Frazil ice growth and production during katabatic wind events in the Ross Sea, Antarctica

Lisa De Pace¹, Madison Smith², Jim Thomson², Sharon Stammerjohn³, Steve Ackley⁴, and Brice Loose⁵

¹Department of Science, US Coast Guard Academy, New London CT
²Applied Physics Laboratory, University of Washington, Seattle WA
³Institute for Arctic and Alpine Research, University of Colorado at Boulder, Boulder CO
⁴University of Texas at San Antonio, San Antonio TX
⁵Graduate School of Oceanography, University of Rhode Island, Narragansett RI

Correspondence to: Brice Loose (bloose@uri.edu)

ABSTRACT: During katabatic wind events in the Terra Nova Bay and Ross Sea polynyas, wind speeds exceeded 20 m s⁻¹, air temperatures were below -25 °C, and the mixed layer extended as deep as 600 meters. Yet, upper ocean temperature and salinity profiles were not perfectly homogeneous, as would be expected with vigorous convective heat loss. Instead, the profiles revealed bulges of warm and salty water directly beneath the ocean surface and extending downwards tens of meters. Considering both the colder air above and colder water below, we suggest the increase in temperature and salinity reflects latent heat and salt release during unconsolidated frazil ice production within the upper water column. We use a simplified salt budget to analyze these anomalies to estimate in-situ frazil ice concentration between 332 x 10⁻³ and 24.4 x 10⁻³ kg m⁻³. Contemporaneous estimates of vertical mixing by turbulent kinetic energy dissipation reveal rapid convection in these unstable density profiles, and mixing lifetimes from 2 to 12 minutes. The corresponding median rate of ice production is 26 cm day⁻¹ and compares well with previous empirical and model estimates. Our individual estimates of ice production up to 378 cm day⁻¹ reveal the intensity of short-term ice production events during the windiest episodes of our occupation of Terra Nova Bay Polynya.
1. INTRODUCTION

Latent heat polynyas form in areas where prevailing winds or oceanic currents create divergence in the ice cover, leading to openings either surrounded by extensive pack ice or bounded by land on one side and pack ice on the other (coastal polynyas) (Armstrong, 1972; Park et al., 2018). The open water of polynyas is critical for air-sea heat exchange, since ice covered waters are one to two orders of magnitude better insulated (Fusco et al., 2009; Talley et al., 2011). A key feature of coastal or latent heat polynyas are katabatic winds (Figure 1), which originate as cold, dense air masses that form over the continental ice sheets of Antarctica. These air masses flow as sinking gravity currents, descending off the glaciated continent, or in the case of the Terra Nova Bay Polynya, through the Transantarctic mountain range. These flows are often funneled and strengthened by mountain-valley topography. The katabatic winds create and maintain latent heat polynyas. This research focuses on in-situ measurements taken from two coastal latent heat polynyas in the Ross Sea, the Terra Nova Bay polynya and the Ross Sea polynya.
Figure 1: Schematic of a latent heat or coastal polynya. The polynya is kept open from katabatic winds which drive ice advection, oceanic heat loss and frazil ice formation. Ice formation results in oceanic loss of latent heat to the atmosphere and brine rejection (Talley et al., 2011). Inset is a schematic of Frazil ice formation that depicts the release of latent heat of fusion and brine rejection as a frazil ice crystal is formed.

The extreme oceanic heat loss in polynyas can generate “supercooled” water, which is colder than the eutectic freezing point (Skogseth et al., 2009; Dmitrenk et al., 2010; Matsumura & Ohshima, 2015). Supercooled water is the precursor to ice nucleation and in-situ ice production. The first type of sea ice to appear are found as fine disc-shaped or dendritic crystals called frazil ice. These frazil ice crystals (Figure 1 inset) are about 1 to 4 millimeters in diameter and 1-100 micrometers in thickness (Heorton & Feltham, 2017; Martin, 1981; Ushio & Wakatsuchi, 1993; Wlchinsky et al., 2015). In polynyas, large net heat losses eventually lead to frazil ice production where katabatic winds and cold air temperatures transport of ice crystals away from the formation site near the ocean surface and into the water column. Both conditions are achieved in polynyas by (Coachman, 1966). Katabatic winds sustain the polynya by clearing frazil ice, forming pancake ice which piles up at the polynya edge to form a consolidated ice cover (Morales Maqueda et al., 2004; Ushio and Wakatsuchi, 1993).

Brine rejection (Cox & Weeks, 1983) and latent heat release during ice production, can lead to dense water formation. Over the Antarctic continental shelf, this process produces the precursor to Antarctic Bottom Water (AABW), a water mass known as High Salinity Shelf Water (HSSW) (Talley et al., 2011). In the case of the Ross Sea, the cold, dense HSSW formed on the shelf eventually becomes AABW off the shelf, the densest water in global circulation (Cosimo & Gordon, 1998; Jacobs, 2004; Martin, et al., 2007; Tamura et al.; 2007). Terra Nova Bay polynya produces especially dense HSSW, and produces approximately 1-1.5 Sv of HSSW annually (Buffoni et al., 2002; Orsi & Wiederwohl, 2009; Sansivero et al, 2017; Van Woert 1999a,b).

Given the importance of AABW to global thermohaline circulation, polynya ice production rates have been widely studied and modeled. Gallee (1997), Petrelli et al. (2008), Fusco et al. (2002), and Sansivero et al. (2017) used models to calculate polynya ice production.
rates on the order of tens of centimeters per day. Schick (2018) and Kurtz and Bromwich (1985) used heat fluxes to estimate polynya ice production rates, also on the order of tens of centimeters per day. However, quantitative estimation of polynya ice production is challenging due to the difficulty of obtaining direct measurements (Fusco et al., 2009; Tamura et al., 2007).

1.2 Motivation for this article

During a late autumn oceanographic expedition to the Ross Sea as part of the PIPERS (Polynyas, Ice Production and seasonal Evolution in the Ross Sea) project we measured CTD profiles in the Ross Sea coastal polynyas during katabatic wind events. Despite air temperatures that were well below freezing and strong winds frequently in excess of the katabatic threshold, these CTD profiles presented signatures of warmer water near the surface. The excess temperature was accompanied by similar signatures of saltier water. During this period, we also observed long wind rows of frazil ice. We hypothesized that the excess temperature was evidence of latent heat of fusion from frazil ice formation, and that the excess salinity was evidence of brine rejection from frazil ice formation. To test these hypotheses, we had to first evaluate the fidelity of these CTD measurements by comparing the shape and size of the profile anomalies with estimates of the CTD precision and stability, and by using supporting evidence of the atmospheric conditions that are thought to drive frazil ice formation (e.g. temperature and wind speed). This analysis is described below, followed by our estimates of frazil ice concentration using the temperature and salinity anomalies (§4). To better understand the importance of frazil formation, we computed the lifetime of these anomalies (§5), which in turn yielded frazil ice production rates (§6). Last, we discuss the implications for spatial variability of ice production and application for further polynya sea ice production estimates.

2. STUDY AREA AND DATA

2.1 The Terra Nova Bay Polynya and Ross Sea Polynya
The Ross Sea, a southern extension of the Pacific Ocean, abuts Antarctica along the Transantarctic Mountains and has three recurring latent heat polynyas: Ross Sea polynya (RSP), Terra Nova Bay polynya (TNBP), and McMurdo Sound polynya (MSP) (Martin et al., 2007). The RSP is Antarctica’s largest recurring polynya, the average area of the RSP is 27,000 km² but can grow as large as 50,000 km² depending on environmental conditions (Morales Maqueda et al., 2004; Park et al., 2018). It is located in the central and western Ross Sea to the east of Ross Island, adjacent to the Ross Ice Shelf (Figure 2), and typically extends the entire length of the Ross Ice Shelf (Martin et al., 2007; Morales Maqueda et al., 2004). TNBP is bounded to the south by the Drygalski ice tongue, which serves to control the polynya maximum size (Petrelli et al., 2008). TNBP and MSP, the smallest of the three polynyas, are both located in the western Ross Sea (Figure 2) (Petrelli et al., 2008). The area of TNBP, on average is 1300 km², but can extend up to 5000 km²; the oscillation period of TNBP broadening and contracting is 15-20 days (Bromwich & Kurtz, 1984). This paper focuses primarily on TNBP and secondarily on RSP, where our observations were taken.

During the autumn and winter season, Morales Maqueda et al., (2004) estimated TNBP cumulative ice production to be around 40-60 meters of ice, or approximately 10% of the annual sea ice production that occurs on the Ross Sea continental shelf. The RSP has a lower daily ice production rate, but produces three to six times as much as TNBP annually due to its much larger size (Petrelli et al., 2008).
Figure 2: Map of the Ross Sea and the Terra Nova Bay Polynya. a) Overview of the Ross Sea, Antarctica highlighting the locations of the three recurring polynyas: Ross Sea Polynya (RSP), Terra Nova Bay Polynya (TNBP), and McMurdo Sound Polynya (MSP). Map highlights the 2014 General Bathymetric Chart of the Oceans one-degree grid. b) Terra Nova Bay Polynya Insert as indicated by black box in panel a. MODIS image of TNBP with the 10 CTD stations with anomalies shown. Not included is CTD Station 40, the one station with an anomaly located in the RSP. (CTD Station 40 is represented on Figure 2a as the location of the Ross Sea Polynya.) Date of MODIS image is March 13, 2017; MODIS from during cruise dates could not be used due to the lack of daylight and high cloud cover.

2.2 PIPERS Expedition

We collected these data during late autumn, from April 11 to June 14, 2017 aboard the RVIB Nathaniel B. Palmer (NB Palmer, NBP17-04). More information about the research activities during the PIPERS expedition is available at http://www.utsa.edu/signl/pipers/index.html. Vertical profiles of Conductivity, Temperature, and Depth (CTD) were taken at 58 stations within the Ross Sea. For the purposes of this study, we
focus on the 13 stations (CTD 23-35) that occurred within the TNBP and 4 stations (CTD 37-40) within the RSP during katabatic wind events (Figure 2). In total, 11 of these 17 polynya stations will be selected for use in our analysis, as described in §3.1.

2.3 CTD measurements

The CTD profiles were carried out using a Seabird 911 CTD (SBE 911) attached to a 24 bottle CTD rosette, which is supported and maintained by the Antarctic Support Contract (ASC). The SBE 911 was deployed from the starboard Baltic Room. Each CTD cast contains both down and up cast profiles. In many instances, the upcast recorded a similar thermal and haline anomaly. However the 24 bottle CTD rosette package creates a large wake that disturbs the readings on the upcast, so only the down cast profiles are used.

The instrument resolution is important for this study, because the anomalous profiles were identified by comparing the near surface CTD measurements with other values within the same profiles. The reported initial accuracy for the SBE 911 is ± 0.0003 S m⁻¹, ± 0.001 °C, and 0.015% of the full-scale range of pressure for conductivity, temperature, and depth respectively. Independent of the accuracy stated above, the SBE 911 can resolve differences in conductivity, temperature, and pressure on the order of 0.00004 S m⁻¹, 0.0002 °C and 0.001% of the full range, respectively (SeaBird Scientific, 2018). The SBE 911 samples at 24 Hz with an e-folding time response of 0.05 seconds for conductivity and temperature. The time response for pressure is 0.015 seconds.

The SBE 911 data were post-processed with post-calibrations by Seabird, following standard protocol, and quality control parameters. Profiles were bin-averaged at two size intervals: one-meter depth bins and 0.1-meter depth bins, to compare whether bin averaging influenced the heat and salt budgets. Since we observed no difference between the budget calculations derived from one-meter vs 0.1-meter bins, the results using one-meter bins are presented in this publication. All thermodynamic properties of seawater were evaluated via the Gibbs Seawater toolbox, which uses the International Thermodynamic Equation Of Seawater – 2010 (TEOS-10).
2.4 Weather observations

Multiple katabatic wind events were observed within the TNBP and RSP during the PIPERS expedition. Weather observations from the NB Palmer meteorological suite during these periods were compared with observations from automatic weather stations Manuela, on Inexpressible Island, and Station Vito, on the Ross Ice Shelf (Figure 2a). Observations from all three were normalized to a height of 10 meters (Figure 3). The NB Palmer was in TNB from May 1 through May 13; during this period the hourly wind speed and air temperature data from Weather Station Manuela follow the same pattern, with shipboard observations from the NB Palmer observations being lower in intensity (lower wind speed, warmer temperatures) than Station Manuela. In contrast, the wind speed and air temperature from NB Palmer during its occupation in RSP (May 16-18) is compared to Station Vito. At Station Vito, the air temperature is colder, but the wind speed is less intense. Whereas at Station Manuela (TNBP) the winds are channelized and intensified through adjacent steep mountain valleys, the winds at Station Vito (RSP) are coming off the Ross Ice Shelf, resulting in lower wind speed.

During the CTD sampling in the TNBP there were 4 periods of intense katabatic wind events, with each event lasting for at least 24 hours or longer. During the CTD sampling in the RSP there was just one event of near katabatic winds lasting about 24 hours. During each wind event, the air temperature oscillated in a similar pattern and ranged from approximately -10 °C to -30 °C.
Figure 3: Weather observations from 01 May to 17 May 2017. a.) Wind speed from Station Manuela (blue line), Station Vito (purple line), NB Palmer (green line), and SWIFT (orange marker) deployments adjusted to 10 meters. The commonly used katabatic threshold of 17 m s$^{-1}$ is depicted as a “dotted red line”, as well as the date and start time of each CTD cast. b) Air temperature from Station Manuela, Station Vito, NB Palmer, and SWIFT deployments.

3. EVIDENCE OF FRAZIL ICE FORMATION

3.1 Selection of profiles

We used the following selection criteria to identify profiles from the two polynyas that appeared to be influenced by frazil ice formation: (1) a deep mixed layer extending several
hundred meters (Supplemental Figure 1), (2) in-situ temperature readings below the freezing point in the near-surface water (upper five meters), and (3) an anomalous bolus of warm and/or salty water within the top twenty meters of the profile (Figure 4 and 5). For context, all temperature profiles acquired during PIPERS (with the exception of one profile acquired well north of the Ross Sea continental shelf area at 60°S, 170°E) were plotted to show how polynya profiles compared to those outside of polynyas (Supplemental Figure 1).
Figure 4: Conservative Temperature profiles from CTD down casts from 11 stations showing temperature and/or salinity anomalies. Profiles (a-g) and (j-k) all show an anomalous temperature bulge. They also show supercooled water at the surface with the exceptions of (a) and (j). All of the plots (a-h) have an x-axis representing a 0.02 °C change. Profiles (a-j) are from TNBP, and (k) is from RSP.

Polynya temperature profiles were then evaluated over the top 50 meters of the water column using criteria 2 and 3. Nine TNBP profiles and one RSP profile exhibited the excess temperature anomalies over the top 10-20 m and near-surface temperatures close to the freezing point (Figure 4). Excess salinity anomalies (Figure 5) were observed at the same stations with two exceptions: Station 26 had a measurable temperature anomaly (Figure 4b) but no discernible salinity anomaly (Figure 5b), and Station 33 had a measurable salinity anomaly (Figure 5h) but no discernible temperature anomaly (Figure 4h). The stations of interest are listed in Table 1.
Figure 5: Absolute Salinity profiles from CTD down casts from 11 stations showing temperature and/or salinity anomalies. Profiles (a) and (c-k) show an anomalous salinity bulge in the top 10-20 meters. Two profiles (c and g) show salinity anomalies extending below 40 meters, so the plot was extended down to 80 meters to best highlight those. All of the plots (a-k) have an absolute salinity range of 0.03 g kg⁻¹.

3.2 Evaluating the uncertainty in the temperature and salinity anomalies

To evaluate the uncertainty associated with the temperature and salinity anomalies at each of the polynya stations, we compared each anomaly to the initial accuracy of the SBE 911 temperature and conductivity sensors: ± 0.001 °C and ± 0.0003 S m⁻¹, or ±0.00170 g kg⁻¹ when converted to absolute salinity. To quantify the maximum amount of the temperature anomaly, the baseline excursion, ΔT, was calculated throughout the anomaly ΔT = T_{obs} - T_{b}, where T_{obs} is the in-situ conservative temperature and T_{b} is the in-situ baseline, which is extrapolated from the far field conservative temperature within the well-mixed layer below the anomaly. Taking the single largest baseline excursion from each of the 11 anomalous CTD profiles and averaging them, we compute an average baseline excursion of 0.0064 °C. While this is a small change in the temperature, it is 32 times larger than the stated precision of the SBE 911 (0.0002 °C). The same approach applied to the salinity anomalies yielded an average baseline of 0.0041 S m⁻¹ (or 0.0058 g kg⁻¹ for absolute salinity), which is 100 times larger than the instrument precision (0.00004 S m⁻¹). Table 1 lists the maximum temperature and salinity anomalies for each CTD station.

One concern was that frazil ice crystals could interfere with the conductivity sensor. It is conceivable that ice crystals smaller than 5 mm can be sucked into the conductivity cell, creating spikes in the raw conductance data. Additionally, frazil crystals smaller than 100 µm are theoretically small enough to float between the electrodes and thereby decrease the resistance/conductance that is reported by the instrument (Skogseth & Smedsrud, 2009). To test for ice crystal interference, the raw (unfiltered with no bin averaging) absolute salinity profile
was plotted using raw conductivity compared with the 1-meter binned data for the 11 anomalous CTD Stations (Supplemental Figure 2). The raw data showed varying levels of noise as well as some spikes or excursions to lower levels of conductance; these spikes may have been due to ice crystal interference. However, the bin-averaged data do not appear to be biased or otherwise influenced by the spikes, which tend to fall symmetrically around a baseline. This was demonstrated by bin-averaging over different depth intervals as described in §2.4, Considering the consistency of the temperature and salinity measurements within and below the anomalies, and the repeated observation of anomalies at 11 CTD stations, we infer that the observed anomalies are not an instrumental aberration.

3.3 Camera observations of frazil ice formation

During PIPERS an EISCam (Evaluative Imagery Support Camera, version 2) was operating in time lapse mode, recording photos of the ocean surface from the bridge of the ship every 10 minutes (for more information on the EISCam see Weissling et al, 2009). The images from the time in TNBP and RSP reveal long streaks and large aggregations of frazil ice. A selection of photos from TNBP were captured (Figure 6). The winds were strong enough at all times to generate wave fields and advect frazil ice, thus creating downstream frazil streaks, and eventually pancake ice in most situations. Smaller frazil streaks and a curtain of frazil ice below the frazil streak were also visible.
Figure 6: Images from NB Palmer as EISCam (Evaluative Imagery Support Camera) version 2. White areas in the water are loosely consolidated frazil ice crystals being actively formed during a katabatic wind event. Image (d) was brightened to allow for better contrast.

3.4 Conditions for frazil ice formation during lab experiments

Ushio and Wakatsuchi (1993) conducted laboratory experiments to reproduce the conditions observed in polynyas. They exposed their tank, measuring 2-m length, 0.4-m width and 0.6-m depth to air temperatures at -10 °C and wind speeds of 6 m s⁻¹. They observed supercooling in the range of 0.1 to 0.2 °C at the water surface and found that after 20 minutes the rate of super-cooling slowed due to the release of latent heat, coinciding with visually observed frazil ice formation. Simultaneously with the formation of frazil ice crystals, they observed an increase in salinity from the brine rejection. After ten minutes of ice formation, the temperature of the frazil ice layer was 0.07 °C warmer and the layer was 0.5 to 1.0% saltier (Ushio and Wakatsuchi, 1993).
In this study, we found the frazil ice layer to be on average 0.0064 °C warmer than the underlying water. Similarly, the salinity anomaly was on average 0.0058 g kg\(^{-1}\) saltier, which equates to 0.017% saltier than the water below. While our anomalies were significantly smaller than those observed in the lab tank by Ushio and Wakatsuchi (1993), the same trend of super-cooling, followed by frazil ice formation and the appearance of a salinity anomaly was observed during PIPERS. However, the forcing conditions and spatial constraints of the tank experiment likely explain why there are discrepancies between the magnitudes of the temperature and salinity anomalies observed in the lab versus in the field.

3.5 Temperature and salinity profiles in the presence of platelet ice formation

The mechanism for supercooling under ice shelves occurs via a different process than in polynyas, but with similar impact on the water column structure. In polynyas, katabatic winds and sub-freezing air temperatures create supercooled water near the surface, which drove frazil ice formation. As plumes of Ice Shelf Water approached the surface, the pressure change led to the formation of supercooled water and frazil ice formation (Jones & Wells, 2018). Robinson et al (2017) investigated ice formation through this process under the McMurdo Sound Ice Shelf. As the frazil crystals continue to grow, they maintained their geometry and formed platelet ice. Robinson et al. (2017) found an increase in salinity from brine rejection and an increase in temperature from latent heat released at the depth of ice formation. Though the mechanism for supercooling differs, these vertical trends in temperature and salinity nonetheless are similar to our results.

3.6. The anomalous profiles from TNBP and RSP appear to trace active frazil ice formation

Throughout Sections 2 and 3, we have documented that the anomalous profiles from TNBP and RSP appear to trace frazil ice formation. In §2.4, the strong winds and sub-zero air temperatures supported both ice formation and advection. In §3.1 and §3.2, we showed that the CTD profiles in both temperature and salinity are reproducible and large enough to be
distinguished from the instrument noise. In §3.3 the coincident EISCam measurements reveal significant accumulation of frazil ice crystals on the ocean surface during the time the NB Palmer was in TNBP and RSP. In §3.4 and §3.5, we note the commonalities between the PIPERS polynya profiles and frazil ice formation during platelet ice formation and during laboratory experiments of frazil ice formation. Given the co-occurrence of strong winds, cold air temperatures, sub-zero water temperature, we find no simpler explanation for the apparent warmer, saltier water near the surface in our 11 CTD profiles from TNBP and RSP. Considering the similarity in conditions during the lab experiments and during in-situ platelet ice formation, we conclude that our 11 profiles reflect measurable frazil ice formation in the TNBP and RSP.

4.0 ESTIMATION OF FRAZIL ICE CONCENTRATION USING CTD PROFILES

Having identified a collection of CTD profiles that trace frazil ice formation, we want to know how much frazil ice formation can be inferred from these T and S profiles? Can we attribute a large portion of polynya ice formation to this early stage of ice growth, or is the growth of pack ice at the polynya edge the dominant process? To estimate ice formation, the inventories of heat and salt from each profile can provide independent estimates of frazil ice concentration. To simplify the inventory computations, we neglect the horizontal advection of heat and salt; this is akin to assuming that lateral variations are not important because the neighboring water parcels are also experiencing the same intense vertical gradients in heat and salt. We first describe the computation using temperature in § 4.1 and the computation using salinity in § 4.2.

4.1 Estimation of frazil ice concentration using temperature anomalies

We used the temperature profiles to compute the “excess” heat inside the anomalies. Utilizing the latent heat of fusion as a proxy for frazil ice production we estimated the amount of frazil ice that must be formed in order to create observed anomalies. For each station, we first estimated the enthalpy inside the temperature anomaly (Talley et al, 2011) as follows. Within each CTD bin, we estimated the excess temperature as \( \Delta T = T_{\text{obs}} - T_b \), where \( T_{\text{obs}} \) is the in-situ
conservative temperature and $T_b$ is the in-situ baseline or far field conservative temperature. The excess over the baseline is graphically represented in Figure 7a. Because we lacked multiple profiles at the same location, we were not able to observe the time evolution of these anomalies. Consequently, $T_b$ represents our best inference of the temperature of the water column prior to the onset of ice formation; it is highlighted in Figure 7a with the dashed line. We established $T_b$ by looking for a near constant value of temperature in the profile directly below the temperature bulge. In most cases the temperature trend was nearly linear and close to the freezing point. After selecting the starting location, the conservative temperature was averaged over 10 meters (10 values from the 1-m binned data) to eliminate slight variations and any selection bias.

Figure 7: Conservative temperature, absolute salinity, and potential density anomaly for TNBP CTD Station 35, May 10, 2017. a) Conservative temperature profile showing the temperature anomaly, the selected baseline temperature (dashed line) and the integrated excess temperature (shaded area). b) Absolute salinity profile showing the salinity anomaly, the selected baseline
salinity (dashed line), and integrated excess salinity (shaded area). c) Potential density anomaly showing the selected baseline density (dashed) and the excess density instability (shaded).

To find the excess heat (\( Q_{\text{excess}}^{\text{total}} \)) represented in the total thermal anomaly, we computed the vertical integral of heat per unit area from the surface (\( z=0 \)) to the bottom of the anomaly (\( z=z_f \)):

\[
Q_{\text{excess}}^{\text{total}} = \int_{z=0}^{z=z_f} \varrho \cdot C_p \cdot \Delta T \, dz
\]  

(1)

Here \( \varrho \) = density of seawater, \( z \) = depth range of the anomaly, and \( C_p \) = the specific heat capacity. The concentration of frazil ice is estimated by applying the latent heat of formation (\( L_f = 330 \, \text{kJ kg}^{-1} \)) as a conversion factor to \( Q_{\text{excess}}^{\text{total}} \):

\[
\text{Conc}_{\text{ice}}^{\text{temp}} = \frac{Q_{\text{excess}}^{\text{total}}}{L_f \cdot z_f}
\]  

(2)

Where \( z_f \) is the depth of the temperature anomaly in meters. The concentration of ice derived represents the total concentration of ice, in kg m\(^{-3}\). A more detailed explanation of equations 1 and 2 is contained in Supplemental 1. The mass concentration of ice derived from the temperature anomaly for each station is listed in Table 1.

4.2 Estimation of frazil ice concentration using salinity anomalies

The mass of salt within the salinity anomaly was used to estimate ice formation. Assuming that frazil ice crystals do not retain any brine and assuming there is no evaporation, the salinity anomaly is directly proportional to the ice formed. By using the conservation of mass equations for water and salt, the mass of frazil ice can be estimated by comparing the excess salt (measured as salinity) with the amount of salt initially present in the profile. The conservation of mass equations used and subsequent derivations are in Supplemental 2. The salinity anomaly (\( \Delta S \)) above the baseline salinity (\( S_b \)) is \( \Delta S = S_{\text{obs}} - S_b \), and is shown in Figure 7b. The initial value of salinity (\( S_b \)) was established by observing the trend in the salinity profile directly below the haline bulge; in most cases the salinity trend was nearly linear beneath the bulge,
however in general the salinity profiles were less homogeneous than the temperature profiles.

After selecting the starting location from below the anomaly, the absolute salinity was averaged over the next 10 meters to establish a baseline salinity.

To find the total mass of frazil ice (\(\text{Mass}_{\text{Ice}}^S\), kg m\(^{-2}\)) in the water column, the integral of each component of the salt ratio is taken over the depth range of the anomaly. This integral is multiplied by the total mass of water per area (\(\text{Mass}_{\text{Water}}^\text{Total}\), kg m\(^{-2}\)) initially in the depth range of the anomaly. The concentration of ice (\(\text{Conc}_{\text{Ice}}^\text{salt}\), kg m\(^{-3}\)) can be found by dividing the mass of frazil ice by the depth of the salinity anomaly \((z_f)\). The resulting estimates of ice concentration are listed in Table 1.

\[
\text{Mass}_{\text{Ice}}^S = \text{Mass}_{\text{Water}}^\text{Total} \int_{z=0}^{z=H} \frac{\Delta S \, dz}{S_{\text{salt}}} \quad (3)
\]

\[
\text{Mass}_{\text{Water}}^\text{Total} = Q_b \int_{z=0}^{z=H} dz \quad (4)
\]

\[
\text{Conc}_{\text{Ice}}^\text{salt} = \frac{\text{Mass}_{\text{Ice}}^S}{z_f} \quad (5)
\]

A more detailed explanation of equations 3, 4, and 5 is contained in Supplemental 3.

4.3 Summary of the frazil ice estimates

The derived ice concentrations are listed in Table 1. The inventories of salt produced in-situ frazil ice concentrations from \(24 \times 10^3\) kg m\(^{-3}\) to \(332 \times 10^3\) kg m\(^{-3}\). However, it is noteworthy that the estimates of frazil ice concentration from salt inventories are anywhere from 2 to 9 times greater than the estimates from heat inventories. The difference is likely produced by unquantified heat loss to the atmosphere. The influence of sensible and long wave heat exchanges produces an atmospheric loss term in the heat inventory, which has no corresponding influence on the salt inventory. Therefore, we suggest that derived ice concentrations from the heat anomalies underestimated frazil ice concentration in comparison to the salt inventory.

We also note the salinity calculation does not account for evaporation. However, evaporation could have contributed to excess salinity while simultaneously decreasing the...
Mathiot et al. (2012) found that evaporation was secondary to ice production and contributed 4% to salt flux. In the TNBP, the Palmer meteorological tower revealed high relative humidity (on average 78.3%), so the effects of evaporation on salinity were likely therefore negligible. The effects of evaporation would reduce the mass of ice derived from the salinity anomaly.

Table 1: CTD Stations with temperature and salinity anomalies (See Figures 4-5), showing maximum values of the temperature anomaly, depth range of the temperature anomaly, concentration of ice derived from the temperature anomaly (§4.1), as well as the maximum value of the salinity anomaly, depth range of salinity anomaly, and concentration of ice derived from the salinity anomaly (§4.2).
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</tbody>
</table>

*Station 26 did not have a measurable salinity anomaly but was included due to the clarity of the temperature anomaly. Conversely, **Station 33 did not have a measurable temperature anomaly but was included due to the clarity of the salinity anomaly.

**5.0 ESTIMATION OF TIME SCALE OF ICE PRODUCTION**

How should we interpret the lifetime of these T and S anomalies? Are they short-lived in the absence of forcing, or do they represent an accumulation over some longer ice formation period? One possibility is that the anomalies begin to form at the onset of the katabatic wind event, implying that the time required to accumulate the observed heat and salt anomalies is similar to that of a katabatic wind event (e.g. 12-48 hours). This, in turn would suggest that the estimated frazil ice production occurred over the lifetime of the katabatic wind event. Another interpretation is that the observed anomalies reflect the near-instantaneous production of frazil ice. In this scenario, heat and salt are simultaneously produced and actively mixed away into the far field. In this case, the observed temperature and salinity anomalies reflect the net difference between production and mixing. One way to address the question of lifetime is to ask “if ice production stopped, how long would it take for the heat and salt anomalies to dissipate?” The
answer depends on how vigorously the water column is mixing. In this section, we examine the mixing rate. However, we can first get some indication of the timescale by the density profiles.

5.1 Apparent instabilities in the density profiles

The computed density profiles reveal an unstable water column for all but one of our eleven stations (Figure 8). These suggest that buoyancy production from excess heat did not effectively offset the buoyancy loss from excess salt within each anomaly. It is not common to directly observe water column instability without the aid of microstructure or other instruments designed for measuring turbulence.
Figure 8: Potential density anomalies for all 11 stations with evidence of active frazil ice formation. The integrated excess density and assumed baseline density are depicted to highlight the instability. Note that Station 26 (b) does not present a density anomaly because it does not have a salinity anomaly. In the absence of excess salinity, the temperature anomaly created instead an area of less dense water (i.e., a stable anomaly).

We suggest that an instability in the water column that persists long enough to be measured in a CTD profile, must be the result of a continuous buoyancy loss that is created at a rate faster than it can be eroded by mixing. In other words, the katabatic winds appeared to dynamically maintain these unstable profiles. Continuous ice production leads to the production of observed heat and salt excesses at a rate that exceeds the mixing rate. If the unstable profiles reflect a process of continuous ice production, then the inventory of ice that we infer from our simple heat and salt budgets must reflect ice production during a relatively short period of time, defined by the time it would take to mix the anomalies away, once the wind-driven dynamics and ice production stopped.

Similarly, Robinson et al (2017) found that brine rejection from platelet ice formation (§3.5) also leads to dense water formation and a static instability. Frazil ice formation from continually supplied ISW created a stationary instability, which was observable before being mixed by convection to the underlying homogeneous water column that extended to 200 meters. Similarly, the katabatic winds and cold air temperatures continually supply supercooled water to the polynya supporting the instability.

5.2 Lifetime of the salinity anomalies from Monin-Obukhov length scale

Turbulence theory suggests the largest eddies control the rate of turbulence dissipation (Cushman-Rosin, 2019). A characteristic timescale, $t$, can be approximated by relating the largest eddy size and the rate of turbulent kinetic energy dissipation ($\varepsilon$, Cushman-Rosin, 2019).

$$
t \approx \frac{d}{\sqrt{\varepsilon \kappa}} \approx \left( \frac{\varepsilon}{\kappa} \right)^{\frac{1}{2}}
$$

(6)
Here, $d$ is the characteristic length of the largest eddy and $\varepsilon$ is the turbulent kinetic energy dissipation rate. In this section we discuss and select the best length scale for an environment dominated by buoyancy and wind shear. We use observed parameters to estimate the terms in equation (6).

The dimension, $d$, of the largest eddy in a vigorously mixing water column could be equivalent to the scale of the domain (in this case, the mixed layer depth) which was up to 600 m in some of the PIPERS profiles (Table 2). However, a homogenous mixed-layer does not necessarily imply active mixing throughout the layer (Lombardo and Gregg, 1989). Instead, the characteristic length scale in an environment driven by both buoyancy and wind shear is typically the Monin-Obukhov length ($L_{M-O}$) (Monin & Obukhov, 1954). When $L_{M-O}$ is small and positive, buoyant forces are dominant and when $L_{M-O}$ is large and positive, wind shear forces are dominant (Lombardo & Gregg, 1989). While the $L_{M-O}$ can be expressed using several different estimates of shear and buoyancy, we focus on the salt-driven buoyancy flux, because those anomalies come closest to capturing the process of frazil ice production (see §4.3 for more detail).

$$L_{M-O} = - \frac{u_*^2}{k \beta g w \Delta S}$$  \hspace{1cm} (7)

where $u_*$ is the wind-driven friction velocity at the water surface, $g$ is gravitational acceleration, $w$ is the water vertical velocity, $\Delta S$ is the salt flux, $\beta$ is the coefficient of haline contraction, and $k$ is the von Karman constant. A more detailed explanation, along with the specific values are listed in Supplemental 4.

Wind-driven friction velocity is estimated using the NB Palmer wind speed ($U_{palmer}$) record from a masthead height of $z_{palmer} = 24$ m, adjusted to a 10 meter reference ($U_{10}$) by assuming a logarithmic profile (Manwell et al., 2010).

$$U_{10} = U_{palmer} \cdot \frac{\ln(\text{zar})}{\ln(z_{palmer}/10)}$$ \hspace{1cm} (8)
Roughness class 0 was used in the calculation and has a roughness length of 0.0002 m. These values are used to estimate the wind stress, $\tau$ as,

$$\tau = C_D \rho_{air} U_{10}^2$$

(9)

where $\rho_{air}$ represents the density of air, with a value of 1.3406 kg m$^{-3}$ calculated using averages from NB Palmer air temperature (-18.73 °C), air pressure (979.4 mbars) and relative humidity (78.3%). $C_D$ represents a dimensionless drag coefficient and was calculated as $1.525 \times 10^{-3}$, using COARE 3 code, modified to incorporate wave height and speed (Fairall et al, 2003). The average weather data from NB Palmer was paired with the wave height and wave period from the SWIFT deployment (defined below) on 04 May to find $C_D$. A more detailed explanation and the specific values are listed in Supplemental 5.

We determined the aqueous friction velocity ($u_*$) at the air-sea interface using:

$$u_* = \sqrt{\frac{\tau}{\rho_{sw}}}$$

(10)

We used a SWIFT (Surface Wave Instrument Float with Tracking) buoy to provide estimates of turbulent kinetic energy dissipation and vertical velocity. (Thomson et al., 2016; Zippel & Thomson, 2016). SWIFT deployments occurred during the period of CTD observations, as shown in the timeline of events (Supplemental Figure 3). The SWIFT deployments do not always coincide in time and space with the CTD profiles. For the vertical velocity estimation we identified the May 04 and May 09 SWIFT deployments as most relevant to CTD stations analyzed here based on similarity in wind speeds. The average wind speed at all the CTD stations with anomalies was 10.2 m s$^{-1}$. For the May 4 SWIFT deployment, the wind speed was 9.36 m s$^{-1}$. CTD Station 32, more than two standard deviations from the average, experienced the most intense winds of the CTD stations at 18.9 m s$^{-1}$. For CTD Station 32, the May 9 SWIFT deployment was used, which had a wind speed of 20.05 m s$^{-1}$. For May 04 and May 09, the average vertical velocity ($w$) was measured in the upper meter of the column. May 04 had an average value of $w = 0.015$ m s$^{-1}$. May 09 had an average value of $w = 0.025$ m s$^{-1}$.
Thomson et al., 2016 & Zippel & Thomson, 2016 for details on how these measurements are made.

The TKE dissipation rates are expected to vary with wind speed, wave height, ice thickness and concentration (Smith & Thomson, 2019). Wind stress ($\tau_{\text{wind}}$) is the source of momentum to the upper ocean, but this is modulated by scaling parameter ($c_e$, Smith & Thomson, 2019). If the input of TKE is in balance with the TKE dissipation rate over an active depth layer, the following expression can be applied:

$$c_e \times \tau_{\text{wind}} \propto \int \varepsilon(z) \, dz$$  \hspace{1cm} (11)

where the density of water ($\rho$) is assumed to be 1027 kg m$^{-3}$ for all stations. The scaling parameter incorporates both wave and ice conditions; more ice produces more efficient wind energy transfer, while simultaneously damping surface waves, with the effective transfer velocity in ice, based on the assumption that local wind input and dissipation are balanced (Smith & Thompson, 2019).

$$c_e = a \left( \frac{z_{\text{ic}}}{H_s} \right)^b$$  \hspace{1cm} (12)

Here, $A$ is the fractional coverage of ice, with a maximum value of 1, $z_{\text{ic}}$ is the thickness of ice, and $H_s$ is the significant wave height. Using Antarctic Sea ice Processes and Climate or ASPeCt visual ice observations (www.aspect.aq) from NB Palmer, the fractional ice cover and thickness of ice were found at the hour closest to both SWIFT deployments and CTD profiles (Knuth & Ackley, 2006; Ozsoy-Cicek et al., 2008; Worby et al., 2008). The significant wave height for each SWIFT deployment was used. We lacked time series data for $H_s$ during the time of CTD casts, so the average value from May 04 of 0.58 m was used for all the CTD profiles. To get the most robust data set possible, in total, 13 vertical SWIFT profiles from May 2, May 4, and May 9 were used to evaluate equation 12 over an active depth range of 0.62 meters.

Using the estimates of $c_e$, $\tau$, and $\varepsilon$ from the SWIFT, we parameterized the relationship between wind stress and $\varepsilon$ that is reflected in equation (11). A log-linear fit ($y = 10^{(1.4572 \times \log_{10}(\varepsilon) +0.2299)}$),
was then applied to NB Palmer wind stress data to derive turbulent kinetic
dissipation estimates that coincided with the ambient wind conditions during each CTD station
(Table 2).

Figure 9: Logarithmic linear fit of the input flux of TKE into the ocean versus the TKE
dissipation rate over the active depth range.

Following estimation of the environmental parameters, Equation 7 can now be used to
estimate $L_{M-O}$. For these calculations a value of 0.41 was used for the von Karman constant, $k$.
Haline contraction, $\beta$, was calculated from Gibbs Seawater toolbox and averaged over the depth
range of the anomaly. The excess salt, $\Delta S$, was found using the average value of $\Delta S$ for each profile anomaly. The values of $L_{M-O}$ range from 6 m to 330 m (Table 2). In general, $L_{M-O}$ was greater than the length of the salinity anomaly but smaller than the mixed layer depth. Using $L_{M-O}$ and the estimates of $\varepsilon$, the characteristic lifetime of the salinity anomalies ranged from 2 to 12 minutes, but most values cluster near the average of 9 min. The average timescale is similar to the frazil ice lifetime found in Michel (1967). These lifetimes suggest that frazil ice production and the observed density instabilities relax to a neutral profile within ten minutes of a diminution in wind forcing.

### 6.0 RATE OF FRAZIL ICE PRODUCTION

We can extend the analysis of anomaly lifetime to estimate a frazil ice production rate by invoking the prior assumption of steady state TKE forcing and dissipation. In this case, the mass of ice reflected by the salinity anomaly ($\text{Conc}_{\text{ice}}^{\text{salt}}$, in kg m$^{-3}$) was produced during the time interval corresponding to the mixing lifetime ($t$) that was determined from TKE dissipation in §5.2.

$$\text{Production rate} = \frac{\text{Conc}_{\text{ice}}^{\text{salt}} \cdot z_s}{t \cdot \rho_{\text{ice}}}$$

Here, $\rho_{\text{ice}} = 920$ kg m$^{-3}$, $t$=lifetime, in days, and $z_s$ = the depth of the salinity anomaly (m). The results are summarized in Table 2. A more detailed explanation and the specific values are listed in Supplemental 6.

#### 6.1 Variability in the frazil ice production rate

The ten estimates of frazil ice production rate, expressed as ice thickness per unit time, ranged from 7 to 378 cm day$^{-1}$. These frazil ice production rates show some spatial trends across the Terra Nova Bay polynya that correspond with variable environmental conditions in different sectors of the polynya. As shown in Figure 10, a longitudinal gradient emerges along the axis of the TNBP when looking at a subsection of stations under similar wind conditions Station 30
(U_{10}=11.50 \text{ m s}^{-1}), \text{ Station } 27 (U_{10}=10.68 \text{ m s}^{-1}), \text{ and Station } 25 (U_{10}=11.77 \text{ m s}^{-1}). \text{ Beginning upstream near the Nansen Ice shelf (Station 30) and moving downstream along the predominant wind direction toward the northeast, the ice production rate decreases. The upstream production rate is } 69.38 \text{ cm day}^{-1} \text{ followed by midstream values of } 28.43 \text{ cm day}^{-1}, \text{ and lastly downstream values of } 9.83 \text{ cm day}^{-1}. \text{ This pattern is similar to the pattern modeled by Gallee (1997). The production rate at Station 35, was significantly higher than that at all other stations, but this large excess is reflected in both the heat and salt anomalies. The salt inventory at station 35 is a factor of 2.6 greater than the nearest station (Station 34), and profiles 34 and 35 were separated in time by less than 5 hours. This other variations in ice production rate may reflect real variability brought on by submesoscale fronts, eddies and other flow structures that are not easily captured by coarse sampling. }

We used the student t-distribution to derive confidence intervals for TKE dissipation rate at each CTD station was used to bound the range of ice production rates, which are reported in Table 2. Uncertainty in the heat and salt inventories were not included in the uncertainty estimates, because we observed negligible difference in the inventory while testing the inventory for effects associated with bin averaging bin averaging of the CTD profiles (Section 2.3). Another small source of error arises from the neglect of evaporation. To quantify the amount of error introduced by that assumption, we used the bulk aerodynamic formula for latent heat flux and found the effects of evaporation across the CTD stations to be 1.8% [0.07-3.45%] (Zhang, 1997). This error due to the effects of evaporation found are similar to Mathiot et al (2012). On average, the lower limit of ice production was 30% below the estimate and the upper limit was some 44% larger than the estimated production.

Table 2: Summary of mass of ice derived from salinity, lifetime, and production rates.

<table>
<thead>
<tr>
<th>Station</th>
<th>Conc_{ice}^S (kg m^{-3})</th>
<th>z_s (m)</th>
<th>L_{M-O} (m)</th>
<th>TKE diss. ( \varepsilon ) (m^2 s^{-3})</th>
<th>Est MLD (m)</th>
<th>Lifetime (min)</th>
<th>Production rate (cm day^{-1})</th>
<th>Production rate 95% CI (cm day^{-1})</th>
</tr>
</thead>
</table>

32
<table>
<thead>
<tr>
<th>Station</th>
<th>Salinity Anomaly</th>
<th>Temperature Anomaly</th>
<th>Temperature Anomaly uncertainty</th>
</tr>
</thead>
<tbody>
<tr>
<td>25</td>
<td>77.76 x 10⁻³</td>
<td>13.4</td>
<td>9.648 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>26*</td>
<td>--</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>27</td>
<td>48.01 x 10⁻³</td>
<td>41.2</td>
<td>8.188 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>28</td>
<td>24.37 x 10⁻³</td>
<td>17.5</td>
<td>1.622 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>29</td>
<td>58.55 x 10⁻³</td>
<td>21.6</td>
<td>5.375 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>30</td>
<td>116.63 x 10⁻³</td>
<td>36</td>
<td>3.771 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>32</td>
<td>121.90 x 10⁻³</td>
<td>47</td>
<td>3.466 x 10⁻⁶⁴</td>
</tr>
<tr>
<td>33</td>
<td>32.38 x 10⁻³</td>
<td>23.7</td>
<td>2.844 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>34</td>
<td>80.29 x 10⁻³</td>
<td>19.6</td>
<td>6.397 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>35</td>
<td>332.16 x 10⁻³</td>
<td>14.4</td>
<td>2.343 x 10⁻⁶⁵</td>
</tr>
<tr>
<td>40</td>
<td>48.84 x 10⁻³</td>
<td>18.6</td>
<td>9.603 x 10⁻⁶⁵</td>
</tr>
</tbody>
</table>

*Station 26 did not have a measurable salinity anomaly but was included due to the clarity of the temperature anomaly.
6.2 Comparison to prior model and field estimates of ice production

Calculated production rates from PIPERS ranged from 7 to 378 cm day\(^{-1}\) (Figure 10). The median ice production rate, 26.31 cm day\(^{-1}\), is similar to Schick (2018), who estimated an average ice production rate, 16.8 cm day\(^{-1}\), for the month of May, (calculated using atmospheric heat fluxes). Our median is also similar to Kurtz and Bromwich (1985), who also used a heat budget to estimate an average ice production rate of 30 cm day\(^{-1}\) for the month of May. All of these estimates are smaller than the winter average from Sansiviero et al (2017) of 48.08 cm day\(^{-1}\) using a sea-ice model. Petrelli, Bindoff, & Bergamasco (2008) modeled a wintertime maximum production rates of 26.4 cm day\(^{-1}\) using a coupled atmospheric-sea ice model. Fusco et al (2002) applied a model for latent heat polynyas and modeled production rate at 85 cm day\(^{-1}\) for 1993 and 72 cm day\(^{-1}\) for 1994.

The spatial trend we observed somewhat mimics the model 3D model of TNBP from Gallee (1997). During a four-day simulation, Gallee found highest ice production rates near the coast (e.g. our Station 35) of 50 cm day\(^{-1}\), and decreasing production to 0 cm day\(^{-1}\) downstream and at the outer boundaries, further west than PIPERS Station 33 (Figure 10). While some of the ice production rates derived from PIPERS CTD profiles exceed prior results, we attribute that excess to the relatively short time scale of these ice production “snapshots”. These estimates integrate over minutes to tens of minutes, instead of days to months, therefore they are more likely to capture the high frequency variability in this ephemeral process. As the katabatic winds oscillate, the polynyas enter periods of slower ice production, driving average rates down.
Figure 10: TNBP map of ice production rates. Map of TNBP CTD stations with anomalies and ice production rates. The CTD station number is listed in black and circled. Listed next to the station is the respective ice production rate in cm day$^{-1}$. The production rates are colored by wind speed: Green indicates wind speeds less than 10 m s$^{-1}$ (Stations 28, 29, 33, 34, 35), Orange indicates wind speeds between 10 and 15 m s$^{-1}$ (Stations 25, 27, 30), and Red indicated wind speeds over 15 m s$^{-1}$ (Station 32).

7. CONCLUSIONS
Polynyas have been regarded as ice production factories with a wide range of modeled production rates. During a late autumn oceanographic expedition to the Ross Sea, PIPERS acquired CTD profiles in the ocean during strong katabatic wind events in both the Terra Nova Bay polynya and the Ross Sea polynya. In those profiles we found near surface temperature and salinity anomalies, which provided a new method for quantifying ice production rates in-situ. Salinity and temperature anomalies observed at 11 CTD stations indicated frazil ice formation and were used to estimate polynya ice production. Our estimated frazil ice production rates varied from 7 to 378 cm day$^{-1}$. The wide range is likely capturing frazil ice production on very short timescales (minutes). We note that the robustness of these estimates could be improved by collecting consecutive CTD casts at the same location.

The polynyas in the Ross Sea show high ice production rates and are significant contributors to Antarctic Bottom Water formation. Since 2015, sea ice extent around Antarctica has decreased, with 2017 being an abnormally low year (Supplemental Figure 5; Fetterer et al, 2017). One of the goals of PIPERS was to understand if sea ice extent in the Ross Sea was controlled primarily by ice production at the coast. If true, the decreased ice extent in recent years may be related to changes in ice production in the polynyas. To further address these questions, our estimates of polynya ice production can be paired with other ice products derived from remote sensing, such as ice thickness from airborne and satellite lidar and ice area from radar and passive microwave to better address the observed year-to-year changes. A decrease in ice production rate correlates to freshening of Antarctic bottom water which would have global impacts.
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Preprint. Discussion started: 30 October 2019
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8. ACKNOWLEDGEMENTS

The authors appreciate the support of the National Science Foundation through NSF Award Nos 1744562 (B. Loose, L. de Pace, URI); 134717 (S.F. Ackley, UTSA); 1341513 (E. Maksym, WHOI); 1341725 (P.Guest, NPS); 1341606 (S. Stammerjohn and J. Cassano, U Colo). The authors appreciate the support of the University of Wisconsin-Madison Automatic Weather Station Program for the data set, data display, and information.

9. DATA AVAILABILITY

The data used in this publication are publicly available from the US Antarctic Program Data Center http://www.usap-dc.org/view/dataset/601192

10. AUTHOR CONTRIBUTIONS

LD prepared the manuscript including all analysis. MS and JT provided SWIFT data and guidance for upper ocean turbulence analysis. SS prepared and processed the PIPERS CTD data and provided water mass insights during manuscript preparation; SA lead the PIPERS expedition
and supported ice interpretations. BL participated in PIPERS expedition, inferred possibility of frazil ice growth and advised LD during manuscript preparation.

11. COMPETING INTERESTS

The authors declare that they have no conflict of interest."