \text{C} \) and \( \text{AvgT2m} > 0 ^\circ \text{C} \) (Fig. S2b). We found that neither total MAR melt occurrence nor the relative agreement with observed sources varied substantially between thresholds until a threshold of 4 mm w.e. Consequently, we use a meltwater production threshold of 0.4 mm w.e. to define melt occurrence for the remainder of the study due to its sensitivity at the northern Larsen C ice shelf. The differences in sensitivity for each satellite-based criteria for melt occurrences, as well as associated temperature biases, are discussed in detail in Section 3.1.

MAR meltwater production is compared to melt outputs from the RCM RACMO2 3p2, a hydrostatic model which has been run extensively over polar regions and over the AP at a 5.5km resolution at 40 vertical levels. RACMO2 3p2 is forced at the boundaries by ERA-Interim every six hours, as with MAR in this study. (Van Wessem et al., 2018). Model results over the AP for RACMO 2.3p2 did not vary substantially from RACMO2.3, which was evaluated extensively in previous work (Van Wessem et al., 2015a; Van Wessem et al., 2015b).
2.2 Microwave satellite estimates of melt extent, duration

Spaceborne microwave sensors can detect the presence of liquid water in snow over those regions where poor or no observations and unlike sensors in the visible range, microwave sensors are only weakly affected by the presence of clouds. In the case of active measurements (e.g., radar, scatterometer), the presence of wet snow is associated with a sharp decline in backscatter ($\sigma^0$) (Ashcraft and Long, 2000), whereas in the case of passive microwave data the detection is associated with an increase in brightness temperature ($T_b$) (Mote et al., 1993; Tedesco et al., 2007). In either passive or active microwave estimates, even the presence of a relatively small amount of liquid water (i.e. a few percent) triggers a substantial increase in the imaginary part of the dielectric constant (Ashcraft and Long, 2006; Ulaby and Stiles, 1980).

2.2.1 Active Microwave Data: QuikSCAT

We employ a wet snow high-resolution product (~22.25 km) described in Steiner and Tedesco (2014) to derive melt occurrence from active microwave data. Both melt occurrence and raw backscatter values used in this analysis use normalized backscattering values as measured by the Seawinds sensor onboard the QuikSCAT satellite at Ku band (13.4 GHz), with the enhanced resolution provided by the application of the Scatterometer Image Reconstruction (SIR) algorithm (Long and Hicks, 2010). Both Ku- and C-band scatterometers have been used extensively to detect melt onset and freeze-up in Antarctica and Greenland (Drinkwater and Liu, 2000; Steiner and Tedesco, 2014; Ashcraft and Long, 2006; Kunz and Long, 2006).

Threshold-based approaches with active microwave data, as used in this study, identify the point of melt onset based on the departure in $\sigma^0$ from values in dry-snow with various thresholds (Ashcraft and Long, 2000; Ashcraft and Long, 2006; Trusel et al., 2012). The approach used here derives melt occurrence from a threshold-based method (H3), which identifies melt when backscatter falls 3 dB below the preceding winter mean (Steiner and Tedesco, 2014; Ashcraft and Long, 2006). This method, along with a wavelet approach have been evaluated over the AP with AWS data at 5 locations; melt was assumed to occur at the AWS location when 2 m air temperature exceeded 0°C for more than 6 hours (Steiner and Tedesco, 2014). In addition to the binary detection of melt, several methods have been proposed which relate seasonally-integrated backscatter reduction to measures for melt intensity (Wismann, 2000; Smith, 2003; Trusel et al., 2013). As these methods provide seasonally-cumulative values, we do not employ them in this study, although we do examine raw backscatter values as a proxy for melt flux.

2.2.2 Passive Microwave Data

We complement the assessment of MAR with estimates of melt extent and duration obtained from passive microwave observations which have been used in the past to assess melt occurrence in Antarctica and Greenland using brightness temperature at 19.35 GHz with a horizontal polarization (Tedesco, 2007). One of the major disadvantages of passive microwave is the relatively coarse horizontal spatial resolution (25 km) with respect to the fine-scale topography characterizing the AP. However, the historical record for passive microwave data extends as far back as 1972. Threshold-based methods for melt detection from passive microwave data range from a combination of multiple
frequencies and polarizations (Abdalati and Steffen, 1995) to using a single frequency, single polarization (e.g., Mote et al. 1993, Tedesco 2009), as is used in this study. Three algorithms are used here which are described in detail in (Tedesco, 2009). These include the 240-algorithm where the threshold was determined as the value above which an increase in liquid water content above 1% no longer produces an increase in $T_b$ based on output of an electromagnetic model. The original threshold of 245K was found to be insufficiently sensitive and reduced to 240K for this study (Tedesco, 2007) (M. Tedesco, personal communication). The second algorithm uses the winter mean threshold-based method ALA:

$$T_c = T_{\text{winter}} + \Delta T$$

where snowmelt is assumed to occur when the brightness temperature ($T_b$) exceeds a threshold brightness temperature ($T_c$) based on the mean winter (JJA) $T_b$ (T$_{\text{mean}}$, equal to 273K) and a mixing coefficient ($\alpha$, equal to 0.47). For the ALA algorithm, Ashcraft and Long (2006) presume a wet layer of 4.7 cm and a Liquid Water Content of 1%. Finally, the third algorithm (zwa), determines melt occurrence when $T_b$ exceeds a threshold value $T_c$ which is based on the on the winter mean threshold ($T_{\text{winter}}$) and a threshold value (AT), in this case 30K (Zwally and Fiegles, 1994).

2.3 AWS measurements

We evaluate the MAR simulation of the near-surface atmosphere using pressure, temperature and wind speed data collected by three automatic weather stations (AWS) on the AP (Fig 1). The comparison between MAR outputs and AWS data for surface pressure are provided in supplementary data. Data from the Larsen Ice Shelf (AWS) is obtained from the University of Wisconsin Madison (AMRC, SSEC, UW-Madison) at a 3-hourly temporal resolution. AWS data from two additional sites on the Larsen Ice Shelf (AWS14 and AWS15) are obtained from the Institute for Marine and Atmospheric Research at Utrecht University (IMAU) at an hourly resolution (Kuipers Munneke et al., 2012). We note that the Larsen Ice Shelf (AWS -67.00°S, -61.60°W) and AWS14 (-67.00°S, -61.5°W) fall within the same MAR grid cell.

AWS values are temporally averaged to obtain mean daily values for the comparison with MAR outputs. Metrics are computed for December-January-February (DJF, summer). We did not compute a seasonal average when more than 5 consecutive days of data were missing. The five-day period was chosen as an upper limit for the length of a synoptic event, corresponding spatially to approximately 145 MAR grid cells (or half the model domain) of continuous flow in a single direction for an average windspeed of 3.4 m/s, which is the expected value (i.e. the predicted mean based on the Weibul distribution) for Larsen Ice Shelf AWS in DJF from 1999-2014 (Fig 2c). Near-surface (2m) air temperature values are corrected for a difference between AWS station elevation and the elevation averaged by the corresponding MAR gridcell by calculating the elevation gradient from surrounding MAR gridcells and interpolating the final value for the AWS location’s recorded elevation using the Bedmap2 DEM (Fretwell et al., 2013). Differences in elevation values between MAR at 10km resolution and those recorded at AWS stations were as large as 23 m. Maximum daily 2m air temperature (MaxT2m) is calculated as well because this measure may help capture sporadic melt events. MaxT2m values are extracted from available 3-hourly values and are used only when
no more than one 3-hour measurement is missing during the day. Pressure values from AWS stations are also observed at approximately 2m above the surface, and compared to MAR values at the first atmospheric layer in MAR. Because the height of this layer is generally between 2 and 3 m above the surface, this is treated as an acceptable proxy for 2m pressure values. Pressure values from MAR are corrected for elevation using the hypsometric equation (Wallace and Hobbs, 1977).

2.4 Statistical Methods

To evaluate and quantify the differences between MAR outputs and AWS data for temperature and wind speed we use a mean bias. Additional statistical measures shown in supplemental data include the coefficient of determination (R^2), root mean squared error (RMSE) and mean error (ME) (Wilks, 1995). We assess the extent to which each station is representative of larger scale climate variability by constructing correlation (R^2) maps between MAR values colocated with AWS stations vs all other gridpoints in the full MAR domain (Fig. S7). We ignore all R^2 statistics where the p-value exceeds 0.05.

To capture wind speed frequency distributions, we fit available data for each season for MAR (for the full 2000-2009 period), AWS (when AWS data are available) and MAR-R (MAR values collected only when AWS data is available) with a Weibull distribution (Wilks, 1995). The shape (β) parameter roughly captures the degree of skew, with higher values being closer to a normal distribution. The scale (λ) parameter approximates the peak frequency (we note that this is not equivalent to the arithmetic mean). We report expected values (i.e. first moment or mean) for each windspeed distribution using the best Weibull fit.

3. Results: Melt Occurrence and Meltwater Production

In this section, we show results concerning total meltwater production in the AP and compare melt occurrence estimated by MAR with estimates from three passive microwave algorithms as well as QuikSCAT ft3. The relative sensitivity of each melt occurrence criteria, as well as their associated temperature biases, are first compared at the location of the Larsen Ice Shelf AWS. We then identify spatial biases for melt occurrence at the domain scale, finding substantial differences in the center of the Larsen C Ice Shelf as well as to the north and west of the NE basin, a region which includes the former Larsens A and B ice shelves as well as the northernmost portions of the Larsen C ice shelf (Section 3.2). These differences could result from either weaknesses in the MAR representation of wind dynamics (discussed in Section 4) or from limitations of the satellite sensor or algorithm. Finally, we compare the climatology and inter-annual variability of melt extent (calculated by multiple algorithms) over the CL and NL region (Section 3.3).

3.1 Meltwater production over the AP

We show MAR meltwater production over the 1999-2009 period (Fig. 2). The total annual meltwater production estimated by MAR shows substantial inter-annual variation with the NE basin accounting for the highest meltwater production, closely followed by the SW basin (in green). The NE basin is divided into three regions: the NL and CL...
masks (discussed in Section 3.2) and the remainder of the basin. We note that the SW basin does not covary with the NE basin and the subregions of the NE basin do not consistently covary with one another. The meltwater production shown here does not account for refreezing and we note that the effects of refrozen melt on the snowpack will vary regionally depending on local properties. The NL region dominates meltwater production in the NE basin in most years except for 1999-2000, 2002-2003 and 2003-2004. The 2001-2002 melt season shows the second lowest overall melt production during the study period (only the preceding year is lower). Declining aggregate meltwater production across the AP does not necessarily correspond to declining meltwater production in the most vulnerable regions of the northeastern AP (including the Larsen C Ice Shelf). Because melt in the NL region is particularly sensitive to föhn-induced melt, we note that changes in circulation patterns may affect the northwest regions differently than the southern regions. The strong relationship between wind direction and temperature bias points to the need for isolating dominant inter-annual patterns of melt in the Northern Larsen C Ice Shelf and associating them with large-scale atmospheric drivers.

A comparison between mean annual meltwater production from 2000-2009 calculated using RACMO2.3p2 (5.5 km) vs MAR (10 km) is shown in Fig. 3. MAR shows higher meltwater production overall (Fig. 3b vs 3a), with a difference of over 150 mm w.e. on the Larsen C Ice shelf north of 67°S latitude. Over the NE basin, MAR meltwater shows enhanced meltwater production near the AP mountains, including towards the southern edges, and declines eastward and southward. By comparison, meltwater production from RACMO2.3p2 melts declines southward, but no similar west-to-east gradient is apparent. Although inter-annual standard deviations over the northern Larsen C Ice shelf are generally above 100 mm w.e. in both models, there are major differences in other regions, with MAR meltwater production exceeding RACMO2.3p2 values by 30 mm w.e. on the southern Larsen C Ice shelf as well as the George VI Ice shelf (Fig. 3d vs 3c). Van Wessem et al. (2015a) suggest that even at 5.5 km resolution, the underestimation of the height and slope of the orographic barrier may result in an underestimation of föhn winds as well as precipitation in RACMO2.3p2. We note that in addition to the difference in horizontal model resolution, RACMO2.3p2 contains 40 atmospheric layers while MAR implements 23 layers. While the differences in total meltwater production from RACMO2.3p2 and MAR could be a product of dissimilar physics, the potential effect of model resolution on meltwater production in MAR is specifically discussed in Section 5. While melt occurrence and meltwater production are not related in any linear fashion, we note that the spatial pattern produced by MAR, i.e. the eastward gradient from the edge of the AP, is also shown in observed melt occurrence estimates, most notably from the PMW zwa and QS algorithms (Fig. 5f,g), as discussed in greater detail in the next section.

3.3 Melt occurrence over the AP

Fig. 4 shows melt occurrence (in days) at the Larsen Ice Shelf AWS location (shown in Fig. 1) as estimated from the satellite-based algorithms QuikSCAT o3 (Section 2.2.1), three passive microwave algorithms (Section 2.2.2), temperature-based criteria from the AWS station (MaxT2m > 0°C and AvgT2m > 0°C), and the MF24 metric derived from MAR (Section 2.1). At this location, we find that QuickSCAT o3 and PMW ZWA show the greatest sensitivity to melt occurrence. Of the AWS-based metrics, M (MaxT2m > 0°C) shows a sensitivity to melt occurrence comparable to PMW ALA while the T metric (AvgT2m > 0°C) compares poorly to satellite-based measures (Fig. 4a). We find
that at colder temperatures (when MAXT2m < 0°C), AvgT2m values reported by MAR are substantially higher than those reported by the AWS when only MAR reports melt (Fig. 4b). However, at higher temperatures (where MaxT2m >= 0°C), the AWS reports higher MaxT2m temperatures than MAR and biases are even stronger when only observation-based metrics report melt (Fig. 4e). We note that the Larsen Ice Shelf AWS is located on the eastern edge of the Larsen C Ice shelf and the major discrepancies in melt occurrence at this location will be explored further in Section 4, where we further expand the analysis of melt occurrence and temperature biases to include wind direction biases as well.

In Fig. 5, we show melt occurrence over the full domain derived from satellite sources, both metrics derived from MAR (Section 2.1) as well as the MF_{lat} criteria applied to RACMO2.3p2. QuikSCAT f3 generally estimates higher average yearly melt occurrence than either of the MAR melt metrics over the full domain. In the NE basin, the difference is on the order of 25 more days than the MAR MF_{lat} melt metric (Fig. 5a). Differences between QuikSCAT f3 and MF_{lat} also show a strong latitudinal dependence in the NE basin, shifting from near agreement in the northern regions of the Larsen C Ice Shelf to QuikSCAT f3 reporting over 500% of the melt days reported by MAR towards the southern edge. Melt onset is on the order of 22 days earlier in QuikSCAT f3 than in MF_{lat} in the NE basin, except at the northern edge of the Larsen C ice shelf, where MF_{lat} reports average yearly melt onset as much as 25 days earlier than QuikSCAT f3 (Fig. S3). A comparison between the two MAR melt metrics shows that MF_{lat} reports as much as 40 more days of melt than LWC_{lat} at the northern tip of the Larsen C Ice Shelf (Fig. 5b vs Fig 5a). The portion of the Larsen C ice shelf which experiences an average of 25 days of melt or more extends as far south as 80.0°S on the eastern side of the Larsen C ice shelf according to the MF_{lat} metric but extends only to 70.5°S according to LWC_{lat}. Towards the very south of the Larsen C Ice Shelf, the two MAR metrics show similar values, although LWC_{lat} reports melt onset as late as early January (Fig. S3b) while MF_{lat} reports melt onset in December (Fig S3b). The formulation for the MF_{lat} metric, which considers melt at any time of the day for the full depth of the snowpack, suggests that the early season melt observed only by MF_{lat} is either sporadic (i.e. can refreeze) and/or percolates below 1 m in the snowpack in the south of the Larsen C Ice Shelf, i.e. below the depth range at which LWC_{lat} is calculated. Whereas QuikSCAT f3 and MAR melt metrics report maximum melt occurrence in the north and west of the Larsen C Ice Shelf (MF_{lat} reporting > 60 days, Fig. 5p), PMW algorithms report maximum melt occurrence in the center-east of the Larsen C Ice Shelf, specifically 43 days (240, Fig. 5p), 57 days (ALA, Fig. 5d) and 69 days (ZWA, Fig. 5e). RACMO2.3p2 reports substantially higher melt occurrence than MAR at the center of the Larsen C ice shelf as well as a comparatively limited west to east gradient. Because overall average annual meltwater production in MAR was shown to be substantially higher, with a stronger west to east gradient away from the AP (Fig. 3), we conclude that in comparison to RACMO2.3p2, MAR produces melt less frequently, but with greater intensity.

In summary, a comparison between observed and modeled data sources show two distinct spatial patterns for maximum melt occurrence. QuikSCAT f3 as well as both MAR melt metrics show the highest range of melt days in the northern and western edges of the Larsen C Ice Shelf (including both high and low elevation regions) while PMW algorithms show the highest number of melt days in the center of the Larsen C Ice Shelf, where elevations are lower and topography is less complex. We hypothesize that the major difference in spatial patterns between algorithms/melt metrics is related to the different resolutions of the data sources (1-2.2225 km for QuikSCAT, 10km for MAR and
25km for PMW), such that QuikSCAT is better able to resolve melt where topography is complex, such as near the spine of the AP. Secondly, the differences are a product of the depths presumed for the calculation of meltwater content. This is true for both the MAR metrics and for the three PMW algorithms; the “ALA” algorithm, for example, presumes a 4.7cm depth and a 1% liquid water content. (see Section 2). To confirm this, we find the maximum depth to which meltwater percolates (according to MAR) associated with the number of days when melt occurs (according to PMW algorithms). Histograms for total PMW melt days in Fig. S4 show three peaks (two major inflection points) for each algorithm which are used to create three classes for meltwater occurrence (“low”, “medium” and “high”). For these classes, the maximum depth to which meltwater percolates (in MAR) is shown in Fig. S6 and the associated elevation and MAR meltwater production is shown in Table S1.

Spatial regions defined as having “low” melt occurrence are highly heterogeneous with regard to elevation, meltwater percolation and the relative sensitivity of PMW algorithms. Low melt occurrence regions largely include the spine of the AP and regions just east of it. Bedmap2 (Fretwell et al., 2013) reports a large range of elevations while MAR reports low coincident meltwater production and a relatively shallow meltwater depth. Both the ALA and ZWA algorithms report melt at higher elevations (above approximately 1300m and 1900m, respectively) than the 240 algorithm, which neither reports any melt occurrence above 1100m in the NE basin nor at lower elevations to the north and south. (Table S1, rows 1,4,7 and Fig. S6). Where melt occurrence is low, the 240 and ALA algorithms generally detect melt only where MAR reports a maximum meltwater percolation depth below 0.4 m, (Fig S6a,b), whereas the ZWA algorithm can detect melt at a substantially shallower depth of 0.1 m (Fig S6c). Although generally meltwater in MAR rarely percolates below 3m, in low melt-occurrence regions, modeled meltwater occasionally percolates below 10m in the beginning of the melt season (Fig. S6a,b,c, column “N”, indicating November). We remind the reader that melt occurrence within the firm layer (as calculated by MAR $MF_{a,c}$) will capture melt that can refreeze immediately, so this does not necessarily correspond to melt which is retained in the snowpack. Rather, the snowpack layer depth represents the deepest layer which is affected by the melt process according to MAR.

By contrast, where PMW reports high melt occurrence in the NE basin, MAR consistently reports high coincident meltwater production, low elevations and the deepest average meltwater percolation in the region. In the month of January, we find that where PMW algorithms report melt, coincident MAR meltwater percolates to 2 m into the snowpack for 35-47% of the total day-pixels in the NE basin which report any melt, and as deep as 3 meters for more than 30% of total day-pixels (Table S1, 240-H, ALA-H, ZWA-H, Fig. S6 a,h,i).

To quantify the two major spatial trends for maximum melt occurrence, i.e. (1) PMW in the center-east of the Larsen C ice shelf and (2) QuikSCAT B3 and MAR in the northwest of the NE basin, we (a) explicitly calculate concurrent melt occurrence in all PMW algorithms (PMWAll) for the first region and (b) define the latter geographically in order to include most of the NE basin, but deliberately exclude center-east of the Larsen C ice shelf region where PMW melt is highest. The first region “CL” (Center Larsen, as the entire region is restricted to the Larsen C ice shelf), where all PMW algorithms agree on high melt occurrence, is defined where PMWAll reports average yearly total melt days exceeding one standard deviation from the mean for the NE basin. Mean elevation for the CL region is 42.70±17.70m (where $\sigma$ is one standard deviation), PMWAll reports a mean annual 36 days of melt occurrence (vs 21 days derived from $MF_{a,c}$) and the mean annual MAR meltwater production calculated only where
PMWAll reports melt occurrence is 96 mm w.e./100km² (vs 143 mm w.e./100km² when MFₚₐ reports melt) (Table S1, row 11,12).

The “NL” (Northern Larsen) mask is defined by finding the mean latitude of the CL region and including all portions of the NE basin above this latitude, but excluding the CL region (Fig. 2, inset). In the NL region, elevation is highly-variable, with a mean value ~600m and MAR and QS detect melt both earlier and more often than for PMW algorithms. The NL region includes the eastern spine of the AP and most inlets (including Cabinet Inlet and SCAR Inlet), a small portion of the northern Larsen C ice shelf and all regions surrounding the former Larsen A and Larsen B ice shelves.

3.3 Climatological and inter-annual trends for melt extent at the sub-basin scale

We compare the seasonal cycle and interannual variability of melt as modeled by MAR vs observations for both the CL and NL regions by computing regional melt extent over the 2000-2009 period (total melt extent area for each day in NDJF), for each year as well as the climatological average. The PMWAll algorithm is typically treated as the most restrictive condition while the PMW zwa and QuikSCAT ft3 are the most sensitive. Melt extent is defined as the total area reporting melt daily between Nov 1st and February 28th (austral summer, including November to show early melt) (Fig. 9).

The melt extent climatology for PMWAll in the CL region shows the initial increase in sustained melt occurring around December 15th with melt extent peaking in January, followed by a series of increasingly smaller melt pulses ending with refreezing at the end of February. While MAR shows peak melt extent at the same point in the season, the progression from melt onset is more gradual, average peak melt extent is generally smaller and interannual variability (indicated by the grey envelope) during peak melt extent is substantial, with PMWAll reporting a larger melt extent than MAR towards the middle of the melt season in most years (Fig. 6d), but not necessarily during melt onset or its ending. In the CL region, PMWAll reports a larger melt extent throughout the melt season during 2000-2001 and 2001-2002 (Fig. 6a). During three periods, MAR reports a larger melt extent than PMWAll, including 1999-2000, the latter half of the 2002-2003 season and the 2003-2004 season. While the highly-sensitive PMW ZWA algorithm (Fig. 6f) reports sporadic periods where MAR melt extent is larger (during the 1999-2000 and 2003-2004 melt seasons, for example), ZWA generally reports either a larger melt extent or general agreement with MAR. Similarly, melt extent derived from the QuikSCAT ft3 algorithm consistently shows a larger melt extent than MAR, except for a few short periods towards the end of the season in 1999-2000 and 2003-2004 (Fig. 6g,h). We note that for several years, both QuikSCAT ft3 and PMW ZWA report substantial melt occurrence early in the season (~Nov 15th) and that the QuikSCAT ft3 climatology frequently reports melt occurrence in the CL region well after February (Fig. 6g). The NL region includes areas which reported low melt occurrence in all PMW algorithms, variable meltwater percolation depth in MAR was variable, and a large range of elevations was observed (Section 3.2), implying that the mask defined by the combined PMWAll algorithm is less clearly linked to consistent modeled physical properties in this region. Here, the MAR melt extent climatology (Fig. 7a,b) is consistently larger than PMWAll throughout the
season (Fig. 7.c,d). In comparison to the ZWA (Fig. 6.g) and QuikSCAT f3 (Fig. 6.h) algorithms, MAR reports less melt extent in the middle of the season (with peak melt extent in January), but larger melt extent at the beginning and end of the melt season. As compared with the CL region, the MAR climatological melt extent shows less inter-annual variability (grey envelope, Fig. 2a). During the 2000-2001 and 2001-2002 melt seasons, MAR shows a larger melt extent than PMWAll (Fig. 3d), but less than the PMW ZWA (Fig. 3f) or QuikSCAT f3 (Fig. 3g) algorithms. We find that during the 2005-2006 season, MAR shows greater melt extent than PMWAll, consistently less than QuikSCAT f3, but reports a greater melt extent than ZWA only towards the end of the season. We consider the condition where only QuikSCAT f3 or PMW ZWA show a greater melt extent than MAR to be potentially indicative of sporadic surface melt.

In summary, we conclude that in the CL region, MAR reports a larger melt extent from 2000-2001 than PMWAll (which is highly-restrictive), but a smaller melt extent than either the PMW ZWA or QuikSCAT f3 algorithms, which are more sensitive. Notably, MAR melt occurrence is comparatively low during the peak melt period. By contrast, in the NL region, MAR reports greater melt occurrence than the most restrictive measure (PMWAll) during peak melt, but far less than the highly-sensitive QuikSCAT f3 algorithm. The interannual comparison suggests that MAR shows substantially less melt occurrence than observations during the 2000-2001 and 2001-2002 seasons in the CL region, but not the NL region.

4. Results: Wind and Temperature Biases at the Larsen Ice Shelf station

The eastern AP is generally substantially colder than the western AP, and temperature-driven melt primarily results from either large-scale advection from lower latitudes or from westerly föhn flow over the spine of the AP (Marshall et al., 2006). Here, we assess the bias in temperature and melt occurrence associated with wind direction at three AWS locations on the Larsen C Ice Shelf (shown in Fig. 1). We first discuss wind direction and wind speed biases during the summer season at all three locations (without regard to melt occurrence) (Section 4.1). For prominent wind direction biases, we quantify the associated temperature and melt occurrence biases in order to capture atmospheric conditions where MAR reports less melt occurrence than observations (Section 4.2). All MAR and satellite data used are co-located to the grid cell associated with the AWS (Fig. 1), and we remind the reader that all three stations, at the eastern edge of the CL region (Fig. 2 inset), are located where MAR reported substantially less melt occurrence than PMW algorithms, QuikSCAT f3 or AWS temperature-based criteria.

4.1 Aggregate wind direction biases

Fig. 8 shows wind frequency distributions during the summer season, color-coded for wind direction as represented by the pie graph at the right. We note that AWS data are 3-hourly averages and ERA-Interim are 6-hourly averages for wind speed and direction, while MAR produces daily-averaged outputs. For this reason, a direct comparison between Weibull parameters derived from MAR vs AWS data is not fully justified. The Larsen Ice Shelf AWS has full temporal coverage during the QuikSCAT period while AWS14 and AWS15 were installed after termination of the QuikSCAT mission. These last two stations are used in this study to demonstrate that (a) similar wind biases
4.2 Wind and temperature biases concurrent with observed melt occurrence

When daily-averaged temperature (AvgT2m) values are high, it is more likely that melt is sustained, while high maximum daily temperatures (MaxT2m) can also occur during sporadic melt. Melt occurrence is strongly influenced by the temperature of the snow column as well as at the surface; internal melting can occur even when the surface is frozen due to net outgoing longwave radiation (Holmgren, 1971; Hock, 2005). It is therefore possible for melt to occur despite a cold bias. In general, we find a small, but consistent warm MAR bias for AvgT2m, and a consistent cold MaxT2m bias (Table 2, rows 12, 13). However, when we restrict the dataset to days when AWS-recorded temperatures exceed 0°C, a condition where melt is most likely, MAR indicates a cold bias for AvgT2m and an enhanced cold bias for MaxT2m (Table 2, rows 15, 16). This implies that MAR is colder than observations at the temperature ranges where melt is likely, although melt is still possible due to other components of the energy balance.

The cold MaxT2m temperature bias is strongest during northerly flow in general (Table 2, row 13, 16, col 2.5), but strongest during easterly flow on the days when MAR reports melt (Table 2, row 23, 26, col 2.3). Satellite-based melt is detected primarily when AWS-recorded flow is northeasterly (0°–90°) or southwesterly (180°–270°), with PMW(QS) reporting 42% (36%) northeasterly flow and 29% (26%) southwesterly flow. On days when MAR reports melt (Table 2, rows 19, 20), southeasterly flow in MAR is more prominent (while AWS values decline) while the proportion of northwesterly flow declines (but increases at the AWS). We find that the major flow biases account for a relatively small proportion of melt which is captured by observations but not by MAR. The easterly flow bias accounts for 8% (9%) of days when PMWAll(QS) melt occurrence is not also captured by MAR (Table S9) while the southerly flow bias accounts for 6% (6%) of days when PMW(QS) melt occurrence is not also reported by MAR (Table
For these wind direction biases, Fig. 9 presents temperature values when observed sources, either PMW All or QuikSCAT f3, report melt, but MAR does not. We refer to the condition where PMWAll reports melt (but MAR does not) as “PMWEx” (i.e. PMW exclusive-or), with the equivalent condition for QuikSCAT f3 called “QSEx”. We limit the melt days shown in each figure panel to a specific wind bias, thus showing how the wind bias directly influences temperature-driven melt in both satellite-based observations as well as MAR. Tables S8-S12 contain relative proportions of each case (flow bias) divided for each restriction (i.e. MAR, QSEx or PMWEx), as well as the timeseries mean and biases for AvgT2m, AvgT2m>0°C (excluding days when AvgT2m values from AWS are below 0°C), MaxT2m and MaxT2m>0°C.

For the main biases, i.e. when MAR either reports northerly winds as southerly (Fig. 9a,b) or westerly winds as easterly (Fig. 9c,d), modelled temperature values are clustered around 0°C, whereas AWS-observed temperatures, especially when only satellite-observed melt occurs, are higher. When MAR reports melt, MAR AvgT2m values cluster near 0°C, with a small overall warm bias (Tables S8,S9, row 4, col 8). Under omission conditions (PMWEx and QSEx), AvgT2m values are lower, and the MAR bias is slightly negative, although the standard deviation is high (Tables S8, S9, row 5,6, col 7). With all flow cases, only QuikSCAT f3 shows melt at very low observed AvgT2m values. By contrast, AWS MaxT2m values are substantially higher than MAR values (the latter clustering around 0°C) (Fig. 9b,d). Temperature biases associated with southwesterly flow are similar to those shown by the overall bias towards easterly flow in MAR, and are shown in Table S10,S11.

Northwesterly winds are most likely to produce föhn-induced melt and we find that on days when MAR reports melt, only 13.2% of winds are northwesterly while AWS reports 25.2% of flow as northwesterly (Table 2, rows 9,10, col 5). Northwesterly winds show the highest expected windspeeds as well as the highest standard deviation for both MAR and AWS (Table 2, rows 19,20, col 5). While the temperature bias when wind directions are in agreement is relatively minimal, the temperature bias when northwesterly winds are misrepresented is substantial. When MAR reports melt but misrepresents northwesterly winds (this condition accounts for 3% of all MAR melt days), the cool bias for MaxT2m > 0°C is above 4°C (Table S12, row 4, col 10). For the PMWEx condition (when PMW reports melt but MAR does not), AWS MaxT2m values exceed MAR values by more than 5°C (Table S12, row 5, col 10). Despite the strength of the temperature bias, this wind direction bias accounts for only 3% of melt in MAR and only 3%(4%) of melt occurrence reported by PMWEx(QSEx). By contrast, when westerly flow is modelled accurately, MAR captures higher AvgT2m values, which frequently exceed 0°C, with a slight cool MAR bias when AvgT2m > 0°C (Fig. 9e). The PMWEx and QSEx conditions still report melt at lower temperature values, and the MAR bias remains positive. Although a cold MAR bias persists, MaxT2m values are generally in better agreement at the Larsen IS AWS location during this condition (Fig 9f, Table S12).

5. Discussion and Conclusions

We conclude that MAR captures melt which occurs just east of the AP (which is normally the product of westerly föhn flow) with acceptable accuracy according to satellite estimates, but that that melt is underestimated with respect to both AWS and satellite estimates in the eastern part of the Larsen C Ice Shelf. This is partially the result of limited westerly flow in MAR towards the eastern part of the Larsen C Ice Shelf, as compared to AWS estimates. Specifically,
MAR shows lower melt occurrence than satellite estimates in the center and east of the Larsen C Ice Shelf (i.e., the CL region, where eastward flow is likely limited in MAR), while in the north and west of the NE basin (i.e., the NL region which is most immediately affected by westerly flow), MAR reports melt occurrence largely concurrent with satellite estimates. The NL region fits a spatial pattern of föhn-induced melt just lee of the AP and extending eastward from inlets which has been shown in previous studies (Grosvenor et al., 2014) and particularly in the northernmost portion of the NE basin surrounding the Larsen B ice shelf, where the correlation between föhn winds and satellite-based melt occurrence has been shown to be as high as 0.5 between 1999-2002 (Cape et al., 2015, Fig. 12). For example, within the CL region, there are periods during the 2001-2002 season when MAR reports no meltwater production, but raw QuikSCAT backscatter values report periods where over 300 km² of surface area show backscatter values dipping below -15 dB (Fig. S9e).

MAR reports warmer temperature compared to AWS observations recorded on the east of the Larsen C ice shelf at temperatures below 0°C, when melt is less likely to occur, but which may still impact the refreezing process. However, when maximum daily temperatures (MaxT2m) and average daily temperatures (AvgT2m) exceed 0°C, MAR shows a substantial cold bias. This is particularly evident when MAR misrepresents westerly winds or northerly winds, and the temperature bias is most extreme when northwesterly flow is misrepresented, i.e., the condition when the most intense föhn flow would be likely. However, this represents only a small proportion of the melt occurrence bias, i.e., melt occurrence reported by satellite estimates, but not by MAR.

We demonstrate the impact of westerly winds on melt during a single season, specifically during both December and the beginning of January of the 2001-2002 season. During both of these periods, satellite-based melt extent in the CL region increases substantially, while MAR melt extent declines after an initial pulse (Fig. S9a). In December, MAR shows an increase in northwesterly flow, both at the station and throughout the region while AWS reports northwesterly winds at slightly higher speeds. Beginning approximately on January 1st, the NL region reports substantial northwesterly flow, followed by southwesterly flow, although neither is reported at the Larsen Ice Shelf AWS station east of the NL region. Over January, while both AWS and MAR report northeasterly flow, the AWS station also reports substantial high-speed southwesterly flow not captured by MAR. After this period (beginning on approximately Jan. 1st), AWS AvgT2m temperatures consistently exceed MAR AvgT2m values until the end of the season (Fig. S11), suggesting that because MAR did not accurately model the initial intrusion of westerly winds, subsequent temperature-induced melt was limited over the eastern Larsen C ice shelf, where this AWS is located.

Presuming that the flow characteristics are largely similar in this relatively flat region, we conclude that the underestimation of melt in the CL region is partially due to the absence of westerly flow, but that this flow is adequately captured directly east of the AP (comprising the NL region).

Previous work has suggested that southwesterly föhn winds can result from gap flow (Elvidge et al., 2015), although we note that the southwesterly jets studied in this single campaign were typically cooler and moister than surrounding air, i.e., föhn flow produced from isentropic drawdown. While a version with a higher spatial resolution may potentially resolve topography sufficiently to include the initial intrusion of southwesterly gap flow, as well as northwesterly föhn flow, it may also further inhibit subsequent eastward flow when the hydrostatic assumption is retained. While a higher resolution of MAR v3.5.2 (used throughout this study) was not run due to computational...
constraints, the enhanced computational efficiency of a newer version of the MAR model (MAR v3.9, Section 2.1) could enable higher resolution runs over extended periods in the future.

To assess both the potential future application of MAR v3.9 over the AP as well as the effects of both vertical and horizontal resolution on modelled melt estimates, we compare melt occurrence and flow characteristics from Nov 1, 2004 to March 31, 2005 between multiple versions of the MAR model. This included three versions of v3.9 (Section 2.1), with two 5km and 10km resolution versions run with 24 vertical layers as well as an additional 10km resolution version with 32 vertical layers (10km V). The effect of the enhanced horizontal resolution on topography is substantial; the maximum height of the AP in the 5km version of the model is 2567m, but only 2340m in the 10km version. We find that the effect of increasing horizontal resolution to 5km is to limit the consistent strong melt production just leeward of the AP and that an increase in either horizontal resolution or vertical discretization limits eastward flow (Fig. S12). As compared to AWS data at the Larsen IS AWS, all MAR configurations largely replicated the dominant southwesterly and northeasterly flow, although we found an enhanced bias for southeasterly flow with the enhanced-resolution versions of the model (Fig. S13). The effects of local topography on wind speed should be relatively limited as the region surrounding the Larsen ice shelf AWS station is relatively flat. Bedmap2 (Fretwell et al., 2013) reports mean (standard deviation) elevation values of 37.38m (0.53m) in the 5km surrounding the station and 37.37m (0.78m) in the 10km surrounding the station. The mean (standard deviation) values for slope are 0.015°(0.018°) at both resolutions. We conclude that a further increase in vertical discretization or horizontal resolution may potentially reduce flow towards the eastern edge of the Larsen C ice shelf, although the effect of better-resolved topography may allow more westerly flow in MAR to cross the AP.

As has been suggested by previous studies (Van Wessem et al., 2015a), the implementation of a non-hydrostatic model may improve the representation of westerly föhn flow over the eastern Larsen Ice Shelf (Habert Gallée, personal communication). We note that previous work has suggested that a 5km non-hydrostatic model was still unable to capture föhn flow on the eastern portion of the Larsen C ice shelf (according to the AWS records), partially due to the inability to simulate southwesterly föhn jets, and that resolutions as high as 1.5km are required to simulate föhn flow accurately (Turton et al., 2017). However, recent work found that spatial resolutions as high as 2km in the non-hydrostatic WRF model were still unable to fully-resolve the steep surface temperature increases associated with the beginning of föhn flow (Bozkurt et al., 2018), suggesting that neither increased spatial resolution nor a non-hydrostatic model may be sufficient to fully capture the effects of föhn flow. We conclude from the main analysis that reduced eastward propagation of westerly winds may contribute to a lack of MAR melt in the CL region as compared to satellite estimates but that melt just east of the AP (the NL) region is represented with relative accuracy. This is further confirmed by the similarity between the spatial trends for melt occurrence as compared to QuikSCAT estimates. We remind the reader that previous work has suggested that föhn flow occurred only 20% of the time during a single melt season, and that substantial melt occurred in conditions where föhn winds are not present (King et al., 2017), suggesting that other factors contributing to surface melt energy may be equally, if not more, important for developing accurate melt estimates in RCMs. Because the current class of RCMs which employ the hydrostatic assumption, such as MAR, can be run for relatively long periods and contain relatively realistic representations of the snowpack, they can provide additional insights into the cumulative effects of surface melt over multiple seasons, with the
understanding that the surface melt produced by föhn flow will likely be under-represented in the eastern regions of
the Larsen C ice shelf.

Previous literature has pointed to several limitations in the remote sensing data sources used here which are either
intrinsic to the satellite data itself or a product of the algorithm selected for melt detection (Ashcraft and Long, 2006).
Products derived from QuikSCAT are limited in temporal resolution because the satellite passes daily, and may
therefore ignore sporadic melt occurring at other times of the day. However, previous studies have compared total
melt days from the QuikSCAT f03 algorithm with a measure derived from surface temperature at seven automatic
weather stations and shown a positive QuikSCAT f03 bias compared to AWS (Steiner and Tedesco, 2014). Similarly,
all PMW algorithms are limited by a relatively low resolution (25km) and twice-daily passes. Periods of melt
occurrence have also been shown to be sensitive to the choice of algorithm (Tedesco 2009).

In future work, we will extend this model run to the 1982-2017 period as well as explore a higher-resolution run
of a newer version of MAR, producing hourly outputs for the near-surface atmosphere. These runs will allow us to
examine the frequency of föhn winds, the concurrent meltwater production and the effects of föhn-induced melt on
the snowpack. We will use this multi-decadal record to examine interannual trends of föhn winds in all seasons as
well as the cumulative effect of a changing regional climate on the snowpack of the NE basin.

Acknowledgements. This research was funded by the National Science Foundation, grant 1131973. Thanks to the
University of Wisconsin Madison for the use of AWS data, to Dr. Nick Steiner for help with the processing of
QuikSCAT melt occurrence and especially to Dr. Patrick Alexander and Dr. Kate Briggs for their constructive reviews
of early versions of the manuscript.
References


Figure 1: Full MAR domain showing topographic relief from bedmap2 (https://www.bas.ac.uk/project/bedmap-2) at 1km, former ice shelves with dates of collapse, locations of automatic weather stations (Larsen IS and AWS 14 stations are located within the same MAR gridcell) and basins corresponding to SW (basin 24) NW (basin 25) NE (basin 26) SE (basin 27) from Zwally et al. 2012.
<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>MAR model</strong> : criteria for melt occurrence (Section 2.1)</td>
<td></td>
</tr>
<tr>
<td>LWC&lt;sub&gt;0.4&lt;/sub&gt;</td>
<td>liquid water content in the first meter is greater than 0.4 mm we (water equivalent)</td>
</tr>
<tr>
<td>MF&lt;sub&gt;0.4&lt;/sub&gt;</td>
<td>total meltwater production over the day exceeds 0.4 mmwe</td>
</tr>
<tr>
<td><strong>Passive microwave</strong> : criteria for melt occurrence (Section 2.2.2)</td>
<td></td>
</tr>
<tr>
<td>zwa</td>
<td>threshold based on winter mean temperature brightness, Zwally and Fiegles, 1994</td>
</tr>
<tr>
<td>ALA</td>
<td>threshold based on winter mean temperature brightness, Ashcroft and Long, 2006</td>
</tr>
<tr>
<td>240</td>
<td>fixed threshold method (Tedesco, 2007)</td>
</tr>
<tr>
<td>PMWAll</td>
<td>Condition when zwa, ALA, 240 all report melt occurrence</td>
</tr>
<tr>
<td><strong>Active microwave (QuikSCAT)</strong> : criteria for melt occurrence (Section 2.2.1)</td>
<td></td>
</tr>
<tr>
<td>QuikSCAT ft3</td>
<td>threshold based on winter mean backscatter (Steiner and Tedesco, 2014)</td>
</tr>
<tr>
<td><strong>Observation-based regions of high melt occurrence (Section 3.2)</strong></td>
<td></td>
</tr>
<tr>
<td>CL region</td>
<td>high melt at the center-east of the Larsen C ice shelf, melt days exceeding 1 std dev of PMWAll mean melt occurrence</td>
</tr>
<tr>
<td>NL region</td>
<td>high melt in the north and west of the NE basin, consisting of the NE basin above the mean latitude of CL region which excludes the CL region</td>
</tr>
<tr>
<td><strong>Conditions for melt occurrence (Section 4.2)</strong></td>
<td></td>
</tr>
<tr>
<td>PMWEx</td>
<td>PMWAll reports melt occurrence but MAR does not</td>
</tr>
<tr>
<td>QSEx</td>
<td>QuikSCAT ft3 reports melt occurrence but MAR does not</td>
</tr>
<tr>
<td>MAR-R</td>
<td>criteria when MAR data is used only when AWS data is available</td>
</tr>
</tbody>
</table>

Table 1: Abbreviations used throughout text
Figure 2 Annual meltwater production from MAR [Gt/yr] shown for masks shown in inset. ‘2001’ corresponds to meltwater production from July 2000 - June 2001. NW, SW, SE basins are shown as in Fig. 1. NE basin is divided into the NL mask, the CL mask and the remaining portion of the NE basin (NE – (CL+NL)). The CL and NL masks are described in text.
Figure 3 Meltwater production (2000-2009). RACMO2.3p2 at 5.5 km resolution, mean annual meltwater production (a) and standard deviation (c) and MAR v. 3.5.2 at a 10km resolution, mean annual meltwater production (b) and standard deviation (d).
Melt Occurrence at the Larsen Ice Shelf AWS Threshold = 0.4 mm w.e.

MAXT2m(AWS) < 0°C  MAXT2m(AWS) ≥ 0°C

Data Sources:

Satellite
Z PMW 240
A PMW ALA
Z PMW zwa
Q QuikSCAT

AWS-based
T AvgT2m > 0°C
M MaxT2m > 0°C

Melt Occurrence:

- MAR Only
- MAR & Obs
- Obs. Only

Figure 4 Melt Occurrence and Temperature Biases at the Larsen Ice Shelf AWS Station. Percentage of total days (DJF, 2001-2010) showing melt occurrence from observational sources as compared to MAR v3.5.3 melt occurrence using the MF0.4 metric (a) Temperature biases (MAR-AWS) for AvgT2m (b,c) and MaxT2m (d,e) when Max T2m is less than 0°C (b,d) or greater than 0°C (c,e).

Formatted: Keep with next

Formatted: Caption, Left, Indent: Left: 0°
Figure 5 Average number of melt days (2000-2009) from multiple sources (a) MAR, Liquid Water Content > 0.4% for three consecutive days, (b) MAR Total Melt Flux > 0.4 mm w.e. for 1 day or more (c) RACMO2.3p2, Melt Flux > 0.4 mm w.e. Satellite-based metrics include (d) PMW 240 algorithm (e) PMW ALA (f) PMW Zwa (g) QuikSCAT. All satellite-based estimates include a melt day only when part of a sustained three-day period of melt.
Figure 6 CL-region, described in text and shown in inset in for (a), average and inter-annual melt occurrence in MAR, PMW and QuikSCAT data. (a) MF0.4 melt extent climatology with one standard deviation shown in grey envelope (b) melt extent for MF0.4 from 1999-2009 (c) melt climatology PMW All (d) interannual difference melt extent PMWAll - MAR (e) melt climatology PMW zwa (f) interannual difference in melt extent PMWzwal - MAR (g) melt climatology QuikSCAT ft3 (h) interannual difference in melt extent QuikSCAT ft3 - MAR
Figure 7: NL-region, described in text and shown inset in (c), average and inter-annual melt occurrence in MAR, PMW and QuikSCAT data. (a) $MF_0.4$ melt extent climatology with one standard deviation shown in grey envelope (b) melt extent for $MF_0.4$ from 1999-2009 (c) melt climatology PMW All (d) interannual difference melt extent PMWAll - MAR (e) melt climatology PMW zwa (f) interannual difference in melt extent PMW zwa - MAR (g) melt climatology QuikSCAT ft3 (h) interannual difference in melt extent QuikSCAT ft3 - MAR
Figure 8 Probability distribution (y-axis) of summer (DJF) wind speeds (x-axis) and direction proportions inset. Wind directions corresponding to colors in 45° increments shown right of (g). Curve shows Weibull curve shape (β) and scale (λ, m/s). Datasets for AWS (col 1), MAR-R (col 2), MAR from 1999-2014 period (col 3) and ERA-Interim for the AWS-restricted period (col 4). Shown for station Larsen IS (row 1, a,b,c,d), AWS 14 (row 2, e,f,g), AWS 15 (row 3, h,i,j,k) Values below figures are expected values.
<table>
<thead>
<tr>
<th></th>
<th>DJF All Days</th>
<th>MAR shows wind direction</th>
<th>AWS shows wind direction</th>
<th>DJF, MAR reports melt</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>DJF (0°-90°)</td>
<td>MAR (90°-180°)</td>
<td>SW (180°-270°)</td>
<td>NW (270°-360°)</td>
</tr>
<tr>
<td>Max T2m</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Avg T2m</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Temp.</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Avg T2m</td>
<td>-0.09°C</td>
<td>-1.11°C</td>
<td>-0.53°C</td>
<td>-0.98°C</td>
</tr>
<tr>
<td>Max T2m</td>
<td>-2.57°C</td>
<td>-3.10°C</td>
<td>-1.06°C</td>
<td>-1.61°C</td>
</tr>
<tr>
<td>Temp. biases</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Temp. bias where T2m &gt; 0°C (MAR-AWS)</td>
<td>0.05°C</td>
<td>0.05°C</td>
<td>0.05°C</td>
<td>0.05°C</td>
</tr>
<tr>
<td>Max T2m</td>
<td>-2.11°C</td>
<td>-2.20°C</td>
<td>-0.92°C</td>
<td>-1.43°C</td>
</tr>
<tr>
<td>Temp. biases</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Temp. bias where T2m &gt; 0°C (MAR-AWS)</td>
<td>0.09°C</td>
<td>0.09°C</td>
<td>0.09°C</td>
<td>0.09°C</td>
</tr>
<tr>
<td>Avg T2m</td>
<td>-0.09°C</td>
<td>-1.11°C</td>
<td>-0.53°C</td>
<td>-0.98°C</td>
</tr>
<tr>
<td>Max T2m</td>
<td>-2.57°C</td>
<td>-3.10°C</td>
<td>-1.06°C</td>
<td>-1.61°C</td>
</tr>
</tbody>
</table>

Table 2: Proportions for wind direction and associated temperature biases at the Larsen Ice Shelf AWS station from 2000-2009 restricted to the summer season (DJF)
Figure 9 MAR vs AWS temperatures at the Larsen Ice Shelf AWS station for DJF from 2001-2009 for melt occurrence criteria as shown bottom-right and described in text. Wind direction biases are shown for when northerly AWS flow is reported as southerly in MAR (a) AvgT2m (b) MaxT2m, when westerly AWS flow is reported as easterly in MAR (c) AvgT2m (d) MaxT2 and when AWS and MAR both report westerly flow (e) AvgT2m (f) MaxT2m.
The underestimation of föhn flow in the east of the Larsen C may potentially be resolved by removing the hydrostatic assumption in MAR or increasing spatial resolution. The underestimation of southwesterly flow in particular may be reduced by using higher-resolution topography.

This is primarily caused by more exposure to open water in combination with prevailing westerly winds on the west AP and southerly winds on the east AP. Moreover, when strong westerly winds cross the bisecting mountain range of the AP (Fig. 1), the resulting föhn winds can produce pulses of warming on the East Antarctic Peninsula’s ice shelves (Marshall, 2007). Föhn winds are a warm, dry air flow on the lee slopes of a mountain range (Beran 1967). This resultant warming can be produced by four main mechanisms. Elvidge (2016) uses a modeling approach to trace four physical processes that occur during föhn flow in the East AP, namely isentropic drawdown (sourcing of föhn air from higher altitudes), latent heating and precipitation (where cooling during uplift on the windward side promotes precipitation), mechanical mixing (turbulent sensible heating and drying of low-level flow) and radiative heating (where cloudless conditions on the lee side increase the availability of shortwave radiation for heating). The relative importance of each of these mechanisms for surface melt has been shown to be related to the source of föhn flow in the East AP (Elvidge et al., 2015; Grosvenor et al., 2014). For example, southwesterly föhn jets descending from gap flow (from lower-elevation passages in the mountain range) have been shown to be cooler and moister than surrounding föhn flow descending from higher elevations (Elvidge et al., 2015). Recent warming in the East AP has been linked to an increase in föhn winds during recent warming (Cape et al., 2015), which were possibly related to an increase in the speed of warm northwesterly winds which have been associated with positive phases of the Southern Annular Mode (SAM), (Van den Broeke and Van Lipzig, 2003). Because melt in the East AP is as vulnerable to wind dynamics as it is to regional temperature changes, an accurate depiction of föhn flow is crucial for accurate estimates of meltwater production.

Fig. 8 shows wind frequency distributions during the summer season, color-coded for wind direction as represented by the pie graph at the right. We note that AWS data are 3-hourly averages and ERA-Interim are 6-hourly averages for wind speed and direction, while MAR produced daily-averaged outputs. For this reason, a direct comparison between Weibull parameters derived from MAR vs AWS data is not fully justified. The Larsen Ice Shelf AWS has full temporal coverage during the QuikSCAT period while AWS14 and AWS15 were installed after termination of the QuikSCAT mission. These last two stations are used in this study to demonstrate that (a) similar wind biases persisted after the QuikSCAT period at multiple locations, as AWS 14 the Larsen Ice Shelf AWSs are co-located to the same MAR grid cell and that (b) wind biases vary slightly by latitude, AWS15 being located slightly to the south.

Both MAR and AWS are dominated by northerly winds at lower windspeeds (in yellow and blue) although AWS data shows a greater frequency of southwesterly winds when windspeeds are higher (> 8 m/s). This is especially relevant at the southern AWS15, where modeled temperature correlates with a larger portion of the southern Larsen
C Ice Shelf than for AWS14 (Supplemental Fig. S7). All AWSs show more southwesterly flow and slightly more northwesterly flow than either MAR-R or MAR, which show a substantially higher percentage of easterly flow instead, a trend which is more pronounced at the southernmost AWS15 (Fig. 7i,j). ERA-Interim reports substantially more northwesterly flow than either AWS or MAR and a smaller proportion of southwesterly flow in the 180°-225° range (especially at the southernmost AWS15 location), although the proportion of easterly flow is similar to that reported by AWSs. We note that although ERA-Interim has been shown to reproduce the basic structure of föhn flow (Grosvenor et al., 2014), the resolution may be too coarse to adequately capture southwesterly gap flow here. As discussed further in Sect. 5, westerly flow towards the stations used in this study may be strongly affected by the fine-scale representation of topography (which is coarse in ERA-Interim) and the lowered orographic barrier in the northwest in ERA-Interim may contribute to the enhanced northwesterly flow shown here.

Specifically, at the Larsen Ice Shelf AWS location, both AWS and MAR reports dominant northeasterly flow (Table 2, rows 4, 8, col2). However, the AWS reports slightly more flow which is either southwesterly (28.9% vs. 23.2% in MAR) or northwesterly (19.3% vs. 14.1% in MAR) while MAR reports more southeasterly flow overall (23.5% vs. 17.4% in AWS). Melt occurrence (from PMW and QS) is observed primarily when AWS-observed flow is northeasterly (0°-90°) or southwesterly (180°-270°), with QS(PMW) reporting that 36%(42%) northeasterly flow and 29%(26%) southwesterly flow. On days when MAR reports melt (Table 2, rows 19, 20), southeasterly flow in MAR is even more dominant (but declines at the AWS) while northwesterly flow decreases (while it increases at the AWS). The bias towards easterly flow affects 26% of all days and 10% of melt days in MAR, 21%(18%) of all days where QS(PMW) report melt occurrence, but only 8%(9%) of days where PMW(QS) melt occurrence is not also captured by MAR. Similarly, the bias towards southerly flow captures 26% of all days and 8% of melt days in MAR, 13%(15%) of days where QS(PMW) report melt occurrence, but only 6%(6%) of days when PMW(QS) melt occurrence is not also reported by MAR. Most notably, for 4% of all melt days in MAR, AWS reports southwesterly winds while MAR reports southeasterly winds and this bias accounts for 3%(4%) of days when PMW(QS) report melt but MAR does not. In summary, despite biases in wind directions reported by MAR, the overall impact on melt occurrence is fairly limited according to comparisons with satellite estimates. Within the next section we prominent wind direction biases in greater detail.

Fig. 5 shows wind frequency distributions during the summer season (AWS, MAR-R, MAR from left to right columns, Larsen IS station, AWS 14 and AWS 15 from top to bottom), color-coded for wind direction as represented by the pie graph at the top. We note that AWS data uses 3-hourly data for wind speed and direction without daily-averaging, while MAR produced daily-averaged outputs. For this reason, a direct comparison between Weibull parameters derived from MAR vs AWS data is not meaningful, although we show comparisons between different stations (for the same data source). The Larsen IS station has full temporal coverage during the QuikSCAT period while AWS14 and AWS15 were installed after termination of the QuikSCAT mission. These last two stations are used in this study to demonstrate the consistency of wind biases at multiple locations, as well as how wind biases vary by latitude (AWS15 being located slightly to the south). Whereas MAR is dominated northerly winds at it’s lower range of windspeeds (in yellow and blue), AWS data shows a greater frequency of southwesterly winds at the higher range of
winds speeds (> 8 m/s). In general, even at lower wind speeds (2-5 m/s), AWS data shows more southwesterly winds than either MAR-R or MAR. This is especially relevant at the southern AWS15 station, where modeled temperature correlates with a larger region of the Larsen IS than temperatures modeled by MAR for the AWS14 station (Supplemental Fig. 6). Observed wind direction (without consideration for wind speed) at AWS15 shows more southwesterly flow (Fig. 5g) than either MAR-R or MAR (Fig. 5 h,i), which show a substantially higher percentage of southeast erly and northerly flow instead.

Specifically, while both MAR and AWS show a higher proportion of northerly (vs southerly) winds, the proportion of northerly winds in MAR is slightly higher (Table 1). While both MAR and AWS report a larger proportion of easterly (vs westerly) flow, MAR reports 64% of flow to be easterly where AWS reports only 55% easterly flow. We show that MAR melt occurrence at the Larsen Ice Shelf station is concurrent with both increased northerly and westerly flows. On days when MAR reports melt (compared to days with no melt), northerly winds are more frequent (according to both MAR and AWS estimates), and the proportion of westerly winds increases slightly in MAR but decreases slightly in AWS data (Table 1).

When daily-averaged temperature (AvgT2m) values are high, it is more likely that melt is sustained, while high maximum daily temperatures (MaxT2m) can also occur during sporadic melt. Melt occurrence is strongly influenced by the temperature of the snow column as well as at the surface; internal melting can occur even when the surface is frozen due to net outgoing longwave radiation (Holmgren 1971)(Hock 2005) In general, we find a small, but consistent warm MAR bias for AvgT2m, and a consistent cold MaxT2m bias. However, when we restrict the dataset to days when AWS 2m-temperature estimates exceed 0°C, (a condition where melt is most likely), MAR indicates a cold bias for AvgT2m and an enhanced cold bias for MaxT2m, i.e. while MAR shows an overall warm bias, this bias is reversed at the temperature ranges where melt is likely (although melt is still possible due to other elements in the energy balance). On days when MAR meltwater production meets the $MF_{0.4}$ criteria, we find that the magnitude of all biases is greatly reduced. In particular, we note that the MaxT2m bias for westerly winds (for the ALL condition) is substantial, showing a -2.42°C cool bias in MAR for all available data and -3.04°C when data is restricted to days when MaxT2m > 0°C.

Three wind direction biases are dominant on days when observed sources (either PMW All or QuikSCAT ft3) report melt, but MAR does not. We refer to the condition where PMWAll reports melt (but MAR does not) as “PMWEx” (i.e. PMW exclusive-or), with the equivalent condition for QuikSCAT ft3 called “QSEx”. We focus on these specific wind biases and find the associated temperature biases. Supplemental tables 2-7 include $R^2$, RMSE and mean bias values for both surface pressure and daily AvgT2m at all three stations.

4.2.1 Observed northeasterly flow

The largest proportion of melt occurrence for either MAR, PMWEx or QSEx is reported when northeasterly winds are dominant, specifically when winds are recorded by the Larsen Ice Shelf AWS station as northeasterly (0°- 90°). Northeasterly AWS flow accounts for a large proportion of general flow and an even larger proportion of flow when
either MAR or PMWEx report melt, i.e. 36% of general flow, 39% of days where MAR reports melt, 42% of days where PMWEx reports melt and 36% of days where QSEx reports melt. On days when AWS reports northeasterly winds, MAR primarily reports northeasterly flow (0°-90°, case 1), but also reports a substantial bias for northwesterly flow (270°-360°, case 2). In case 2, associated temperature biases may be influenced by the inclusion of warmer westerly winds in MAR. We add that the majority of northeasterly AWS flow is actually captured in a narrower northeasterly band in MAR from 0-45°, accounting for 13.4% of ALL days. We examine these two cases separately to quantify how a modeled westerly wind bias affects temperature and melt (in comparison to the case when wind direction matches observed estimates). Supplemental tables 8-10 contain relative proportions of each case (flow bias) divided by the general melt restriction (i.e. MAR, QSEx or PMWEx), as well as the timeseries mean and biases for AvgT2m, AvgT2m>0°C (excluding days when AvgT2m values from AWS are below 0°C), MaxT2m and MaxT2m>0°C.

In the instance where northeasterly flow is modeled accurately (case 1, Fig. 6a), modeled temperature values are clustered around 0°C, whereas AWS-observed temperatures (especially when only satellite-observed melt occurs) are higher. When MAR reports melt, MAR AvgT2m values cluster near 0°C, with a small overall warm bias (0.69°C). Under omission conditions (PMWEx and QSEx), AvgT2m values are lower, and MAR bias is slightly cold, although standard deviation is high. As with all flow cases, only QuikSCAT ft3 shows melt at very low observed AvgT2m values. By contrast, AWS MaxT2m values are substantially higher than MAR values (the latter clustering around 0°C) (Fig. 6b). We find that where QuikSCAT ft3 uniquely reports melt (QSEx), AvgT2m values at the lower range of temperatures report a stronger cool MAR bias for MaxT2m.

In case 2, AvgT2m values show a small warm MAR bias in all melt conditions, i.e. for ALL data points (0.79°C), for when MAR shows melt (0.82°C), and finally for both the PMWEx (0.44°C) and QSEx(0.65°C) conditions (Supplemental Table 8). However, when MAR reports melt and AWS AvgT2m values exceed 0°C, there is a substantial cold bias (-1.13°C)(Fig 6c), which may lead to reduced meltwater production. In contrast to case 1, modeled MaxT2m values in case 2 do not cluster around 0°C (Fig. 6b,d) and MAR melt days report larger MaxT2m values.

In summary, when MAR reports westerly flow, Avg T_s values are higher (as is melt occurrence). As with the comparison with case 1 and case 2, we find that in all instances when MAR reports northeasterly flow (with all AWS-observed wind directions considered), AvgT2m and MaxT2m temperatures cluster near 0°C (Fig. 6e,f), whereas when MAR reports northwesterly flow (with all AWS wind directions taken into consideration), MaxT2m values are, on average, higher, and the temperature bias is narrowed (Fig. 6g,h). Expected values for windspeeds for each condition based on a Weibull fit show comparable expected values, but larger standard deviations for AWS-estimated windspeeds when MAR reports northwesterly flow (Fig. 6 c,d g,h).

### 4.2.2 Observed southwesterly flow

For all days in the summer season (“ALL”, i.e. without regard for melt occurrence), we find that MAR reports 15% of winds to be southwesterly at the Larsen Ice Shelf station location while AWS reports ~ 30% southwesterly flow. We note that the relative proportions of southwesterly/southeasterly flow in AWS is approximately reversed in MAR,
with AWS reporting 18.3% of flow to be southwesterly and MAR reporting 29.2% southwesterly flow. We focus specifically on the condition where both MAR and AWS report southwesterly flow (between 180° and 270°), which accounts for 5.7% of total flow and only 4.9% of MAR melt days, but a larger proportion of days where only observed sources report melt, i.e. 5.6% for the PMWEx condition and 8% for the QSEx condition.

As with case 2 for northwesterly winds in section 4.2.1, MAR captures higher AvgT2m values which frequently exceed 0°C, with a slight cool MAR bias when AvgT2m > 0°C (Fig 7a). The PMWEx and QSEx conditions report melt at lower temperature values, where the MAR bias is slightly warmer. Although a cold MAR bias persists, MaxT2m values are, in general, higher with AWS and MAR values showing greater agreement (Fig 7b). Expected windspeeds for southwesterly winds are substantially higher (with a greater standard deviation) than the base condition when wind direction is not considered, with AWS reporting an even higher standard deviation (i.e. high-speed sporadic winds).

In the aggregate, we conclude that MAR shows lower melt occurrence than satellite estimates in the center and east of the Larsen C ice shelf (i.e. the CL region, where eastward föhn flow is likely limited in MAR ), while in the north and west of the NE basin (i.e. the NL region which is most immediately affected by föhn flow), MAR reports melt occurrence largely concurrent with satellite estimates. For example, within the CL region, there are periods during the 2001-2002 season when MAR reports no meltwater production, but raw QuikSCAT backscatter values report periods where over 300 km² of surface area show backscatter values dipping below -15 dB (Supplemental Fig. 8e). We remind the reader that raw backscatter values from QuikSCAT have previously been used to estimates melt flux over the AP (Trusel et al., 2013; Trusel et al., 2012).

In comparison to AWS estimates, MAR displays a general warm bias in the East AP at lower temperatures where melt is less likely to occur, but which may still impact the refreeze process. However, when maximum daily temperatures (MaxT2m) and average daily temperatures (AvgT2m) exceed 0°C, MAR shows a substantial cold bias which may limit melting. We note a smaller proportion of westerly winds in MAR compared to observed values at the Larsen Ice Shelf AWS station, especially an absence of southwesterly flow, which tends to have higher observed windspeeds. The general cold bias in MAR is partially closed when observed northeasterly winds are reported by MAR as northwesterly, i.e. MAR reports both higher AvgT2m and MaxT2m values as well as greater melt occurrence. Similar biases are shown for southwesterly flow, which accounts for a disproportionate amount of satellite-observed melt which is not captured by MAR. The importance of westerly winds is demonstrated during mid-December in the 2001-2002 season, at which point satellite-based melt extent in the CL region increases substantially, while MAR melt extent declines (Supplemental Fig. 8a). This period is concurrent with an increase in westerly winds at the Larsen IS AWS station which are not modeled MAR (Supplemental Fig 9b vs f). Additionally, we note that shortly after this point, AWS AvgT2m temperatures consistently exceed MAR AvgT2m values until the end of the season (Supplementary Fig. 10).

Previous work has suggested that southwesterly föhn winds can result from gap flow (Elvidge et al. 2015), although we note that the southwesterly jets studied in this single campaign were typically cooler and moister than surrounding air (i.e. föhn flow produced from isentropic drawdown). We note that the low windspeed bias in MAR...
may have a minor impact overall, but could strongly impact melt in the East AP if southwesterly flow is more accurately captured in future versions of MAR. We hypothesize that the underestimation of westerly flow at the eastern reaches of the Larsen C ice shelf is likely due to the hydrostatic assumption (allowing for no vertical acceleration of air mass) preventing eastward, downward flow in the near-surface atmosphere. The implementation of a non-hydrostatic model will likely be required to fully capture fohn flow in the East AP (Hubert Gallée, personal communication). We conclude that the relative absence of fast, warm westerly and southwesterly winds contributes to a lack of MAR melt in the CL region as compared to satellite estimates.

Previous literature has pointed to several limitations in the remote sensing data sources used here which are either intrinsic to the satellite data itself or a product of the algorithm selected for melt detection. Products derived from QuikSCAT are limited in temporal resolution because the satellite passes at a twice-daily scale, and may therefore ignore sporadic melt occurring at other times of day. However, previous studies have compared total melt days from the QuikSCAT ft3 algorithm with a measure derived from surface temperature at seven automatic weather stations and shown a positive QuikSCAT ft3 bias compared to AWS (Steiner and Tedesco, 2014). Similarly, all PMW algorithms are limited by a relatively low resolution (25km) and twice-daily passes. Periods of melt occurrence have also been shown to be highly sensitive to the choice of algorithm (Tedesco 2009). Because of the high topographic variability of the NL region (especially near the spine of the AP), it is possible that PMW algorithms are under-reporting melt occurrence due to low horizontal spatial resolution. A higher-resolution passive microwave product may better resolve this issue.

In the northernmost portions of the NL region, sporadic MAR-modeled meltwater percolates deeply into the snowpack in November (as deep as 10m early in the season in some years), which is consistent with MAR $MF_{0.4}$, PMW zwa and QuikSCAT ft3 reporting melt occurrence at this point while other algorithms/melt metrics do not. The deep percolation of meltwater is potentially enabled by low density snow early in the season. This early-season melt is frequently followed by a near-complete refreeze. Future work will focus on the interannual variability of early-season melt as this may have a substantial impact on the density of the firn layer in the Larsen C ice shelf.

In light of the biases reviewed here, we report MAR meltwater production over the 1999-2009 period (Fig. 8) and consider the potential implications of the wind/temperature biases found in this analysis on regional meltwater production. Over the full study domain, the total annual meltwater production estimated by MAR shows substantial inter-annual variation with the NE basin accounting for the highest aggregate meltwater production, closely followed by the SW basin (in green). The NE basin is divided into three regions: the NL and CL masks and the remainder of the basin. We note that the SW basin does not covary with the NE basin (with all subregions taken together) and the subregions of the NE basin do not consistently covary with one another. The meltwater production shown here does not account for refreeze and we note that the effects of refrozen melt on the snowpack will vary regionally depending on local properties. The NL region dominates meltwater production in the NE basin in most years except for 1999-2000, 2002-2003 and 2003-2004. The 2001-2002 melt season shows the second lowest overall melt production during the study period (only the preceding year is lower). Declining aggregate meltwater production across the AP does not necessarily correspond to declining meltwater production in the most vulnerable regions of the northeastern AP (including the Larsen C ice shelf). Because melt in the NL region is particularly sensitive to fohn-induced melt, we
note that changes in circulation patterns may affect the northwest regions differently than the southern regions. The
strong relationship between wind direction and temperature bias points to the need for isolating dominant inter-annual
patterns of melt in the Northern Larsen C Ice Shelf and associating them with large-scale atmospheric drivers.

References
Abdalati, W., and Steffen K.: Passive Microwave-Derived Snow Melt Regions on the Greenland Ice Sheet,
Ahrens, C. Donald.: Meteorology Today : An Introduction to Weather, Climate, and the Environment. Eighth
H.: Comparison between Observed and Simulated Aeolian Snow Mass Fluxes in Adélie Land, East Antarctica., The
Ashcraft, I. S., and Long. D.G.: Comparison of Methods for Melt Detection over Greenland Using Active and
Hosking, J. S.: Trends in Antarctic Peninsula Surface Melting Conditions from Observations and Regional
Climate Modeling, Journal of Geophysical Research: Earth Surface, 118 (1), 315–330,
Bell, R. E., Chu, W. Kingslake, J., Das, I., Tedesco, M., Tinto, K.J., Zappa, C.J., Frezzotti, M., Boghosian, A., and
Braithwaite, R. J.: On Glacier Energy Balance, Ablation, and Air Temperature, Journal of Glaciology, 27, 381–91,
1981


Trusel, L. D., Frey, K. E., and Das, S. B.: Antarctic Surface Melting Dynamics: Enhanced Perspectives from Radar


Figure 1: Full MAR domain showing topographic relief, former ice shelves with dates of collapse, locations of automatic weather stations and basins corresponding to SW (basin 24) NW (basin 25) NE (basin 26), SE (basin 27) from Zwally, et. al. 2012
Figure 2: Average number of melt days from multiple sources (a) MAR, Liquid Water Content > 0.4% for three consecutive days. (b) MAR Total Melt Flux > 0.4 mmwefor 1 day or more. Satellite-based: (c) PMW 240 algorithm (d) PMW ALA (e) PMW Zwa (f) QuikSCAT. All satellite-based estimates include a melt day only when part of a sustained three-day period of melt. All averages are taken from the 2000-2009 period to retain consistency with the availability of QuikSCAT data.
MAR melt climatology

(b) MAR interannual

(c) PMWAll melt climatology

(d) PMWAll - MAR interannual

(e) zwa melt climatology

(f) zwa - MAR interannual
Figure 3: CL-region, described in text and shown in inset in (a), average and inter-annual melt occurrence in MAR, PMW and QuikSCAT data. (a) $M_{0.4}$ melt extent climatology with one standard deviation shown in grey envelope (b) melt extent for $M_{0.4}$ from 1999-2009 (c) melt climatology PMW
All (d) interannual difference melt extent PMWAll - MAR (e) melt climatology PMW zwa (f) interannual difference in melt extent PMWzwal – MAR

(g) melt climatology QuikSCAT ft3 (h) interannual difference in melt extent QuikSCAT ft3 - MAR

MAR melt climatology (b) MAR interannual

(c) PMWAll melt climatology  
(d) PMWAll - MAR interannual
Figure 4: NL-region, described in text and shown inset in (c), average and inter-annual melt occurrence in MAR, PMW and QuikSCAT data. (a) MF$_{0.4}$ melt extent climatology with one standard deviation shown in grey envelope (b) melt extent for MF$_{0.4}$ from 1999-2009 (c) melt climatology PMW All (d) interannual difference melt extent PMWAll - MAR (e) melt climatology PMW zwa (f) interannual difference in melt extent PMWzwal – MAR (g) melt climatology QuikSCAT ft3 (h) interannual difference in melt extent QuikSCAT ft3 - MAR
Larsen IS (1999-2014)

4.07 ±5.95σ m/s  
(a) All Days (AWS)

3.47±2.78σ m/s  
(b) All Days (MAR-R)

3.42±2.68σ m/s  
(c) All Days (MAR)

AWS 14 (2009-2014)

3.80 ±5.03σ m/s  
(d) All Days (AWS)

3.26±2.43σ m/s  
(e) All Days (MAR-R)

3.42±2.68σ m/s  
(f) All Days (MAR)
Figure 5: Probability distribution (y-axis) of summer (DJF) wind speeds (x-axis) with wind speed direction proportions shown in each inset. Wind directions corresponding to inset colors shown in 45° increments in inset left of (a). Curve shows best-fit Weibull curve with parameters for shape ($\beta$) and scale ($\lambda$, m/s). Datasets are for AWS data (left column), MAR outputs restricted to AWS availability (middle column) and MAR data for the 2001-2014 period (right column). Shown for (a) Larsen, AWS (b) Larsen, MAR at AWS availability (c) Larsen, MAR full period (d) AWS 14, (e) AWS 14, MAR at AWS availability (f) AWS 14, MAR full period (g) AWS 15, AWS (h) AWS 15, MAR at AWS availability (i) AWS 15, MAR full period. Values shown below figures are expected values from the Weibull distribution

AWS 15 (2008-2014)
**DJF All Days**

<table>
<thead>
<tr>
<th>Wind Direction</th>
<th>Northerly</th>
<th>Southerly</th>
<th>Easterly</th>
<th>Westerly</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Where MAR shows wind direction</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>MAR wind direction percentage</td>
<td>55.6%</td>
<td>44.4%</td>
<td>63.9%</td>
<td>36.3%</td>
</tr>
<tr>
<td>MAR expected wind speed [m/s]</td>
<td>3.49(±3.12)</td>
<td>4.21(±4.83)</td>
<td>3.75(±3.09)</td>
<td>4.04(±5.69)</td>
</tr>
<tr>
<td>AWS expected wind speed [m/s]</td>
<td>3.82(±5.23)</td>
<td>5.10(±9.14)</td>
<td>4.4(±6.87)</td>
<td>4.44(±8.73)</td>
</tr>
<tr>
<td><strong>Where AWS shows wind direction</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>AWS wind direction percentage</td>
<td>51.2%</td>
<td>48.8%</td>
<td>54.5%</td>
<td>45.5%</td>
</tr>
<tr>
<td>MAR expected wind speed [m/s]</td>
<td>4.04(±5.31)</td>
<td>3.72(±3.07)</td>
<td>3.48(±2.37)</td>
<td>4.43(±6.26)</td>
</tr>
<tr>
<td>AWS expected wind speed [m/s]</td>
<td>4.52(±8.33)</td>
<td>4.39(±6.98)</td>
<td>3.90(±4.77)</td>
<td>5.22(±11.03)</td>
</tr>
</tbody>
</table>

**Temperature biases**

| Average T2m Bias (MAR - AWS) | -0.58°C | 0.77°C | 0.43°C | 0.76°C |
| Maximum T2m Bias (MAR - AWS) | -2.21°C | -1.11°C | -1.75°C | -2.42°C |

**Temperature bias where T2m > 0°C**

| Average T2m Bias (MAR - AWS) | -1.25°C | -1.42°C | -1.32°C | -1.29°C |
| Maximum T2m Bias (MAR - AWS) | -2.80°C | -2.40°C | -2.84°C | -3.04°C |

**DJF, MAR reports melt**

<table>
<thead>
<tr>
<th>Wind Direction</th>
<th>Northerly</th>
<th>Southerly</th>
<th>Easterly</th>
<th>Westerly</th>
</tr>
</thead>
<tbody>
<tr>
<td>MAR wind direction</td>
<td>59.5%</td>
<td>40.5%</td>
<td>62.8%</td>
<td>37.2%</td>
</tr>
<tr>
<td>AWS wind direction</td>
<td>55%</td>
<td>45%</td>
<td>56.7%</td>
<td>43.3%</td>
</tr>
</tbody>
</table>

**Temperature biases**

| Average T2m Bias (MAR - AWS) | 0.28°C | 0.18°C | 0.25°C | 0.24°C |
| Maximum T2m Bias (MAR - AWS) | -0.78°C | -0.41°C | -0.73°C | -0.43°C |

**Temperature bias where T2m > 0°C**

| Average T2m Bias (MAR - AWS) | -0.69°C | -0.76°C | -0.47°C | -0.63°C |
| Maximum T2m Bias (MAR - AWS) | -1.06°C | -0.92°C | -1.31°C | -0.64°C |
Table 1: Proportions for wind direction and associated temperature biases at the Larsen Ice Shelf AWS station from 2000-2009 restricted to the summer season (DJF)
Figure 6: MAR vs AWS temperatures at the Larsen Ice Shelf AWS station for DJF from 2001-2009 for wind biases AWS -> MAR shown by pie charts (when AWS data is available). Red shows days where MAR shows melt. Blue shows “QSex” days, i.e. when QuikSCAT reports melt (and MAR does
not), Cyan indicates “PMWEx” when PMWAll shows melt and MAR does not. Green indicates when PMWEx an QSEx both report melt (and MAR does not). Yellow (only shown for g,h) indicates all other days for completeness. Northeasterly agreement (case 1) (a) AvgTs (b) MaxTs, Northeasterly
AWS winds reported as northwesterly in MAR (case 2) (c) AvgTs (d)MaxTs, All wind directions in AWS reported as northeasterly in MAR (e) AvgTs (f) MaxTs

MAR: 4.31(±4.74) m/s  AWS: 4.81 (±7.44) m/s

MAR: 3.83(±3.53) m/s  AWS: 4.36 (±6.26) m/s
Figure 7: MAR vs AWS temperatures at the Larsen Ice Shelf AWS station for DJF from 2001-2009 for wind biases AWS → MAR shown by pie charts (when AWS data is available). Red shows days where MAR shows melt. Blue shows “QSEx” days, i.e. when QuikSCAT reports melt (and MAR does not), Cyan indicates “PMWEx” when PMWAll shows melt and MAR does not. Green indicates when PMWEx an QSEx both report melt (and MAR does not). Yellow (only shown for g,h) indicates all other days for completeness. Southwesterly wind direction agreement (a) AvgTs (b) MaxTs, All wind directions in AWS which are reported as southwesterly in MAR (c) AvgTs (d) MaxTs,
Figure 8: Annual meltwater production from MAR [Gt/yr] shown for masks shown in inset (‘2001’ corresponds to the meltwater from July, 2000-June, 2001). NW, SW, SE basins are kept intact as in Fig. 1. NE basin is divided into the NL mask, the CL mask and the remaining portion of the NE basin (NE – (CL+NL)). The CL and NL masks are described in text.
<table>
<thead>
<tr>
<th>Page 38: [31] Deleted</th>
<th>Microsoft Office User</th>
<th>5/19/18 10:49:00 AM</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>4.44</td>
</tr>
<tr>
<td>Page 38: [32] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:55:00 AM</td>
</tr>
<tr>
<td>Left, Line spacing: single</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Page 38: [33] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:52:00 AM</td>
</tr>
<tr>
<td>Line spacing: single</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Page 38: [34] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:55:00 AM</td>
</tr>
<tr>
<td>Left, Line spacing: single</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Page 38: [35] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:52:00 AM</td>
</tr>
<tr>
<td>Line spacing: single</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Page 38: [36] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:55:00 AM</td>
</tr>
<tr>
<td>Left, Line spacing: single</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Page 38: [37] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:52:00 AM</td>
</tr>
<tr>
<td>Line spacing: single</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Page 38: [38] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 10:52:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>4.04</td>
</tr>
<tr>
<td>Page 38: [38] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 10:52:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>4.04</td>
</tr>
<tr>
<td>Page 38: [39] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 10:57:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>72</td>
</tr>
<tr>
<td>Page 38: [39] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 10:57:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>72</td>
</tr>
<tr>
<td>Page 38: [40] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 11:00:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>3.48</td>
</tr>
<tr>
<td>Page 38: [40] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 11:00:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>3.48</td>
</tr>
<tr>
<td>Page 38: [41] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 11:04:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>4.4</td>
</tr>
<tr>
<td>Page 38: [41] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 11:04:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>4.4</td>
</tr>
<tr>
<td>Page 38: [41] Deleted</td>
<td>Microsoft Office User</td>
<td>5/20/18 11:04:00 AM</td>
</tr>
<tr>
<td></td>
<td></td>
<td>4.4</td>
</tr>
<tr>
<td>Page 38: [42] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:55:00 AM</td>
</tr>
<tr>
<td>Left, Line spacing: single</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Page 38: [43] Formatted</td>
<td>Microsoft Office User</td>
<td>5/19/18 10:52:00 AM</td>
</tr>
<tr>
<td>Line spacing: single</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Missing Northerly flow  Missing Westerly flow

MAR: 4.31 (±4.74) m/s   AWS: 4.81 (±7.44) m/s