Weakening of Cold Halocline Layer Exposes Sea Ice to Oceanic Heat in the Eastern Arctic Ocean

IGOR V. POLYAKOV,a,b,c TOM P. RIPPETH,d ILKER FER,e MATTHEW B. ALKIRE,f,m TILL M. BAUMANN,a,b,c EDDY C. CARMACK,g RANDI INGVALDSEN,h VLADIMIR V. IVANOV,i,j MARKUS JANOUT,k SIGRID LIND,h LAURIE PADMAN,l ANDREY V. PNYUSHKOV,a AND ROBERT REMBERa

a International Arctic Research Center, University of Alaska Fairbanks, Fairbanks, Alaska
b College of Natural Science and Mathematics, University of Alaska Fairbanks, Fairbanks, Alaska
c Finnish Meteorological Institute, Helsinki, Finland
d School of Ocean Sciences, Bangor University, Menai Bridge, United Kingdom
e Geophysical Institute, University of Bergen and Bjerknes Centre for Climate Research, Bergen, Norway
f Polar Science Center, Applied Physics Lab, University of Washington, Seattle, Washington
g Institute of Ocean Sciences, Fisheries and Oceans Canada, Sidney, British Columbia, Canada
h Institute of Marine Research, Bergen, Norway
i Geography Department, Lomonosov Moscow State University, Moscow, Russia
j Arctic and Antarctic Research Institute, Ocean-air interaction department, St.Petersburg, Russia
k Alfred Wegener Institute, Helmholtz Centre for Polar and Marine Research, Bremerhaven, Germany
l Earth and Space Research, Corvallis, Oregon

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ABSTRACT

A 15-yr duration record of mooring observations from the eastern (>70°E) Eurasian Basin (EB) of the Arctic Ocean is used to show and quantify the recently increased oceanic heat flux from intermediate-depth (150–900 m) warm Atlantic Water (AW) to the surface mixed layer and sea ice. The upward release of AW heat is regulated by the stability of the overlying halocline, which we show has weakened substantially in recent years. Shoaling of the AW has also contributed, with observations in winter 2017–18 showing AW at only 80 m depth, just below the wintertime surface mixed layer, the shallowest in our mooring records. The weakening of the halocline for several months at this time implies that AW heat was linked to winter convection associated with brine rejection during sea ice formation. This resulted in a substantial increase of upward oceanic heat flux during the winter season, from an average of 3–4 W m⁻² in 2007–08 to >10 W m⁻² in 2016–18. This seasonal AW heat loss in the eastern EB is equivalent to a more than a twofold reduction of winter ice growth. These changes imply a positive feedback as reduced sea ice cover permits increased mixing, augmenting the summer-dominated ice-albedo feedback.

1. Introduction

In recent decades there has been a dramatic decline in seasonal sea ice extent in the Arctic Ocean, with a more recent year-around decline in sea ice extent, area, and volume (Kwok 2018; Stroeve and Notz 2018). This change has shifted the local radiative balance resulting in a positive ice-albedo feedback mechanism as increasing lead fraction and surface melt pond areas in decaying Arctic sea ice facilitate enhanced upper-ocean solar heating and more rapid melting of ice floes (e.g., Perovich et al. 2008; Toole et al. 2010). Moreover, it was hypothesized that the declining sea ice has larger-scale hemispheric impacts on the North Atlantic Oscillation and, in consequence, midlatitude weather patterns (e.g., Francis et al. 2017; García-Serrano et al. 2015; Kolstad and Screen 2019).

Heat associated with oceanic currents originating from lower latitudes provides an important, and year-round, source of heat to the Arctic Ocean (e.g., Carmack...
et al. 2015). The dominant external source of oceanic heat is the warm (temperature 0.8°C) and salty water of Atlantic origin [Atlantic Water (AW)], which is distributed throughout the deep basins at intermediate depths (150–900 m; Fig. 1) and holds sufficient heat to melt the Arctic sea ice 3–4 times over (Carmack et al. 2015). Across much of the eastern (>70°E) Eurasian Basin (EB) this heat is isolated from the surface, and hence the sea ice, by large vertical density gradients associated with the Arctic halocline (60–150 m; Fig. 1). The presence of the halocline impedes the transport of AW heat upward toward the surface across much of the Arctic Ocean (e.g., Fer 2009). The exception to this is the western (<70°E) Nansen Basin where substantial turbulent mixing linked to the tides (Fer et al. 2010; Padman and Dillon 1991; Rippeth et al. 2015; Renner et al. 2018) and wind events (e.g., Provost et al. 2017; Graham et al. 2019) supports heat fluxes in excess of 50 W m⁻².

Inflowing AW is warming (Barton et al. 2018) driving a regime shift in sea ice cover over the past decade in the Barents Sea (Onarheim et al. 2018). There is also a growing body of evidence that the characteristics of the Arctic halocline are changing; for example, the halocline has weakened in the eastern EB since the 1970s (Steele and Boyd 1998; Polyakov et al. 2010). These changes have accelerated over the past decade (Polyakov et al. 2020a) with continuous time series from moored instruments capturing the significant weakening of the cold halocline layer (the upper part of the halocline with temperatures near freezing and negligible vertical temperature gradient) and shoaling of the AW in 2013–15 (Polyakov et al. 2017).

The combination of weaker stratification and shoaling of the AW in the EB, coupled with the loss of sea ice, has allowed progressively deeper winter ventilation in the eastern EB in recent years (Polyakov et al. 2017). This process further enhances the annually averaged upward AW heat fluxes. The shift in sea ice state and upper ocean stratification to conditions previously unique to the western Nansen Basin has been termed “Atlantification” (Polyakov et al. 2017) and represents a transition toward a new Arctic climate state, in which the geographical influence of the AW heat on sea ice volume is spreading eastward.

Since the increased oceanic heat fluxes associated with Atlantification drive sea ice melt, and reduced sea ice increases oceanic heat fluxes through increased convective entrainment in winter, this process represents a positive ice–ocean-heat feedback mechanism. This mechanism is analogous and complementary to the ice-albedo feedback, in which atmospheric warming leads to a reduction of ice and snow coverage and decreasing albedo, resulting in further snow and sea ice retreat (Manabe and Stouffer 1980).

The strength of the ice–ocean-heat feedback is determined by the vertical flux of AW heat across the halocline into the surface-forced seasonal convective layer. Polyakov et al. (2017) estimated seasonal changes of heat content Q in the eastern EB halocline (65–130 m) and an equivalent divergent heat flux (the difference of fluxes at two depth levels for which a 1D equation of heat balance for a unit-area water column is integrated) of δF₂h ~ 12 W m⁻² over this depth range for winter 2013/14, and ~8 W m⁻² for winter 2014/15. They argued that these inferred values of δF₂h exceeded previous regional estimates (e.g., Lenn et al. 2009; Polyakov et al. 2013) by a factor of 2–4 and potentially account for an additional loss of up to 18–40 cm of sea ice over this
period of time associated with the increase in upward AW heat transport. In consequence the impact of the oceanic heat flux on sea ice formation in 2013–15 was comparable to that of the atmospheric thermodynamic forcing (Polyakov et al. 2017).

The aim of this paper is to quantify the changes in the upper ocean heat content, and the consequent release of heat from the AW up into the halocline and to the surface mixed layer in the key eastern Eurasian Basin of the Arctic Ocean. We improve on the Polyakov et al. (2017) study by including new data collected over the period 2015–18 to quantify changes in the upper ocean heat content, and the consequent release of heat from the AW up into the halocline and to the surface mixed layer in the EB. We then compare these regional estimates with earlier estimates.

2. Data

Our analyses utilize observations of ocean temperature, salinity, and currents from moorings deployed in the eastern EB (Fig. 2, Table 1). Observations at the M14 mooring site began in August 2002, with several collocated moorings deployed and recovered annually prior to 2009, and longer duration of deployments since 2013 (Table 1).

Moorings deployed in summer 2013 and recovered in summer 2015 provided 2-yr-long records for most instruments except for the M15 upper ocean acoustic Doppler current profiler (ADCP), which worked for 10 months only. Mooring M3 located off the Novisibirskiiye Islands was deployed at water depth of 1350 m. Six moorings (M1–M14) formed a ~350-km cross-slope section spanning from the 250 to 3400 m isobaths along 126°E. A topographically steered boundary current flows along slope across this section (Pyushkov et al. 2015, 2018a). Averaged over 2013–15, the maximum current speed of ~11 cm s⁻¹ was found at the shallowest mooring M1 (on the 250 m isobath), with only ~0.5 cm s⁻¹ in the deep basin at moorings M13 and M16. The AW core defined by the maximum water temperature is typically located at a depth of ~250 m at the M14 mooring site.

Deployment of moorings in 2015–18 repeated the mooring distribution used for 2013–15 except that the M16 mooring was not redeployed (Table 1). Almost all mooring instruments provided full 3-yr-long records; the M13 McLane Moored Profiler (MMP) stopped recording after two years. In addition, a short-term mooring, M14-short, was deployed for 18 days only (2–20 September 2018) close to the M14 climatologic mooring site (Table 1). The short-term mooring was designed to provide current and CTD data with the most rapid possible sampling rate in the upper 200 m.

![Fig. 2. Map showing the focus of the study together with the positions of moorings and location of CTD profiles made in summer 2013, 2015, and 2018 reported in this study. The Gakkel Ridge (GR) divides the Eurasian Basin (EB) into the Nansen Basin and the Amundsen Basin. The Lomonosov Ridge (LR), Novosibirskiiye Islands (NI), Severnaya Zemlya (SZ), Franz Joseph Land (FJL), and Makarov Basin (MB) are indicated. Gray solid lines show depth in meters. The eastern EB region used for calculation of blue time series in Fig. 10 is identified by the green line.](image-url)
column (Table 1). Moorings with no MMP were equipped with long-range 75-kHz ADCPs covering deeper layers (Table 1). ADCPs provided current velocities, averaged over 2 m (prior to 2013) or 4 m (after 2013) vertical cells, with 1-h time resolution. The manufacturer’s estimates for 300-kHz ADCP accuracies are 0.5% of measured speed and 2° for current direction.

Moorings equipped with MMPs provided current velocity profiles with above mentioned profiling intervals and 0.25-m vertical resolution. The MMPs were equipped with a Falmouth Scientific Inc. (FSI) microCTD sensor in 2002–04 and a Sea-Bird Electronics (SBE) 41CP CTD sensor starting from 2004, with temperature and conductivity measurement accuracies of about ±0.002°C and ±0.0003 S m⁻¹, respectively. Prior to 2013, each MMP carried an FSI Acoustic Current Meter (ACM); after 2013, the ACMs were substituted with the FSI ACM-PLUS-MP (http://www.falmouth.com/product-information.html). The velocity precision of the FSI ACM (ACM-PLUS-MP) carried on the MMP is

<table>
<thead>
<tr>
<th>Mooring</th>
<th>Latitude (°N), longitude (°E)</th>
<th>Depth (m)</th>
<th>Instrument</th>
<th>Depth range (m)</th>
<th>Beginning of record</th>
<th>End of record</th>
</tr>
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<td>MMP</td>
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<td>2690</td>
<td>ADCP</td>
<td>5–50</td>
<td>14 Sep 2004</td>
<td>15 Sep 2005</td>
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<td></td>
<td>125, 40.194</td>
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<td>MMP</td>
<td>72–900</td>
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<td>5–57</td>
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Moorings deployed in 2002–09 and collocated with M14 mooring

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<th>Mooring</th>
<th>Latitude (°N), longitude (°E)</th>
<th>Depth (m)</th>
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<th>Depth range (m)</th>
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<td>M12</td>
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<td>787</td>
<td>ADCP</td>
<td>5–63</td>
<td>27 Oct 2013</td>
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<td>MMP</td>
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Moorings M11–M15, 2013–15

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<th>Depth (m)</th>
<th>Instrument</th>
<th>Depth range (m)</th>
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<td>23–83</td>
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<td>125, 42.673</td>
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<td>MMP</td>
<td>60–754</td>
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Moorings M11–M15, 2015–18

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<th>Latitude (°N), longitude (°E)</th>
<th>Depth (m)</th>
<th>Instrument</th>
<th>Depth range (m)</th>
<th>Beginning of record</th>
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<td>20 Sep 2018</td>
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<td>125, 58.924</td>
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Moorings M3

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<th>Latitude (°N), longitude (°E)</th>
<th>Depth (m)</th>
<th>Instrument</th>
<th>Depth range (m)</th>
<th>Beginning of record</th>
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<td>142, 14.688</td>
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<td>SBE</td>
<td>41, 45, 57, 64, 130, 270, 600</td>
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<td>M3f</td>
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<td>1357</td>
<td>ADCP</td>
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<td>142, 15.216</td>
<td></td>
<td>SBE</td>
<td>30, 50, 133, 217, 268, 614</td>
<td>7 Sep 2015</td>
<td>6 Sep 2018</td>
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</table>

Mooring M14-short (2–20 Sep 2018)
reported to be ±2% (1%) of reading and ±0.5 cm s\(^{-1}\) for velocity resolution. Compass accuracy is ±2°. All MMP sensors were calibrated before their deployment and immediately after their recovery using McLane facilities.

c. Shipborne CTD data

Mooring observations were complemented by repeated hydrographic profiles collected using a Seabird SBE911plus CTD system in 2013, 2015, and 2018 at the M14 mooring site (Fig. 2). The effective vertical resolution, considering the different sensor characteristics, is about 0.25 m. Individual temperature and conductivity measurements are accurate to ±0.002°C and ±0.0003 S m\(^{-1}\).

3. Methods

a. Defining a proxy for the Richardson number

The Richardson number (Ri) is a measure of the stability of the water column; when Ri < 0.25 the vertical shear in the flow is sufficient to generate dynamic instabilities and turbulent mixing. As such, Ri estimates provide a useful indicator for the likelihood of shear instability and mixing. The correct scale for the estimation of Ri is the Ozmidov scale [which in this case we estimate to be O (0.1) m]. However, the vertical resolution of the Ri estimate is limited by the positions of instruments on the moorings, which have a vertical resolution of 20 m. While the 20-m Ri estimates are likely to smooth out the fine structure of individual instabilities, we argue that the smaller the large-scale Ri value is, the greater the likelihood of shear instability (and so turbulence and mixing). As such, the 20-m Ri provides a useful proxy for the likelihood of shear instability. Moreover, trends in the 20-m (proxy) Ri estimate will expose trends in the likelihood of shear instability, the key interpretation here. This approach is supported by direct comparisons of dissipation and low resolution Ri estimates (e.g., Mead Silvester et al. 2014).

The mooring-based estimates of Ri (Fig. 3) are based on MMP measurements of stratification and velocity. Stratification over the 100–140 m layer is quantified using buoyancy frequency \(N\), where \(N^2 = - (g/\rho_o) \partial \rho / \partial z\), where \(\rho\) is the potential density of seawater, \(\rho_o\) is the reference density (1030 kg m\(^{-3}\)), and \(g\) is the acceleration due to gravity. The limited depth range of 100–140 m was chosen due to insufficient data coverage in early years (see Table 1). The Ri proxy was estimated as \(\mathrm{Ri} = N^2 / |U_z|^2\), where \(|U_z|\) is the magnitude of the vertical shear of the horizontal currents; \(|U_z|\) and \(N\) were calculated averaging gradients over 20-m vertical scale for all points within the 100–140 m depth range.

b. Defining timing and depth of seasonal upper ocean ventilation and divergent heat flux \(\delta F_h\)

For this analysis, temperature observations at M13, M14, M14, and M3 moorings in 2013–18 were used. SBE-37 data from non-MMP moorings M12 (2015–18), M13, M14, and M3 were complemented by MMP profiles from the M12 (2013–15) mooring. SBE-37 observations were linearly interpolated to match the MMP vertical resolution. We are interested in the analysis of seasonal ventilation of the halocline. Accordingly, temperature observations were filtered using wavelet transformations to keep seasonal variations only (and thus the different temporal sampling by MMP and SBE-37 did not affect our results). A standard package of wavelet programs was used based on the derivative of Gaussian (DoG) mother function. Estimates of heat content \(Q\) (J m\(^{-3}\), with freezing point taken as a reference temperature at a given salinity) for the halocline (65–140 m) are shown in Fig. 4. To check that the use of SBE-37 point measurements with relatively coarse vertical resolution and continuous MMP profiles for estimates of \(Q\) did not affect our results we calculated \(Q\) using MMP temperature record from M12 mooring (2015–18) twice, first with original MMP resolution and again with subsampled coarser resolution matching SBE-37 depth levels (Table 1). The estimates of \(Q\) integrated over the halocline depth range and averaged in time over the entire record length differed by 8%.

The aim is to define the timing and amplitude of upward heat flux associated with winter ventilation. To this end, we identified timing and amplitude of the maximum \(Q\) (as accumulated over the warm phase of the seasonal cycle) and the minimum of \(Q\) (associated with winter ventilation) using \(Q\) vertically integrated over 65–140 m. The depth of the ventilation is defined as the deepest point where a distinct minimum of \(Q\) was found. The maximum of vertically integrated \(Q\) was then recalculated using the depth of ventilation.

Following Polyakov et al. (2017), we limited the boundary of the winter ventilation layer to 140 m. For some years, the boundary of the layer was deeper than 140 m (as shown in Fig. 4 by the black horizontal segments located at the very bottom of the panels with \(Q\)). Therefore, our choice of the ventilation layer is conservative and estimates of divergent heat fluxes \(\delta F_h\) derived from change of heat content \(\Delta Q\) during each winter season represent the lower bound, consistent with the objectives of the study. For the upper boundary of the layer for which \(Q\) is estimated, we selected the depth 65 m, chosen because this best determines the halocline layer in which heat from the AW is stored and released (Polyakov et al. 2013, 2017). We evaluated the
sensitivity of our estimates to the choice of the boundary of the ventilation layer by calculating $\delta F_h$ for 65–140 m and 65–150 m layers. The 10-m increase in layer thickness increases $\delta F_h$ by less than 8%.

Following Polyakov et al. (2013), we estimated $\delta F_h$ (W m$^{-2}$) between two depth levels as the change, in time, of vertically integrated $Q$. This approach is based on the assumption that all change in heat content is due to vertical exchange (i.e., 1D). Note that these values are flux differences between two depth levels, and total heat fluxes may be larger than these values due to additional nondivergent heat transports; thus, our inferred estimates of divergent heat fluxes represent lower bounds for the total heat flux [for details, see Polyakov et al. (2013)].

4. Results

a. AW warming and weakening of halocline stratification in the eastern Eurasian Basin

Time series of the AW temperature show significant interannual variability (Fig. 5a). The AW in the eastern EB began warming in the early 2010s, with the AW temperature in 2018 being, on average, 0.5°C–0.7°C higher than in 2011 (Fig. 5a). This recent warming is particularly noticeable at shallower depths, with the increase in temperature at 150 m exceeding 1.5°C between 2011 and 2018. This warming over the depth range 150–750 m between September 2013–May 2014 and September 2016–May 2017 is partially associated with shoaling of the upper halocline boundary.
(Fig. 5c) and a substantial increase in AW layer thickness (Fig. 6).

Cross-correlation analysis of time series of AW temperature measured at 250 m from 1997–2018 in Fram Strait, the entry point of AW into the Arctic, and from 2002–18 in the eastern EB (red time series in Fig. 5a) shows the strongest correlation, $R = 0.67$, for a lag of 682 days (Fram Strait series leads; Fig. 5b). The fit between the two time series is better over the last 7–8 years than it is over the earlier period. The ~2-yr lag suggests that warm pulses of AW that entered the Arctic Ocean through Fram Strait are traveling toward the eastern

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**Fig. 4.** The 65–140 m layer depth vs time of (left) water temperature and (center) annual component of heat content $Q$. Annual components are obtained via bandpass filtering using wavelet transformations. Horizontal black segments identify the depth of seasonal ventilation; dates identified by their ends are used to compute vertically integrated $Q$ shown in the lower parts of panels in the right column. (right) Vertically integrated $Q$ for the beginning (warm phase) and end (cold phase) of seasonal ventilation (shown in lower parts of the panels) and divergent heat fluxes $\delta F_h$ (upper parts) for four moorings.
EB at a speed 2–2.5 times faster than that estimated for a warm AW pulse that entered the eastern EB in 2004 (Polyakov et al. 2005). This implies that the rate of advection has increased over time. However, noisy data due to gaps in the EB record preclude meaningful statistical analysis using just the early part of the time series. Assuming that the lagged correlation between the two time series will persist in the near future, the latest part of the Fram Strait series (not shown) implies that the AW temperature in the eastern EB reached its peak in late 2018 (these data are not yet available) and will slowly decrease over the next 1–2 years.

Temperature and salinity profiles in the eastern EB from CTD during 2013–18 and MMP during 2003–18 recorded a decline of stratification ($N^2$) over the 110–140 m depth range of the halocline (Figs. 1c and 3a,b), which may be a result of both the shoaling of AW and weakening of halocline stratification. Polyakov et al. (2018) used available potential energy defined for the variable-depth halocline to show overall weakening stratification in the EB since the 1980s, with accelerated tendencies in the 2010s compared with the 2000s. However, there was substantial weakening of halocline stability from 2013 to 2015 (Polyakov et al. 2017), which continued in 2015–18, and which was also partially associated with shoaling of the AW (Fig. 6) found at 80 m depth, as inferred from the most recent observations in winter 2017/18 (Fig. 5c). This represents the shallowest depth the AW has been observed in the 15 years of mooring deployments. As these estimates used a linear interpolation of CTD time series made at 38 and 107 m at mooring M14, we are not able to definitively conclude that the cold halocline layer was present (albeit very thin) during the winter of 2017/18. However, the record suggests the extreme thinning (or even absence) of the Arctic cold halocline layer for several months at this time (Figs. 5c and 6), implying that AW heat was exposed to winter convection associated with sea ice formation and brine rejection.
b. Increased oceanic heat fluxes and ice loss in the eastern Eurasian Basin

The weakening stratification, shoaling of the AW layer and increase of current shear in recent years (e.g., Polyakov et al. 2020b, manuscript submitted to Geophys. Res. Lett.) have altered the seasonal cycle of upward AW heat transport (Fig. 4). Estimated change in heat content $Q$ from the halocline (65–140 m) during winter, averaged at four moorings, is equivalent to mean divergent heat fluxes (see section 3) of $dF_h = 12.0 \pm 5.5, 3.5 \pm 2.2, 3.0 \pm 1.9, 12.9 \pm 1.7,$ and $20.6 \pm 6.8$ W m$^{-2}$ for five winters from 2013/14 through 2017/18 (Figs. 4 and 7). For three of these winters (2013/14, 2016/17, and 2017/18), $dF_h$ greatly exceeded (from threefold to fivefold) the previous estimates derived from summer 2007/08 microstructure observations over the Laptev Sea slope (Lenn et al. 2009; Polyakov et al. 2019) and winter 2009/10 ITP-37 observations in the central Amundsen Basin (Polyakov et al. 2013). For the winters of 2014/15 and 2015/16, estimates of $dF_h$ were comparable to upward heat fluxes of about 3–4 W m$^{-2}$ from 2007/08. We attribute the decrease of $dF_h$ in 2015/16 (cf. Polyakov et al. 2017) to an anomalous freshening event in the upper ocean. This freshening is evident in data collected at mooring M13 (Fig. 8), which shows that strong upper ($<75$ m) ocean stratification (evidenced by high $N^2$ values) in 2016 precluded seasonal ventilation beyond the SML. Stronger stratification in winter 2015 compared with winters of 2014 and 2017 (Fig. 8d) limited seasonal ventilation to the upper ~115 m, thus not extending deeply enough to reach the main pool of AW heat (Fig. 8b). In consequence the heat flux is limited. The strongest heat flux is inferred for winter 2017/18 and is associated with the weakest stratification (Fig. 3), providing further evidence for the key role of stratification in mediating upper ocean ventilation.

The new estimates of seasonal ventilation of heat evaluated from the $\delta F_h$ for the winter seasons of 2016/17 and 2017/18 are equivalent to $78 \pm 4$ and $93 \pm 29$ cm reductions in ice growth, respectively, for the eastern EB (Fig. 7), given that one year of a heat flux of 1 W m$^{-2}$ in isolation is equivalent to about 10 cm of sea ice loss. This represents a twofold increase in the sea ice loss rate.
compared to that estimated for 2013/14 (54 cm) and 2014/15 (40 cm) (Polyakov et al. 2017), and so partially explains intensified eastern EB sea ice loss in more recent years (Onarheim et al. 2018; Stroeve and Notz 2018).

Time series from the shallower moorings (M12 and M13) show strong seasonal variations in the AW core temperature, which may be associated with seasonal displacement of the AW core relative to the slope (e.g., Baumann et al. 2018). However, the consistently low correlation between $Q$ and the AW core temperature records, for all mooring sites (Fig. 9), implies that cross-slope shifts in AW temperature core are not a major driver of the seasonal variation in $Q$ in the halocline. The correlation between $Q$ and AW core temperature at the shallowest mooring (M12) where currents are strongest is also weak ($R = 0.29$), indicating that advection does not provide a significant contribution to the seasonal variability of $Q$. This evidence is consistent with the results of Polyakov et al. (2017), who argued that the in-phase seasonal maxima and minima of wavelet transforms of $Q$ at all mooring sites suggests that the observed winter ventilation is driven by surface cooling and sea ice formation—and not by lateral advection. They reasoned that spatially varying water

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**Fig. 8.** (a) Potential temperature, (b) annual component of heat content $Q$ obtained by bandpass filtering of daily heat content using wavelet spectra; horizontal black segments identify the depth of seasonal ventilation, (c) salinity, and (d) squared buoyancy frequency for the M13 mooring.
transports across the slope, ranging from 13 cm s$^{-1}$ [measured over the upper continental slope (250–700 m) by moorings M1$_1$ and M1$_2$] to 1–2 cm s$^{-1}$ (measured at 2700 m and deeper, at mooring locations M1$_4$, M1$_5$, and M1$_6$) make the in-phase pattern of the seasonal signal at all moorings impossible to explain using the advective mechanism. Furthermore, mooring M1$_6$, which was farthest from the near-slope boundary current, in the ocean interior, yielded estimates for $F_h$ with magnitudes and phases consistent with estimates from the other moorings deployed on the eastern EB continental slope in 2013–15 (Polyakov et al. 2017).

The one-dimensional approach adopted here can be further validated by considering the magnitude of the lateral temperature gradient necessary to explain the estimated heat flux, if advection were to dominate. Taking our estimate of along-slope current speed of 2 cm s$^{-1}$ implies that, to explain the estimated heat flux with lateral processes, the lateral temperature gradient $dT/dx$ would need to be about $4 \times 10^{-3}$ $\degree C$ km$^{-1}$, about 5 times larger than the observed value of about 0.75 $\times 10^{-4}$ $\degree C$ km$^{-1}$ between Fram Strait and the central Laptev Sea slope temperature [1.8$\degree C$ of cooling (from 3$\degree C$ to 1.2$\degree C$) over $\sim$2400 km]. Another potential contributor to the observed ventilation rates is lateral eddy
fluxes. Ventilation of halocline by eddies is, however, difficult to quantify using available data. Nevertheless, considering that the typical time of eddy passing across the mooring site is about a week with the average frequency about one eddy per month (Pnyushkov et al. 2018b), it is unlikely that eddies can significantly contribute to changes of the heat content at seasonal time scales. These considerations imply the uncertainty in the 1D flux calculation from lateral advection and diffusion is small.

5. Discussion and conclusions

Time series measurements from a 15-yr mooring record in the eastern EB of the Arctic Ocean demonstrate that the previously identified weakening of stratification over the halocline, which isolates intermediate depth AW from the sea surface, over the period 2003–15 (e.g., Polyakov et al. 2017, 2018), has continued at an increasing rate in more recent years (2015–18). In consequence, oceanic heat fluxes for the winters of 2016–18 are estimated to be greater than 10 W m$^{-2}$. These fluxes are substantially larger than the previously reported winter estimates for the region for 2007/08 of 3–4 W m$^{-2}$ (Lenn et al. 2009; Polyakov et al. 2019) and comparable to the estimates for the winters of 2013–15 (Polyakov et al. 2017), implying a significant enhancement of the role of oceanic heat in this region in recent years.

Moreover, the increased vertical heat fluxes have been accompanied by increased upper-ocean current speeds $|U|$ and the magnitude of vertical shear in the horizontal velocities $|U_z|$ over the period 2015–18 (Polyakov et al. 2020b, manuscript submitted to Geophys. Res. Lett.). Using mooring observations from 2003 to 2018, these authors showed that time-averaged values of $|U|$ and $|U_z|$ in the upper 60 m of the water column increased by about 20% and 40%, respectively. In the lower halocline (110–140 m), $|U|$ was generally larger after 2008, increasing on average from 2.5–3.5 cm s$^{-1}$ in 2003–08 to about 4–5 cm s$^{-1}$ in 2009–18 (Figs. 3c,d) although the change was not as strong in very recent years, 2016 and 2018, when compared to 2009–15. There is also a clear transition in $|U_z|$, with significantly larger shears evident post-2010, and in particular in the summer of 2018 (Figs. 3c,d). However, Pnyushkov et al. (2018a) found no significant change in the mean along-slope water transport over the same period.

The combination of reduced stratification and increased shear implies a decrease of the gradient Richardson number (Ri) defined in section 3 (Figs. 3e,f), consistent with an increased turbulent heat flux, associated with vertical mixing by shear instabilities. Although the Ri estimates are based on 20 m vertical resolution measurements, they show a clear trend toward reduced dynamic stability, which may be interpreted as a tendency toward increased turbulent mixing in recent years, coincident with the increase in maximum halocline heat content (Fig. 4). This tendency is particularly strong in 2018 with amplified velocity shear in the relatively weakly stratified upper ocean (Fig. 3).

The increased shear and weakening of stratification as prerequisites for enhanced turbulent mixing are consistent with the recent transition in the upper ocean to conditions previously unique to the western Nansen Basin, a process called Atlantification (Polyakov et al. 2017). Our analyses confirm that, in part, the loss of stratification in the eastern EB halocline can be attributed to processes originating upstream. For example, the change in halocline salinity, the main contributor to water column stability in the eastern EB, is correlated with upper ocean salinity changes in the northern Barents Sea with a lag of approximately 2 years (Fig. 10) (Lind et al. 2018), revealing coherent interannual variability.
between the two regions. In the Barents Sea, these changes were found to be closely linked to declines in sea ice imports to the Barents Sea (Lind et al. 2018; Barton et al. 2018). The shift toward higher salinities in the eastern EB lags the changes in the northern Barents Sea by about 1 year (Fig. 10), implying an eastward lateral progression of Atlantification. Shelf–basin interactions may also be contributing to the observed warming (e.g., Timmermans et al. 2018).

Our observations point to the shift of this region of the eastern Arctic Ocean toward a new regime that is more typical of the continental slope regions of the western Nansen Basin where surface conditions are strongly influenced by oceanic heat imported from the Atlantic Ocean (Fig. 11). The flux of AW heat to the sea ice cover and the atmosphere has increased, during the winter season, from an average of 3–4 W m$^{-2}$ in 2007–08 to >10 W m$^{-2}$ in 2016–18, equivalent to more than a two-fold reduction of winter ice growth over the last decade.

The process described here represents a positive feedback, analogous to the ice-albedo feedback, since increased ocean heat flux to the sea surface reduces ice thickness and increases its mobility, increasing atmospheric momentum flux into the ocean and reducing the damping of surface-intensified baroclinic tides (Carr et al. 2019). We refer to this process as the ice–ocean-heat feedback. As with the ice-albedo feedback, the contribution of the ice–ocean-heat feedback to long-term sea ice trends depends on the seasonal variability of several factors that affect mixing rates including sea ice concentration and thickness, baroclinic tidal response to seasonally varying stratification, and wind stress impacts on sea ice and on AW shoaling. The transition in dominant mixing regime from double diffusion to shear-driven mixing also affects the relative magnitudes of buoyancy fluxes due to heat and salinity transports; the vertical diffusivities for heat and salt are the same in shear-driven turbulence, but are different for double diffusion (Kelley 1984). Coincident vertical nutrient fluxes, which support oceanic primary productivity, food web structure and carbon export from the atmosphere to the seabed (Bluhm et al. 2015; Falk-Petersen et al. 2015), will also increase. Moreover, the nutricline has shoaled in recent years (Fig. 1d), relieving nutrient limitations, while declining sea ice cover relieves light limitations; both of these changes are influenced by Atlantification, suggesting regional-scale enhancement of biological productivity in the central Arctic Ocean.

As ice thins—through atmospheric forcing, changing ocean heat fluxes, and feedbacks—upper-ocean stratification is responding and a new Arctic state is emerging, which may not be easily reversed. For example, a large

![Conceptual model of shift of the mixing regime in the eastern EB in recent years and associated suite of processes and state conditions including: 1) thinner, more mobile ice; 2) warmer surface mixed layer (SML); 3) weakening/retreat of cold halocline (HC) layer; 4) increased AW vertical heat flux (red arrows) and horizontal currents and their vertical shear (blue arrows); 5) shoaling of upper AW boundary; and 6) replacement of DD by turbulent mixing.](image-url)
anomaly in AW heat input coupled with shoaling may lead, through the ice/ocean-heat feedback, to an expanding and more permanent Atlantic-dominated state wherein the hydrographic structure of the halocline no longer provides sufficient insulation between the intermediate depth AW and the sea ice, even when the heat flux associated with the AW is relaxed. This potential for a permanent transition of the eastern Arctic to a new state emphasizes the pressing need for the incorporation of improved mixing schemes into Arctic climate models in order to better simulate the evolving halocline stratification and its impact on sea ice state.

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![Figure A1](link)
Russian Federation (Project RFMEFI61619X0108). All mooring data used in this study are available at https://doi.org/10.18739/A2N37R and https://doi.org/10.18739/A2HT2GB80.

APPENDIX

Building Long-Term Time Series

Changes in the 110–140 m (halocline) layer at the M14 mooring site shown in Fig. 3 were documented using MMP records for 2003–07 and 2013–18, SBE37 records from M1g mooring in 2008–10, and ADCP records for 2008–10. This layer is the key part of the lower halocline water (Figs. 1a,b) and has sufficient data coverage for the task. All original mooring data were processed to make them comparable. We filtered MMP vertical profiles with a running-mean filter to reduce resolution to 4 m, equivalent to the 2013–18 ADCP observations. We subsampled ADCP and SBE37 data in time to match coarser MMP temporal resolution. The vertical shear is calculated consistently using gradients over 20-m vertical scale. Reconstruction of the record at the M14 mooring site

Fig. A2. Multiple regression reconstruction of (a),(b) salinity and (c),(d) current speed |$|U|$| at the M14 mooring site using data from the M13 and M15 moorings for 170–210 m depth range. Shown are the daily (dotted) and 3-month running mean smoothed time series of salinity in (a) and |$|U|$| in (c) from the M13, M14, and M15 moorings, and the original (blue) and reconstructed (red) time series of salinity in (b) and |$|U|$| in (d) from the M14 mooring. Relatively high correlations between the original and reconstructed time series attests of good quality of reconstruction.
in 2013–18 using MMP data from nearby moorings is described below.

There were no MMP measurements within the 110–140 m depth range at the M14 mooring in 2013–15 and 2015–18 (Table 1). Records for these years and depth range were reconstructed using weighted interpolated estimates from the neighboring M13 and M15 moorings.

This approach is justified by the observed monotonic cross-slope change of current speed from M13, M14, and M15 mooring records for the depth ranges where overlapping data are available for the three moorings (Fig. A1). Estimates of buoyancy frequency $N$ derived from temperature and salinity provided by these three moorings are statistically indistinguishable (Fig. A1).

Multiple regression is used to further validate the use of records from moorings M13 and M15 to reconstruct time series of temperature, salinity, and current speed at mooring M14 for 2013–18. The model of multiple regression is

$$Y = \beta_o + \beta_1 X_1 + \beta_2 X_2,$$

where $\beta_1 = [(r_{XY} - r_{X_1Y})(1 - r_{X_2Y})](\sigma_Y/\sigma_X)$, $\beta_2 = [(r_{XY} - r_{X_1Y})(1 - r_{X_2Y})](\sigma_Y/\sigma_X)$, and $\beta_o = \overline{Y} - \beta_1 \overline{X_1} - \beta_2 \overline{X_2}$; the overbar denotes means, $\sigma$ denotes standard deviations, $r$ is used to denote cross-correlation coefficients, and the random error term is neglected. For independent parameters $X_1$ and $X_2$ the time series from M13 and M15 moorings are used, and the time series from M14 is used as the dependent variable $Y$. We neglected the high-frequency part of the records by applying low-pass 3-month running mean filtering to each time series used in the tests because in this study we mainly focus on longer-term (interannual) trends and changes. Evidence for the validity of this approach is provided in Fig. A2.

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