Higher Ocean Wind Speeds During Marine Cold Air Outbreaks

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- Cold air outbreaks
- Air–sea interaction
- Polar climate
- Wind speed
- Polar lows

**Abstract**

Marine cold air outbreaks (MCAOs) are large-scale events in which cold air masses are advected over open ocean. It is well-known that these events are linked to the formation of polar lows and other mesoscale phenomena associated with high wind speeds, and that they therefore in some cases represent a hazard to maritime activities. However, it is still unknown whether MCAOs are generally conducive to higher wind speeds than normal. Here this is investigated by comparing ocean near-surface wind speeds during MCAOs in atmospheric reanalysis products with different horizontal grid spacings, along with two case studies using a convection-permitting numerical weather prediction model. The study regions are the Labrador Sea and the Greenland–Iceland–Norwegian (GIN) Seas, where MCAOs have been shown to be important for air–sea interaction and deep water formation. One of the main findings is that wind speeds during the strongest MCAO events are higher than normal and higher than wind speeds during less severe events. Limited evidence from the case studies suggests that reanalyses with grid spacings of more than 50 km underestimate winds driven by the large ocean–atmosphere energy fluxes during MCAOs. The peak times of MCAO usually occur when baroclinic waves pass over the regions. Therefore, the strong wind episodes during MCAOs generally last for just a few days. However, MCAOs can persist for 50 days or more.

1. **Introduction**

Events in which cold and dry air masses are transported out from over sea-ice or cold land masses over an area of open ocean are known as marine cold air outbreaks (MCAOs). Such events are important for several reasons. They pose a direct hazard to human activities in that they set up a large energy imbalance which can drive extreme weather. For instance, MCAOs constitute a favourable environment for the formation of polar lows (Kolstad, 2011), intense mesoscale cyclones that form over open ocean at high latitudes (Rasmussen and Turner, 2003). Even when no polar lows form, strong surface wind speeds can occur along Arctic fronts separating cold air masses from warmer air further south (Grønås and Skeie, 1999). The large vertical temperature differences between the ocean surface and the air during MCAOs can give rise to large surface heat fluxes, as measured by numerous flight campaigns (Grossman and Betts, 1990; Chou and Ferguson, 1991; Brümmer, 1996; Renfrew and Moore, 1999) and with buoys (Harden et al., 2015). One of the mechanisms that arise due to these fluxes are roll clouds or cloud streets, with their large wind speed gradients across small distances (Hartmann et al., 1997; Renfrew and Moore, 1999). In short, MCAOs and associated weather features with high wind speeds are important for air–sea interaction at high latitudes (Våge et al., 2008; Condron and Renfrew, 2013; Isachsen et al., 2013; Papritz and Spengler, 2017).

Although MCAOs are linked to known severe weather types, it is still an open question whether MCAOs are generally conducive to high surface wind speeds. If they are, it would mean that MCAOs pose a risk in themselves, even without the presence of polar lows or similar kinds of extreme weather. This question is the main motivation for this study. Using two reanalysis products, the correlations between wind speeds and MCAOs are calculated for two marginal seas in the North Atlantic: the Labrador Sea and the Greenland-
Iceland-Norwegian (GIN) Seas. These regions are known to experience a high frequency of MCAOs during winter (Kolstad et al., 2009; Fletcher et al., 2016). They are also areas in which surface water masses are transformed as a result of interactions with the atmosphere and sea-ice (Dickson et al., 1996; Marshall and Schott, 1999).

Figure 1. (a) Advanced Very High Resolution Radar (AVHRR) image taken at 13:44 UTC on 16 January 2005. (b) AVHRR image taken at 21:06 UTC on 25 January 2007. The images are reproduced with the kind permission of the Dundee Satellite Receiving Station in the UK.

Figure 1 shows two of the most extreme MCAOs in recent decades in the Labrador Sea and the GIN Seas. In the first of these, roll clouds formed due to advection of cold air masses from the North American continent (Figure 1(a)). The driving force is a low-pressure system located between Greenland and Iceland, and a strong Greenland tip jet event (Doyle and Shapiro, 1999; Moore and Renfrew, 2005) is in progress. In the second case, shown in Figure 1(b), the cold air originates over sea-ice and is transported southwards through the Fram Strait between Greenland and Svalbard. The white patches to the south-west of Svalbard are an indication of convection along an elongated high-vorticity shear line. These often form during periods with northerly and north-westerly flow over the Norwegian Sea, and arise due to interactions with the topography and coastline of Spitsbergen (the largest island in the Svalbard archipelago).

There are valid reasons to suppose that wind speeds are higher during MCAOs than in normal conditions. First, the static stability is low during MCAOs, and thus turbulent mixing facilitates a downward flux of momentum (Monin and Obukhov, 1954). Second, as the surface heat fluxes during MCAOs can be intense, a strong flow is needed to prevent the cold air masses from being heated so much from below that the MCAO expires (Papritz and Pfahl, 2015). The third reason has already been mentioned: multiple small-scale features, such as the ones observed in the satellite images in Figure 1, commonly occur in MCAO conditions. Although this does not imply that MCAOs generally produce strong winds, it does indicate that some MCAOs do.

There is also cause to believe that the strong winds during MCAOs are not always captured by reanalyses. For instance, it has been demonstrated that ERA-Interim does not fully reproduce polar lows (Zappa et al., 2014; Pezza et al., 2016). The reason is that the spatial scales of several phenomena are too small for reanalyses
to be even theoretically able to reproduce them. At least four, and sometimes up to ten, grid points are required to resolve wave-like mesoscale features (Walters, 2000). The grid spacing (Δx) of ERA-Interim (Dee et al., 2011) is 80 km, which means that it cannot capture features on scales of less than 300 km. One example of a small-scale feature pertinent for this study is the shear line south of Svalbard, of which an example was shown in Figure 1(b). A case study by Sergeev et al. (2017) indicated that the maximum horizontal wind speed gradient across one such shear line was 25 m s⁻¹ over just 50 km, less than the Δx of ERA-Interim and identical to the Δx of the second version of the Modern Era Retrospective analysis for Research and Applications (MERRA-2) reanalysis (Bosilovich et al., 2015; Molod et al., 2015). A related phenomenon is the topography-induced Greenland tip jet that forms near Cape Farewell. A model-based case study showed that decreases of Δx from 100, 50, and 25 km to 10 km led to stepwise improvements in the representation of tip jets (DuVivier and Cassano, 2013). As for polar lows, decreasing Δx from 12 to 4 km ‘significantly improved’ the simulation of two cases over the Norwegian Sea (McInnes et al., 2011).

A consequence of the coarse resolution of the reanalyses is that reanalysis-based correlations between MCAOs and wind speed may misrepresent the actual relationship between the variables. The evidence of underestimation of wind speeds during MCAOs in reanalysis products abounds. Substantial improvements in the representation wind speeds have been shown for the first version (Δx = 30 km) of the Arctic System Reanalysis (ASRv1 henceforth; Bromwich et al., 2016) for several small-scale features, including southeast Greenland barrier winds and katabatic flows (Moore et al., 2015). Even better results have been demonstrated for the second version of the Arctic System Reanalysis (ASRv2), for which Δx = 15 km, in a study of topographically forced winds near Greenland (Moore et al., 2016), tip jets included. To illustrate how small-scale wind features not resolved by reanalyses can arise during MCAOs, case studies of two of the most extreme MCAOs during the reanalysis period are performed with a convection-permitting numerical simulation (Δx = 4 km).

This study is structured around the search for answers to three broad questions designed to shed light on the role of MCAOs in driving surface wind speeds over the ocean, with implications for maritime activities and air–sea interaction. The first question is: \textit{What is the relationship between the severity of MCAOs and associated wind speeds?} To investigate this, MCAO events are detected using time series of an established MCAO index. The events are then ranked by severity, and the wind speeds during the events are studied. The second question is: \textit{How do MCAOs and associated winds speeds evolve?} By studying the 100 most intense events in each region, composite averages of wind speed, surface pressure and the MCAO index are calculated and studied. The third question addresses the known fact that MCAOs form a favourable environment for mesoscale features not necessarily resolved by reanalysis products, and is formulated as follows: \textit{Do reanalyses with a grid spacing of 50 km or more underestimate the actual relationship between MCAOs and wind speeds?} This is explored by performing case studies of two extreme MCAOs, where wind speeds simulated using a convection-permitting model are compared with wind speeds from the reanalyses.

\section{Data and Methods}

Here I follow Papritz et al. (2015) and use an MCAO index defined as $M = \theta_S - \theta_{850}$, where $\theta_{850}$ is the potential temperature at 850 hPa, and $\theta_S$ is the ‘potential skin temperature’, calculated as $\theta_S = T_S \left( \frac{P_0}{P_S} \right)^{R/c_p}$, where $T_S$ is the skin temperature, $P_S$ is the surface pressure, and $P_0$ is 1000 hPa. November to March (NDJFM) data from ERA-Interim and MERRA-2 (Global Modeling and Assimilation Office (GMAO), 2015b, 2015a) are used for the period 1980–2016, as this period was available for both data sets when the analysis was performed. ASRv1 (Bromwich et al., 2012) and ASRv2 (Polar Meterology Group/Byrd Polar and Climate Research Center/The Ohio State University, 2017) only cover the period 2000–2012 and are therefore only used for the two case studies.
Figure 2. Monthly mean skin temperature (filled contours) and the approximate sea-ice edge (white contours) in MERRA-2 for January 2016. The two study regions are outlined in black: the GIN Seas (20°W–17°E and 66.5°N–76.5°N) and the Labrador Sea (65°W–43°W and 52°N–60°N).

The MCAO index is computed for each grid point and then area-averaged for two regions: the GIN Seas and the Labrador Sea. These regions are drawn as boxes in Figure 2, where the monthly mean MERRA-2 skin temperature and the approximate sea-ice edge (contoured sea-ice fraction for the value 0.55) are also shown. Land grid cells and cells for which the skin temperature is lower than 271.5 K is not included when area-averaging. Ten-meter wind speed and the meridional wind component are also area-averaged for the same grid cells. Anomalies are calculated with respect to the long-term mean values for each month, day and time.

For each region, time series of 12-hourly values of the MCAO index $M$ are used to identify events. A new event is started when $M > 0$ K and ended if $M < 0$ K for two successive time steps. This is done to merge two neighbouring events when they are separated by only one time step with $M < 0$ K. After an event is terminated, the last time that $M > 0$ K is set as the last time of the event. Events are only recorded if they attain an $M$ value of at least 3 K and last for at least 48 hours. The peak of each event is required to occur in NDJFM, but the events are allowed to start in October or end in April.

Different values for $M$ have been used in other studies. Fletcher et al. (2016) used 3 K as a lower threshold for moderate MCAO events, using air temperatures at the 800 hPa level. Papritz et al. (2015) applied five thresholds ranging from 0 K to 8 K, yielding different intensity categories. Here a single threshold of 3 K is used, but for one part of the analysis the MCAO events are ranked according to their highest $M$ value. The results do not change appreciably when other nearby thresholds are used instead.

Two case studies, one for each region, are performed to compare winds speeds from convection-permitting simulations with winds speeds from the reanalysis products during extreme MCAOs. Version 3.7.1 of the Weather and Research Forecasting (WRF) model is used, with three-hourly ERA-Interim data as boundary and initial forcing. The simulations are initiated 48 hours prior to the peaks of the MCAO index in both cases. Inner domains with $\Delta x = 4$ km are nested inside outer domains with $\Delta x = 12$ km. Except for the slightly larger $\Delta x$, the WRF configuration used follows one that was found to successfully represent the main features of a
hurricane-like polar low in the Barents Sea (Kolstad and Bracegirdle, 2017). Cumulus parameterisation is switched off. Note also that spectral nudging (von Storch et al., 2000) of winds and temperature above 700 hPa is applied in the outer domain, with a nudging wavelength of 1100 km. Sea surface temperatures and the sea-ice fraction are kept fixed throughout the simulations. ASRv1 and ASRv2 are used for comparison with WRF and the other reanalyses, and QuikSCAT satellite-derived wind speeds (Hoffman and Leidner, 2005) are used for validation.

Figure 3. The circles show wind speed anomalies at the peak of each MCAO event. The filled circles indicate events for which the meridional wind during the peak of the event was negative (open circles indicate positive meridional winds). The events were sorted by the highest event-wide value of $M$. The events with the highest $M$ value are shown first (rank 1). The lines show the best least-squares linear fit to the wind speed anomalies.

3. Results

3.1. MCAO events ranked by intensity

The first question posed in the introduction relates to the relationship between the severity of MCAOs and their associated wind speeds. For the two study regions, events were identified using the algorithm described in the previous section. Events occur slightly more often in the Labrador Sea than in the GIN Seas (1.8 vs. 1.5 events per month). The circles in Figure 3 show the wind speed anomalies during the peak of each event. About 70 percent of the Labrador Sea events and practically all the events in the GIN Seas occurred in northerly flow. The lines show the best linear fit to the wind speed anomalies, calculated using least squares. To assess the statistical significance of the linear fits, the Spearman’s rank correlations (Spearman, 1904), also known as Spearman’s $\rho$, between the $x$ values and the rank of each $y$ value in each graph were calculated. These are listed above each panel. Each $\rho$ value is statistically significant at the $10^{-5}$ level. The differences between the highest and lowest points along the linear fits shown in Figure 3(a) are 3.7 m s$^{-1}$ in ERA-Interim and 4.6 m s$^{-1}$ in MERRA-2 (Figure 3(b)). In the GIN Seas region, the differences are 1.8 and 1.9 m s$^{-1}$, respectively, as shown in Figures 3(c) and (d).

An alternative way to assess the difference between strong MCAO events and weaker events is to use mean values of $M$ and wind speed anomalies throughout each event instead of instantaneous values during the peak of each event. The results of this exercise are shown in Figure 4, where the events are sorted by the mean $M$ value. Note that results are only shown for MERRA-2, as these are virtually indistinguishable from the results obtained with ERA-Interim. By this measure, the number of events with northerly flow in the Labrador Sea...
increases to 77 percent. The $\rho$ values are all significant at the 1 percent level. According to the linear fits shown in Figure 4(a), the wind speed anomalies during the most intense events in the Labrador Sea are 1.5 m s$^{-1}$ higher than during the least intense events. The corresponding difference is 1.3 m s$^{-1}$ in ERA-Interim. In the GIN Seas, shown in Figure 4(b), the difference in wind speed is 1.1 m s$^{-1}$ in MERRA-2 and 0.9 m s$^{-1}$ in ERA-Interim.

![Figure 4](image)

**Figure 4.** As Figure 3, but here the events are sorted by the event-wide average $M$ values. Results are only shown for MERRA-2.

![Figure 5](image)

**Figure 5.** In each panel, the grey bars show the number of MCAO events active at the specified times relative to the time of the peak of the MCAO index $M$ during each event according to MERRA-2. The black lines show anomalous $M$ values, averaged over the events that were active at each time. Anomalies that are significant at the 5 percent level are marked with circles.

### 3.2. Evolution of MCAO events

The evolutions of the 100 most intense events in MERRA-2 in the two regions are now investigated through a composite approach. I elected to focus on a limited number of events, and the events are ranked by their peak event-wide $M$ value. In other words, the events are the first 100 events shown in Figure 3. Qualitatively similar results were obtained when compositing on all the events. The black lines in Figure 5 show the average $M$ anomalies across all active events at each time relative to the peaks of the events. The circles indicate anomalies that are significant at the 5 percent level (estimated using a two-sided 10,000-member Monte Carlo approach where the years were scrambled). In both regions, the $M$ peaks are well-defined and short-lived, with notably higher $M$ anomalies found only from about 24 hours before till about 24 hours after the peaks. Before and after that short period, the $M$ anomalies hover around 3–4 K throughout the period shown. The anomalies are significantly positive throughout.

![Figure 6](image)

**Figure 6.** Shows the average wind speed anomalies relative to the peaks of the 100 events. Panel (a) shows that the highest average wind speeds over the Labrador Sea are found during the peak times. There is more of a tendency towards higher wind speeds before than after the peaks. In the GIN Seas, the pattern of higher wind speeds before than after the peaks of $M$ is more evident than in the Labrador Sea, as shown in Figure 6(b). In fact, the highest wind speeds occur 12 hours before the peaks of the events, not during the peaks as in the Labrador Sea region. After the peaks, the wind speed anomalies decline and even become significantly negative after a few days.
What are the reasons for the significantly positive wind speed anomalies several days before the peaks of MCAO events in both regions and the subsequent drops in wind speed after the peaks? To investigate this, Figure 7 shows the composite average evolution of MCAO events in the Labrador Sea region. Again, results are only shown for MERRA-2 due to their close agreement with ERA-Interim. Note also that the statistical significances of the anomalies are not computed, as this part of the investigation is meant as a qualitative rather than quantitative survey of the dynamical evolutions of the events.

Figure 7(a) indicates that the most important precursor of MCAO events is a low-pressure anomaly between Greenland and Iceland, already present 48 hours before the peaks of the events. The wind speed anomaly over the region is weakly positive, whereas the wind speeds further east are higher. One day later, an anomalous trough stretches southwards from Cape Farewell, and positive wind speed anomalies have appeared in the anomalous north-westerly flow in the western part of the region (Figure 7(b)). This increase in wind speed was also seen in Figure 6(a). During the peak, as shown in Figure 7(c), the wind speed anomalies have increased substantially (see also Figure 6(a)) relative to 24 hours before. This is due to the emergence of a high-pressure anomaly in the southern part of the region shown, along with a considerable deepening of the low anomaly over southern Greenland. Precisely such a low-pressure system was evident in the satellite image shown in Figure 1(a). The strong winds south of Cape Farewell attest to a high frequency of Greenland tip jets during the peaks of MCAO events. One day after the peaks the wind speeds have subsided, with higher wind speeds to the north-east of the region (also a sign of tip jets). To sum up, the evolution shown in Figure 7 suggests that the peaks of the MCAO events in the Labrador Sea occur as a result of the transient coexistence of high pressure anomalies in the south and strong low pressure anomalies over southern Greenland.
Figure 8. As Figure 7, but for the GIN Seas region.

The average evolution of the 100 strongest MCAO events in the GIN Seas is shown in Figure 8. Panel (a) shows that a large-scale low anomaly east of the prime meridian is a precursor 48 hours before the peaks, along with a less pronounced high anomaly along the east coast of Greenland. Together these pressure anomalies yield anomalous northerly flow into the region. The average wind speed anomaly is already positive, especially in the northern part of the region. The above-average wind speeds agree with Figure 8(b). One day before the peaks, the wind speeds have increased additionally, as shown in Figure 8(b). The chief cause appears to be enhanced pressure gradients, mainly due to a further development of the ridge east of Greenland. Another likely cause is the twin influences of the sea-ice edge and the topography of Spitsbergen (see Figure 1(b)). During the peaks of the MCAO events, the wind speed over the region is still high, as shown in Figure 8(c). Recall that the peak in wind speed on average occurs 12 hours before the peaks (Figure 6(b)). The primary change from 24 hours before is a rapid strengthening of the ridge east of Greenland. Figure 8(d) shows that the pressure gradients are weak one day after the peaks. The average wind speed anomalies are also weak at this point, as already indicated by the rapid decline in wind speed shown in Figure 6(b).

3.3. Case studies of two MCAO events

To illustrate the impact of model grid spacing on the representation of small-scale winds during MCAOs, one event in each region is investigated in detail. For the Labrador Sea, the event with the highest peak $M$ value during 2000–2009 is studied. This period was chosen because there is data available for five datasets: ERA-Interim, MERRA-2, ASRv1, ASRv2, and QuikSCAT. A WRF simulation with $\Delta x = 4$ km was performed, starting at 12:00 UTC on 14 January 2005. The satellite image in Figure 1(a) was taken at 13:33 UTC on 16 January. Figure 9(a–e) shows the 10-metre wind speed at 09:00 UTC on the same day for the reanalyses and the WRF simulation. The black contours indicate the approximate sea-ice edge. The QuikSCAT data shown in Panel (f) was retrieved between 08:05 and 10:06 on the same day. The white areas reflect missing QuikSCAT data over and near sea-ice and land. The white contours in all the panels show the 25 m s$^{-1}$ isoline.

The QuikSCAT image reveals a strong tip jet event. This is also reproduced by all the models, although the magnitudes of the wind speeds vary. WRF produces more small-scale detail than the other models, and some features resemble the roll cloud structures seen in Figure 1(a), although the 4 km grid spacing is probably not high enough to fully resolve roll clouds. The wind speed associated with the tip jet in WRF, ASRv2, ASRv1, and to some extent MERRA-2, agree quite well with QuikSCAT. The tip jet wind speed in ERA-Interim is too low. There is also a region with high wind speeds to the west of 50°W, 55°N which is captured only by WRF, ASRv2 and ASRv1.
The highest MCAO index value in the GIN Seas region in the period 2000–2009 occurred on 26 January 2007. The satellite image in Figure 1(b) was taken at 21:06 UTC on 25 January. Figure 10 shows simulated wind speeds at 21:00 UTC on that day, along with QuikSCAT data retrieved between 19:12 and 23:14 UTC. In these maps the white contours are drawn at 20 m s\(^{-1}\). The wake downstream of Spitsbergen is best reproduced by ASRv1, ASRv2 and WRF. Of these, the WRF wind speeds compare best with QuikSCAT in the area to the west of Spitsbergen. The small-scale structures appear to be resolved gradually better with increasing resolution. The lower-resolution products also give too low wind speeds over much of the GIN Seas. In particular, the strong winds along the shear zone south of Spitsbergen, which are also evident in the satellite image in Figure 1(b), are best resolved by WRF. This is therefore another example of underestimation of the true wind speeds during a strong MCAO in the coarser-resolution reanalyses.
4. Summary and discussion

The first question in the Introduction was: What is the relationship between the severity of MCAOs and associated wind speeds? The results presented here clearly show that the most severe MCAO events coincide with stronger wind speeds than weaker events. Although not discussed earlier, the overall (i.e., using all times, not just during MCAO events) correlations between anomalies of the MCAO index $M$ and wind speed are statistically significant at the 10^{-10} level in both regions, both at lag-0 and when the wind speeds lead $M$ by up to several days. This demonstrates that MCAOs are associated with above-average wind speeds.

The second question was: How do MCAOs and associated wind speeds evolve? The composite analysis of MCAO events indicates that strong MCAOs commence when the MCAO index is already high. In the Labrador Sea region, the existence of a low-pressure anomaly between Iceland and Greenland is a precursor 48 hours before the peaks of strong MCAO events. In the GIN Seas, northerly winds over the region are a precursor 48 hours before the peaks. The peaks in both regions occur when ridges and troughs rapidly develop to create a flow that is optimal for transporting cold air masses over the region. The fast build-up and the overall transient nature of the peaks of the events indicate that they are associated with passages of baroclinic waves. This result is consistent with the findings of Fletcher et al. (2016), who concluded that strong MCAOs occur in the cold air sector of mid-latitude cyclones. Additionally, Papritz et al. (2015) found that about 80 percent of MCAOs in the Pacific sector of the Southern Ocean are induced by the flow induced by the passages of synoptic-scale cyclones. It is also interesting to note that MCAOs are essentially self-destructive. The vigorous transfer of energy from the ocean to the atmosphere becomes more effective when the wind speed is high, which it is during the peaks of the events. This energy transfer leads to a cooling of the sea surface and a warming of the...
lower atmosphere through sensible heat fluxes and latent heating (Papritz and Pfahl, 2015), and thereby forces a decrease in the MCAO index. The MCAO will therefore extinguish itself unless it is reinforced by fresh, incoming cold air masses due to some external factor.

The third and final question was: *Do reanalyses with a grid spacing of 50 km or more underestimate the actual relationship between MCAOs and wind speeds?* As only two case studies were investigated, a complete answer is elusive. The case studies performed here, in which the 30 km ASRv1 and the 15 km ASRv2 were used in addition to the coarser reanalyses, indicate that a dense grid spacing (perhaps less than 30 km) is required to obtain a good reproduction of the actual ocean wind speeds during strong MCAOs. As mentioned, other studies have also shown that reanalyses do not fully resolve some severe weather features associated with MCAOs, such as polar lows. Thus, the answer to the third question is a qualified ‘Yes’.

One key finding is that MCAOs are large-scale phenomena that can last for days and even weeks. The longest events in both regions lasted for over 50 days. It is therefore possible that MCAOs can be forecast on the sub-seasonal time scale, i.e. beyond 5–10 days, which is the typical range of numerical weather prediction. Cai *et al.* (2015) claimed that continental-scale cold air outbreaks can be forecast up to one month ahead by means of a hybrid statistical/dynamical model system, using key indices for stratospheric circulation as predictors. The lagged association between stratospheric polar vortex anomalies and subsequent cold air outbreaks in winter has been the topic of many studies (e.g. Thompson *et al.*, 2002; Kolstad *et al.*, 2010). It is also possible that empirical indices based on SST anomalies and sea-ice can be used for forecasting purposes. If MCAOs can be forecast on the sub-seasonal time scale, there might be a potential for developing early warning systems for severe weather in marine environments. This is proposed as important future work with potentially high relevance for stakeholders in coastal, Arctic regions.

The relationship between MCAOs and wind speeds also has implications for air–sea interaction and thereby possibly also for convection in the ocean. It has for instance been shown that Greenland tip jet events have an impact on deep convection in the Irminger Sea south of Iceland (Våge *et al.*, 2008). The comparisons between the reanalysis products and the high-resolution ASR and WRF simulations also suggested that wind speeds are underestimated in coarse-resolution reanalyses. In a series of papers, Alan Condron and his colleagues first showed that polar mesocyclones (including polar lows) were not adequately represented by the ERA-40 reanalysis (Condron *et al.*, 2006), and then used a ‘bogusing’ approach to parameterise these cyclones in ocean model experiments (Condron *et al.*, 2008). Compared with model runs with no parameterisation, they found ‘increases in the simulated depth, frequency and area of deep convection in the Nordic seas, which in turn leads to a larger northward transport of heat into the region, and southward transport of deep water through Denmark Strait’ (Condron and Renfrew, 2013). Similar results were obtained by Jung *et al.* (2014), who compared ocean model simulations forced with high- and low-resolution atmospheric data. In a related study by Holdsworth and Myers (2015), coupled ocean–ice model simulations forced with hourly atmospheric data were compared with simulations where the high-frequency atmospheric forcing was filtered out. In the Labrador Sea, these differences yielded large changes in the oceanic mixed layer depth. These studies imply that there is a potential for parameterising MCAOs in ocean models, perhaps in ways that can make use of the results presented here.

5. **Conclusions**

Severe MCAO events tend to arise when the background flow is favourable, meaning that the wind direction is such that cold air is already being advected over the ocean. The peaks of the events coincide with the passages of baroclinic waves. A key factor is ridging, which contributes to enhanced pressure gradients and a stronger flow of cold air. High wind speed anomalies occur a few days before and during the peaks of the events. After the peaks, however, the wind speeds subside rapidly. But although the strong wind episodes are transient, MCAOs can be sustained for up to 50 days. High-resolution model experiments indicate that the actual correlation between wind speed and severe MCAOs could be higher than what is indicated by reanalysis products with a grid spacing of 50 km or more.

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