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Modelling the biogeographic boundary shift of *Calanus finmarchicus* reveals drivers of Arctic Atlantification by subarctic zooplankton

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Abstract

Biological communities in the Arctic are changing through the climate-driven encroachment of subarctic species. This "Atlantification" extends to keystone Calanoid copepods, as the small-bodied Calanus finmarchicus increases in abundance in areas where it overlaps with larger Arctic congeners. The environmental factors that are facilitating this shift, whether related to optimal conditions in temperature or seasonality, remain unclear. Assessing these drivers at an Arctic-wide scale is necessary to predict future ecosystem change and impacts. Here we have compiled range-wide occurrences of C. finmarchicus and a suite of seasonal biophysical climatologies to build a boreo-Arctic ecological niche model. The data set was divided into two eras, 1955-1984 and 1985-2017, and an optimized MaxEnt model was used to predict the seasonal distribution of the abiotic niche of C. finmarchicus in both eras. Comparing outputs between eras reveals an increase in habitat suitability at the Arctic range edge. Large and significant increases in suitability are predicted in the regions of the Greenland, Labrador, and Southern Barents Seas that have experienced reduced seaice cover. With the exception of the Barents Sea, these areas also show a seasonal shift in the timing of peak habitat suitability toward an earlier season. Our findings suggest that the Atlantification of Arctic zooplankton communities is accompanied by climate-driven phenology changes. Although seasonality is a critical constraint to the establishment of C. finmarchicus at Arctic latitudes, earlier sea-ice retreat and associated productivity is making these environments increasingly favorable for this subarctic species.

KEYWORDS boreal, climate warming, phenology, plankton, sea-ice, species distributions

1 | INTRODUCTION

In recent decades, the enhanced inflow of warm Atlantic water in to the Eurasian Arctic represents a step toward a new Arctic climate state (Polyakov et al., 2017; Tsubouchi et al., 2021). As well as contributing to rapid rates of warming and sea-ice loss (Polyakov et al., 2017), "Atlantification" or borealization of the Arctic Ocean is also triggering a shift in Arctic biological communities as subarctic species

This is an open access article under the terms of the Creative Commons Attribution License, which permits use, distribution and reproduction in any medium, provided the original work is properly cited. © 2021 The Authors. *Global Change Biology* published by John Wiley & Sons Ltd. encroach northward (Polyakov et al., 2020; Wassmann et al., 2011). The effects of Atlantification are being recorded in Arctic benthic (Kortsch et al., 2012), pelagic (Eriksen et al., 2017), and fish assemblages (Fossheim et al., 2015; Frainer et al., 2017) and are disrupting its food web structure (Kortsch et al., 2015; Vihtakari et al., 2018).

Vital to the Arctic food web are three coexisting species of Calanoid copepods; Calanus hyperboreus, C. glacialis, and C. finmarchicus. This Calanus complex dominates herbivorous mesozooplankton throughout the Arctic and northern seas (Mauchline, 1998). Although they display differences in growth, development, and reproduction as imposed by different optimal environments, all three species are adapted to graze on the seasonal phytoplankton bloom, converting low-energy carbohydrates and proteins into high-energy wax esters (lipids), which they store to aid survival over-winter and fuel reproduction (Falk-Petersen et al., 2009). In doing so, they are a fundamental source of energy for higher trophic levels, sustaining vast fish stocks, seabird colonies, and marine mammal populations (Wassmann et al., 2006).

The Calanus life-cycle is complex. For C. finmarchicus at its northern range, it lasts 1-year. During spring and summer it develops from eggs via six naupliar and four copepodite stages to its major overwintering stage (CV), accumulates lipid reserves, and then descends to an overwintering depth below 500 m. Molting to the final adult stage and mating occur at the end of the winter. C. finmarchicus is generally an income breeder, thus while it accumulates enough lipid reserves to survive the winter, it relies on the nutritional input of the spring bloom upon resurfacing to fuel gonad maturation, egg production, and nauplii development (Hirche et al., 1997; Niehoff et al., 2002).

In the last 30 years, C. finmarchicus-a species typically associated with Atlantic water masses—has undergone a poleward distribution shift in the North Atlantic (Chust et al., 2014). This small, subarctic species has also increased in contribution to overall Calanus biomass in several Arctic regions (Aarflot et al., 2018; Hop et al., 2019; Moller & Nielsen, 2020; Weydmann et al., 2014). The fate of its larger congeners endemic to Arctic water masses is less clear due to a lack of long-term data from the high Arctic. There are indications of a decline in the southern margins of their distribution (Aarflot et al., 2018; Chust et al., 2014) while coastal populations of C. glacialis show stable population size (Hop et al., 2019; Moller & Nielsen, 2020). Overall, there is growing evidence that Atlantification is favoring a shift toward an Arctic Calanus community with smaller body size and less lipid content (Renaud et al., 2018). Shifts in the composition of Arctic Calanus and their functionally important traits (body size and lipid content) are of significant concern as a reduction in lipid production may impact the energy available to higher trophic levels (Kwasniewski et al., 2010; Renaud et al., 2018).

Any effect of C. finmarchicus' increasing presence is likely to be exacerbated under future climate warming. To be better equipped to predict this, we must move from local observations of encroachment to a regional understanding of the underlying mechanisms facilitating it. The thermal tolerance range of C. finmarchicus was found to be important in determining recent and future distribution shifts

within the North Atlantic and subarctic (Beaugrand et al., 2008, 2009; Chust et al., 2014; Helaouet & Beaugrand, 2007; Reygondeau & Beaugrand, 2011; Villarino et al., 2015), yet it is the length of the growing season, that is, the period of phytoplankton availability, which is thought to be a critical limitation in successful recruitment of C. finmarchicus at their Arctic range edge (Hirche & Kosobokova, 2007; Ji et al., 2012). Here, the timing of the phytoplankton bloom is strongly dictated by local light conditions related to seasonal sea-ice cover (Falk-Petersen et al., 2009). Recent studies have highlighted a link between the biogeography of C. glacialis and C. hyperboreus and changing sea-ice characteristics (Ershova et al., 2021). If the biogeographic boundary of C. finmarchicus is also limited more by seasonality and food availability than by temperature per se, areas which have increased in suitability for C. finmarchicus in recent decades would have also experienced the greatest change in seasonality due to reduced sea-ice cover.

Ecological niche models (ENMs) and allied species distribution models are valuable tools for understanding environmental correlates of Calanus biogeography (Albouy-Boyer et al., 2016; Beaugrand et al., 2013; Helaouet & Beaugrand, 2007; Helaouet et al., 2011; Record et al., 2018; Villarino et al., 2015; Wilson et al., 2016). These have largely taken advantage of long-term monitoring data sets within the North Atlantic yet often lack the seasonal resolution necessary to uncover seasonal as well as decadal shifts in suitable habitat. Here, we extend these efforts by compiling rangewide collections of C. finmarchicus and a suite of seasonal biophysical climatologies to build the first boreo-Arctic ENM for this species. We assess how the distribution of their ecological niche, particularly at the poleward boundary, has shifted seasonally between two eras (1955-1984 and 1985-2017), which correspond to cool and warm thermal regimes in the region, respectively (Beaugrand, 2009).

We aim to (a) characterize the spatial and temporal patterns of C. finmarchicus distribution on a pan-Arctic scale and (b) determine the abiotic drivers, including the relative importance of thermal tolerance limits and seasonality, in facilitating the Atlantification of zooplankton communities in the Arctic.

2 MATERIALS AND METHODS

2.1 Species occurrence records

A total of 65,037 georeferenced occurrence records of copepodite stages of C. finmarchicus were compiled from six different online repositories (OBIS, PANGAEA, NSF Arctic Data Center, BODC, COPEPOD global plankton database, NOAA NODC), comprising more than 45 individual data sets. The continuous plankton recorder data set accounts for 85% of records. The month, depth, and year of collection were retained. 98% of records were classes as CV, CVI, or adult life stages and the remaining were classed as CI-CIV. Citations for all data sets used are given in File S2.

After identifying overlapping data sets between repositories, remaining records were thinned to retain only one occurrence per season per grid cell (resolution: $0.25^{\circ} \times 0.25^{\circ}$) using the "spThin" R package (Aiello-Lammens et al., 2015). This removes the fewest records necessary to substantially reduce the effects of sampling bias, while simultaneously retaining the greatest amount of useful information.

Environmental predictors 2.2

Ten environmental predictors were identified as candidate variables for the niche model. These include: temperature, bathymetry, slope, chlorophyll a, sea-ice concentration, salinity, silicate, pH, photosynthetically active radiation (PAR), and current velocity. Representing the sea surface only, these data were obtained for the full study region (longitude: -180°E, 180°W; latitude: 30°N-90°N) using a combination of empirical observations and model re-analyses (see Table S1 in File S1 for details on all data sources). As C. finmarchicus is absent from the Pacific Arctic, these regions were removed from environmental raster data to limit model outputs to the known spatial distribution range of the species. Temperature and PAR were the only variables to be highly correlated (Pearson's r = .71, Figure S1). Methods used to deal with correlation are described in the section "model evaluation and tuning."

For each predictor, seasonal climatologies were obtained for two eras of approximately 30 years; 1955-1984 and 1985-2017. These eras were chosen as they represent two different (cool and warm) oceanographic regimes in the northern North Atlantic which are known to have affected zooplankton community dynamics (Beaugrand, 2009; Figure 1). Based on data availability, the seasonal partitions represent the months: Jan-Feb-Mar (JFM), Apr-May-Jun (AMJ), Jul-Aug-Sep (JAS), and Oct-Nov-Dec (OND).

Occurrence data were matched to the most appropriate environmental data in relation to the season and year in which they were collected by adapting the method from Duffy and Chown (2017). For example, a record collected in May 2002 was assigned the environmental conditions from the AMJ 1985-2017 climatology. This accounts for variation in environmental conditions (particularly sea-ice extent and primary productivity) between seasons and multidecadal time periods, and, as only the most appropriate environmental data are paired with each occurrence record, the accuracy of the ENM can, in principle, be improved (Duffy & Chown, 2017).

MaxEnt ecological niche models 2.3

Occurrence and environmental data were fitted to the presenceonly ecological niche modelling algorithm MaxEnt v. 3.4.1 (Phillips & Dudik, 2008) using the "SDMtune" R package (Vignali et al., 2020). MaxEnt estimates the conditional probability of the presence of a species relative to locations where the species has been observed by sampling the environment at a range of "background" locations across the study region and discriminating these from locations where species is known to be present. MaxEnt assumes background locations adequately cover areas accessible to the species and that the presence of localities are unbiased and cover important environmental gradients (Jarnevich et al., 2015). Although a lack of absence data prevents probability estimates of a species presence and predictions of a species' realized distribution, presence-only outputs more closely represent the existing, fundamental niche of a species (Soberon & Nakamura, 2009).

Sampling bias and background data selection 2.4

Sampling in this region is skewed in favor of sea-ice free areas, leading to spatial and environmental sampling bias. To avoid generating distribution maps which overly reflect sampling effort (Botella et al., 2020), the geographic and temporal distribution of background data were selected based on an estimate of zooplankton sampling effort across the region (Fourcade et al., 2014; Phillips et al., 2009). All occurrence records within the phylum "arthropoda" occurring in the upper 200 m and collected between 1960 and 2017 were downloaded from OBIS.

FIGURE 1 Annual mean sea surface temperature and sea-ice extent within the North Atlantic and European Arctic (-75°W, 75°E, 30°N, 90°N) between 1955 and 2017. Dashed vertical line denotes transition between the two eras compared within the study (1955-1984 and 1985-2017). Data plotted from Hadley Centre Sea Ice and Sea Surface Temperature data set (Rayner et al., 2003) and National Snow and Ice Data Center Gridded Monthly Sea Ice Extent and Concentration 1850 onward data set v2 (Walsh et al., 2019).



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A kernel density surface of these records was generated (Figure S2) and used to weight the selection of 10,000 random points across the region (i.e., more points taken from areas with higher-density values; Figure S2). These points were also randomly assigned to a season and era proportionate to the temporal distribution of zooplankton data. Background points were matched to the most appropriate environmental data as described for the occurrence records.

2.5 Model evaluation and tuning

Cross-validation approaches partition the data in to K number of folds. The model is then run K times, withholding a different fold for model evaluation each time (Araujo et al., 2019). The spatialBlock function of the "blockCV" package in R (Valavi et al., 2019) was used to create folds (K = 5) that account for spatial autocorrelation in the environmental data. Spatial block size was determined by fitting isotropic variogram models using 5000 random points from each environmental predictor raster. This finds the effective range of spatial autocorrelation and the spatial block size was based on median of these ranges. The occurrence and background data within each block were then allocated to a fold. To ensure block-to-fold allocations achieved an even spread of data, 100 iterations were ran with the most even allocation of data being used. See Figure S3 for map of spatial blocks and fold assignment.

The area under the receiving operator curve (AUC) and true skill statistic (TSS) metrics were used to evaluate model discriminatory performance on the evaluation (test) fold. The AUC score is a widely used rank-based measure of predictive accuracy that can be interpreted in the context of MaxEnt as the probability that a randomly chosen presence location is ranked higher than a randomly chosen background point (Merow et al., 2013). A model with no discriminatory power will have an AUC value equal to 0.5 (no better than random), whereas a model with perfect fit would have an AUC value of 1.0. TSS values range from -1 to 1, with values of 0 or less reflecting a model that is no better than random and values closer to 1 being better at discerning presence and background points (Allouche et al., 2006).

To prevent model overfitting, the gridsearch function of "SDMtune" package was used to find the optimal combination of MaxEnt hyperparameters. Varying combinations of regularization parameter (0.2-3) and iteration number parameter (300-900) were tested, and the combination with the highest AUC_{TEST} and TSS_{TEST} scores was considered optimal. Only linear and quadratic feature class settings were used in all combinations. Additional, more complex transformations such as the "hinge" feature class did not improve model performance and generated less realistic response curves to environmental parameters.

The model with optimal hyperparameters was further tuned using (1) the varSel function to remove any correlated variables, removing the one that results in the best performing model when removed and (2) the reduceVar function to find and remove environmental predictors with low model contribution (<3% permutation importance) when their removal did not decrease model performance based upon the model's mean AUC_{TEST} value.

2.6 Model prediction

The final, optimized model was used to predict the habitat suitability of C. finmarchicus across the region of interest. Separate predictions were made for each season (JFM, AMJ, JAS, OND) and era (1955-1984, 1985-2017). For each prediction, outputs from each crossvalidation fold as well as the mean across all folds were retained to explore between-fold variation.

Model output analyses 2.7

To assess changes in the spatial pattern of habitat suitability between the two eras, we subtracted model outputs for the most recent era from the former. To assess seasonal changes between the eras, we used ArcGIS Cell Statistics tool to determine, for both eras, which season gave the highest habitat suitability value in each grid cell. Subtracting these two outputs from each other allowed an assessment of where timings in peak suitability had shifted between eras. To assess the significance of habitat suitability change, the 95% confidence interval (CI) was calculated for each grid cell and era based on variation from the cross-validation folds. Grid cells that had no overlap in CI range between eras were classed as being significantly different.

Sensitivity analyses 2.8

Sensitivity analyses were carried out to account for potential misidentification of occurrences between C. finmarchicus and its congener, C. glacialis (Choquet et al., 2017). In regions where their ranges overlap, a random subset (10% and 20%) of C. finmarchicus occurrences were replaced with a corresponding number of C. glacialis records. These models are named 10%_c.glacialis and 20%_c.glacialis, respectively, and Figure S4 shows localities of dropped/replaced occurrences in each model. To check that the seasonal assignment process of occurrences to environmental data did not influence model results, we carried out a third sensitivity analysis (model name: 30%_season_shift) whereby 50% of records collected in months 1, 4, 7, and 10 (first month of each climatology) were re-assigned to one season earlier. Similarly, 50% of records collected in months 3, 6, 9, and 12 (last month of each climatology) were re-assigned to one season later. This led to 34% of all occurrence records being assigned to a different season and the remaining 66% of occurrences remained unchanged. A final analysis (model name: original_no_ice) was carried out to test the sensitivity of model outputs to the inclusion of seaice concentration. These ENMs were optimized following the same procedure as for the original model. Further details of these analyses are given in Table S2.

3 | RESULTS

3.1 | Spatial change in habitat suitability

The predicted geographic distribution of *C. finmarchicus* for each season and era are shown in Figure 2 with the variation between cross-validation folds given in Figure S5. Habitat suitability is predicted to be highest in the East Atlantic during Jan–Feb–Mar, whilst it is predicted to be highest during JAS at high latitudes including the Barents Sea (Figure 2).

Across its Arctic range edge, the model predicts there to have been large increases in *C. finmarchicus* habitat suitability between eras. This has predominantly occurred where sea-ice cover has declined (Figure 3). In these regions of increased suitability, the 95% CIs in predictions do not overlap between eras, suggesting that the predicted increase in suitability is significant (Figure S6). Regions and levels of increased suitability are consistent in the outputs from sensitivity analyses 10%_c.glacialis, 20%_c.glacialis, and 30%_season_shift (Figure 4). When sea-ice concentration is removed (original_no_ice), predicted increases are present throughout the Arctic range edge of *C. finmarchicus*, but values are lower (Figure S8). In all but the last sensitivity analysis, the vast majority of cells predicted to have increased in suitability remain significant (Figure S6).

3.2 | Temporal change in habitat suitability

We find a seasonal shift in the timing of peak habitat suitability, most notably in the Labrador Sea, northern North Atlantic, and Greenland Sea regions (Figure 5). Within these areas, the timing of peak habitat suitability has advanced forward by one season, that is, changing from being optimal during JAS in the former era to being optimal in AMJ more recently. All cells predicted to have advanced in timing of peak suitability were found to be significant, based on a lack of 95% CI overlap between eras. These regions coincide with areas of retreating sea-ice cover, and this result is consistent across sensitivity analyses with the exception of model *original_no_ice* (Figure S7).

3.3 | Model performance and predictor importance

The optimized MaxEnt model retained five environmental predictors; sea-ice concentration, temperature, chlorophyll a, bathymetry, and salinity. Model performance metrics indicate strong discrimination ability with mean AUC_{TEST} = 0.73, and $TSS_{TEST} = 0.41$. On average, sea-ice concentration had the highest permutation importance at 45.5%, followed by temperature, salinity, chlorophyll a, and bathymetry (Table 1). Response curves



FIGURE 2 Predicted habitat suitability (conditional probability of presence) for *Calanus finmarchicus* for each season during (a) era 1 (1955–1984) and (b) era 2 (1985–2017) using MaxEnt's cloglog transformed output. White line denotes the average seasonal position of the sea-ice edge, defined at 15% sea-ice concentration



FIGURE 4 From left to right, predictions of increased habitat suitability (>0.1) between eras (1985-2017 and 1955-1984) using (a) the original model and (b-e) four different sensitivity analyses (see Section 2 for details). Green and black lines denote the mean position of the sea-ice edge for the older era and the recent era, respectively

(Figures S9-S13) show a strong negative relationship between C. finmarchicus habitat suitability and sea-ice concentration (Figure S9) and indicate an optimal temperature of 8.9°C (Figure S10). Sea-ice concentration and temperature ranges within the upper quartile of suitable habitat values are 0%-24.2% and 4.9-12.7°C, respectively.

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Sensitivity analyses to account for potential misidentification of C. finmarchicus and C. glacialis (Choquet et al., 2017) resulted in minimal change to AUC and TSS performance metrics, a small decline in the contribution of sea-ice concentration (Table 1), and a less severe negative relationship between habitat suitability and sea-ice concentration (Figure S14). Sensitivity analyses to account for seasonal assignment of occurrence records resulted in no change to model performance or variable contribution. A further test that withheld the inclusion of sea-ice concentration (model: *original_no_ice*) was found to have decreased model performance and increased importance of salinity and temperature predictors (Table 1).

Large-scale changes in the abundance and distribution of marine

species are omnipresent and consistent with ocean warming over

4 DISCUSSION

Season of maximum suitability Jan-Feb-Mar sea-ice extent Jul-Aug-Sep 1955-1984 Apr-May-Jun 1985-2017 Jan-Feb-Mar

FIGURE 5 Solid filled areas represent the season containing highest suitability value for Calanus finmarchicus during era 1 (1955-1984). Hatching denotes areas where the highest suitability value advanced forward by one season during era 2 (1985-2017). White and black lines denote the mean position of the sea-ice edge for the older era and the recent era, respectively

the last century (Hastings et al., 2020). Robust to the sensitivity analyses tested, our results reveal that suitable habitat for C. finmarchicus has increased at Arctic latitudes in the last 30 years, extending the previously known range shift within the North Atlantic (Chust et al., 2014). This is consistent with regional observations of boreal plankton and benthic species becoming more dominant within Arctic ecosystems (Aarflot et al., 2018; Dalpadado et al., 2020; Fossheim et al., 2015; Kortsch et al., 2012; Moller & Nielsen, 2020; Polyakov et al., 2020).

Predictions of "suitable habitat" from presence-only ENMs can be interpreted as showing the potential distribution of a species. This is typically broader than their realized distribution because the model does not incorporate absence records and does not account for biological interactions such as competition which further constrain where a species can persist (Soberon & Nakamura, 2009). In this study, results show a similar pattern to the core distribution of C. finmarchicus described by Choquet et al. (2017). Known zones of expatriation, where the species is found through advection but cannot successfully complete its life-cycle, such as in the Arctic Ocean basin, were not predicted to be highly suitable. Thus, our outputs represent regions where surface conditions are suitable for copepodite survival and where population recruitment may occur locally. Moreover, our model predicts that C. finmarchicus habitat is characterized by optimal surface temperatures between 4 and 12°C, with a peak at 9°C. These are consistent with previous regional model estimates (Albouy-Boyer et al., 2016; Beaugrand et al., 2013; Helaouet & Beaugrand, 2007, 2009) and observations (Bonnet et al., 2005; Strand et al., 2020) for this species. Although temperature has a strong influence on the biogeography of C. finmarchicus, our findings suggest that other factors-in addition to temperature-may have influenced the opening up of suitable habitat at their Arctic range edge, as we detail below.

We find a strong overlap between regions of sea-ice retreat and regions predicted to have undergone: (a) an increase in suitable habitat and, (b) a seasonal advancement in suitable habitat. Although this may, in part, be influenced by greater sampling effort in low sea-ice conditions, the importance of sea-ice parameters in determining the biogeography of other Arctic Calanus (C. glacialis and C. hyperboreus) has recently been demonstrated (Ershova et al., 2021; Feng et al., 2016, 2018). Our findings suggest that

TABLE 1 Percent contribution (%) of each environmental variable to model performance (fivefold mean ± 1 SD). Model names are Original = final Calanus finmarchicus model; 10% and 20%_c.glacialis = replaced 10% and 20% of occurrences with C. glacialis; 30%_season_ shift = reassigned 30% of occurrences to earlier/later seasonal climatology; original_no_ice = original model without sea-ice concentration as a variable

Environmental variable	Original	10%_c.glacialis	20%_c.glacialis	30%_season_shift	original_no_ice
Sea-ice concentration	52.56 ± 8.15	51.18 ± 7.64	48.31 ± 7.13	52.31 ± 8.20	-
Temperature	31.08 ± 11.66	32.30 ± 10.77	32.30 ± 11.93	31.12 ± 12.07	50.43 ± 10.59
Salinity	9.59 ± 4.92	10.03 ± 5.93	11.68 ± 5.87	9.28 ± 4.89	33.61 ± 10.46
Chlorophyll a	4.62 ± 1.80	6.51 ± 1.67	5.92 ± 1.18	5.22 ± 1.39	11.78 ± 3.17
Bathymetry	2.16 ± 1.27	-	1.79 ± 1.70	2.08 ± 1.31	4.18 ± 4.93



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phenological changes caused by the retreating ice-edge may also be an important driver of Arctic Atlantification of zooplankton: with sea-ice loss, the seasonal conditions necessary for *C. finmarchicus* to succeed at Arctic latitudes have started to emerge. This gives cause and context to recent empirical observations as areas of earlier and/or increased suitability from this model correspond with Atlantic-Arctic gateway areas known to have experienced biomass increases of *C. finmarchicus* including the Barents Sea (Aarflot et al., 2018), Disko Bay (Moller & Nielsen, 2020), and the Fram Strait (Weydmann et al., 2014).

Our study is limited to inferring indirectly the link between sea-ice concentration and suitable seasonality for *C. finmarchicus*. However, this premise is supported by studies showing a strong correlation between decreased summer sea-ice concentration and an earlier Arctic phytoplankton bloom (Kahru et al., 2011; Song et al., 2021), the peak of which has advanced by up to 50 days in the Baffin Sea (Kahru et al., 2011) and over a month in areas of the Barents Sea (Dalpadado et al., 2020). Studies from the Bering Sea have also found zooplankton community shifts linked to combined effects of temperature and primary production between ice-covered and ice-free years (Kimmel et al., 2018).

A prolonged growing season, enabled by an earlier sea-ice retreat and food availability, has important consequences for the Calanus life-cycle, bringing favorable conditions for growth and development (Feng et al., 2016), reproductive success (Ringuette et al., 2002), and life-cycle duration (Falk-Petersen et al., 2009). Studies from the northward range edge of C. finmarchicus indicate that warmer years may initiate earlier spawning and accelerated development (Weydmann et al., 2018), increasing the time window to build up sufficient reserves to overwinter successfully and reproduce the following season. Indeed, Tarling et al. (in press) assessed the population dynamics of C. finmarchicus within the Fram Strait in 2019. They found that a proportion of the adult population had enough lipid reserves to survive over-winter, and early developmental stages were present in early summer, both of which suggest successful local recruitment. Thus, whilst our predictions of an increase in the latitude and timing of suitable habitat do not conclude a new residency of C. finmarchicus at these locations, there is localized evidence that this may be the case. A full picture of the changing biogeography of C. finmarchicus requires further insights at the population level and complementary, basin-scale predictions of suitable overwintering habitat to confirm locations where their life-cycle can be completed (Melle et al., 2014).

Our results are directed by long-term, seasonal climatologies of environmental conditions. This is useful in understanding of how environmental changes may have influenced *C. finmarchicus* habitat over multi-decadal time periods. However, our results should also be viewed within the context of inter-annual environmental variation. The Arctic congeners *C. hyperboreus* and *C. glacialis* are, to varying extents, able to use energy reserves via capital breeding and have highly flexible life history strategies (Daase et al., 2013; Falk-Petersen et al., 2009; Sainmont et al., 2014; Soreide et al., 2010). Although *C. finmarchicus* does also use some capital resources, it is predominantly an income breeder, that is, requiring food provided by the spring bloom to contribute to facilitate maturation and reproduction, and is, thus, less equipped to deal with high inter-annual variability in the bloom phenology (Falk-Petersen et al., 2009). Its inability to cope with a short growing season in addition to low temperatures is regarded as the main limiting factors to allow it to survive and reproduce in the polar basin (Hirche & Kosobokova, 2007; Ji et al., 2012). Thus, as long as environmental conditions such as timing of ice break up and onset of the spring bloom remain highly variable in the areas defined as newly suitable, *C. finmarchicus* may struggle to reproduce and survive on a year-to-year basis, and successful establishment will be dependent on the constant replenishment of the population from the south.

As the Arctic climate state changes and leads to increased areas of open water, distribution shifts are to be expected for all Arctic Calanus species (Feng et al., 2016). Recently, Ershova et al. (2021) found a similar positive correlation between C. glacialis (and to some extent C. hyperboreus) habitat and low seaice cover, suggesting that both of these Arctic Calanus species also need extended periods of open water to complete their life-cycle. Thus, a continued northward shift of the marginal ice edge may enable these Arctic species to proliferate more within the polar basin. However, while decreasing sea-ice cover may advance the bloom phenology and, thus, open up more suitable habitat earlier in the year, there is a limit as to how far these factors will aid the northward extension of C. finmarchicus (Ji et al., 2012; Slagstad et al., 2011). Even if the Arctic Ocean becomes ice free, the light climate at extreme high latitudes will limit primary production for long periods each year, and high inter-annual variability in sea-ice extent leads to unpredictable bloom phenology, providing conditions in which the Arctic congener species are still better adapted than C. finmarchicus (Daase et al., 2013; Falk-Petersen et al., 2009). In areas of co-existence such as the Barents Sea and in the fjords of Svalbard and Greenland, there is no conclusive evidence that an increase in one Calanus species is directly detrimental to another (Hop et al., 2019; Moller & Nielsen, 2020), although the degree to which their respective niches overlap remains poorly understood. Furthermore, difficulties in distinguishing between C. finmarchicus and C. glacialis morphologically (Choquet et al., 2018), and a lack of long-term data with reliable species identification, limits our ability to assess any changes to their co-occurrence.

Understanding the adaptability and resilience of each *Calanus* species will require life-stage specific estimates of suitable habitat and fitness under a range of Arctic conditions. Although the correlative model applied here is the first step in examining broad-scale patterns of change, potential sources of bias remain (e.g., geographic sampling bias) because of its data-driven approach. Mechanistic approaches that estimate an organism's energetic budget at a fine temporal resolution and that incorporate the Arctic's extreme light environment, will have an important, complementary role in predicting the success of subarctic species at Arctic latitudes (Ljungstrom et al., 2021). Novel observation methods (Vilgrain et al., 2021) and

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increased sampling within sea-ice environments will also be important in overcoming data limitations in the Arctic and for parameterizing and validating model outputs.

Results from this study suggest that the seasonal conditions necessary for C. finmarchicus to survive have emerged at their Arctic range edge in recent decades. In these Arctic gateway regions, encroachment of C. finmarchicus is likely to alter the overturning and availability of energy in the pelagic ecosystem due to their smaller size, lower lipid content, and shorter life-cycle durations compared to Arctic congeners. An unprecedented warm and ice-free year in the Bering Sea saw an increase in small, low-lipid zooplankton, concurrent poor catches of pelagic fish, and low reproductive success and mass mortality at seabird colonies (Duffy-Anderson et al., 2019). Yet trait-based models have demonstrated that a C. finmarchicus-like life-history also brings a shorter generation time and faster population turnover, which may compensate or even enhance the transfer of energy to predators (Renaud et al., 2018). The ecological implications of changes in the Calanus complex remain uncertain. These will depend on the dynamics of sea-ice decline and associated phenology shifts, as well as the adaptability of, and interactions between, Arctic pelagic species.

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CONFLICT OF INTEREST

The authors declare no conflict of interest.

DATA AVAILABILITY STATEMENT

A list of the environmental data used within this study is provided in Table S1 within File S1. A list of the occurrence record data sets used within this study is provided in File S2. The data and R code that support the findings of this study are available from the UK Polar Data Centre at https://doi.org/10.5285/FC660BC3-09AB-4C1A-9D2A-40269 51872C8, reference number GB/NERC/BAS/PDC/01575.

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REFERENCES

Aarflot, J. M., Skjoldal, H. R., Dalpadado, P., & Skern-Mauritzen, M. (2018). Contribution of *Calanus* species to the mesozooplankton biomass in the Barents Sea. *ICES Journal of Marine Science*, 75(7), 2342–2354. https://doi.org/10.1093/icesjms/fsx221

- Aiello-Lammens, M. E., Boria, R. A., Radosavljevic, A., Vilela, B., & Anderson, R. P. (2015). spThin: An R package for spatial thinning of species occurrence records for use in ecological niche models. *Ecography*, 38(5), 541–545. https://doi.org/10.1111/ecog.01132
- Albouy-Boyer, S., Plourde, S., Pepin, P., Johnson, C. L., Lehoux, C., Galbraith, P. S., Hebert, D., Lazin, G., & Lafleur, C. (2016). Habitat modelling of key copepod species in the Northwest Atlantic Ocean based on the Atlantic Zone Monitoring Program. *Journal* of Plankton Research, 38(3), 589-603. https://doi.org/10.1093/ plankt/fbw020
- Allouche, O., Tsoar, A., & Kadmon, R. (2006). Assessing the accuracy of species distribution models: Prevalence, kappa and the true skill statistic (TSS). *Journal of Applied Ecology*, 43(6), 1223–1232. https://doi.org/10.1111/j.1365-2664.2006.01214.x
- Araújo, M. B., Anderson, R. P., Márcia Barbosa, A., Beale, C. M., Dormann,
 C. F., Early, R., Garcia, R. A., Guisan, A., Maiorano, L., Naimi, B.,
 O'Hara, R. B., Zimmermann, N. E., & Rahbek, C. (2019). Standards
 for distribution models in biodiversity assessments. *Science Advances*, 5(1). https://doi.org/10.1126/sciadv.aat4858
- Beaugrand, G. (2009). Decadal changes in climate and ecosystems in the North Atlantic Ocean and adjacent seas. *Deep-Sea Research Part II: Topical Studies in Oceanography*, 56(8–10), 656–673. https://doi. org/10.1016/j.dsr2.2008.12.022
- Beaugrand, G., Edwards, M., Brander, K., Luczak, C., & Ibanez, F. (2008). Causes and projections of abrupt climate-driven ecosystem shifts in the North Atlantic. *Ecology Letters*, 11(11), 1157–1168. https:// doi.org/10.1111/j.1461-0248.2008.01218.x
- Beaugrand, G., Luczak, C., & Edwards, M. (2009). Rapid biogeographical plankton shifts in the North Atlantic Ocean. *Global Change Biology*, 15(7), 1790–1803. https://doi. org/10.1111/j.1365-2486.2009.01848.x
- Beaugrand, G., Mackas, D., & Goberville, E. (2013). Applying the concept of the ecological niche and a macroecological approach to understand how climate influences zooplankton: Advantages, assumptions, limitations and requirements. *Progress in Oceanography*, 111, 75–90. https://doi.org/10.1016/j.pocean.2012.11.002
- Bonnet, D., Richardson, A., Harris, R., Hirst, A., Beaugrand, G., Edwards, M., Ceballos, S., Diekman, R., López-Urrutia, A., Valdes, L., Carlotti, F., Molinero, J. C., Weikert, H., Greve, W., Lucic, D., Albaina, A., Yahia, N. D., Umani, S. F., Miranda, A., ... Fernandez de Puelles, M. L. (2005). An overview of *Calanus helgolandicus* ecology in European waters. *Progress in Oceanography*, 65(1), 1–53. https://doi. org/10.1016/j.pocean.2005.02.002
- Botella, C., Joly, A., Monestiez, P., Bonnet, P., & Munoz, F. (2020). Bias in presence-only niche models related to sampling effort and species niches: Lessons for background point selection. *PLoS One*, *15*(5), e0232078. https://doi.org/10.1371/journal.pone.0232078
- Choquet, M., Hatlebakk, M., Dhanasiri, A. K. S., Kosobokova, K., Smolina, I., Soreide, J. E., Svensen, C., Melle, W., Kwaśniewski, S., Eian, K., Daase, M., Tverberg, V., Skreslet, S., Bucklin, A., & Hoarau, G. (2017). Genetics redraws pelagic biogeography of *Calanus. Biology Letters*, 13(12). https://doi.org/10.1098/rsbl.2017.0588
- Choquet, M., Kosobokova, K., Kwasniewski, S., Hatlebakk, M., Dhanasiri, A. K. S., Melle, W., Daase, M., Svensen, C., Søreide, J. E., & Hoarau, G. (2018). Can morphology reliably distinguish between the copepods Calanus finmarchicus and C-glacialis, or is DNA the only way? *Limnology and Oceanography-Methods*, 16(4), 237–252. https://doi. org/10.1002/lom3.10240
- Chust, G., Castellani, C., Licandro, P., Ibaibarriaga, L., Sagarminaga, Y., & Irigoien, X. (2014). Are *Calanus* spp. shifting poleward in the North Atlantic? A habitat modelling approach. *ICES Journal of Marine Science*, 71(2), 241–253. https://doi.org/10.1093/icesjms/fst147
- Daase, M., Falk-Petersen, S., Varpe, O., Darnis, G., Soreide, J. E., Wold, A., Leu, E., Berge, J., Philippe, B., & Fortier, L. (2013). Timing of reproductive events in the marine copepod *Calanus glacialis*: A

WILEY- 🚍 Global Change Biology

pan-Arctic perspective. Canadian Journal of Fisheries and Aquatic Sciences, 70(6), 871–884. https://doi.org/10.1139/cjfas-2012-0401

- Dalpadado, P., Arrigo, K. R., van Dijken, G. L., Skjoldal, H. R., Bagøien, E., Dolgov, A. V., Prokopchuk, I. P., & Sperfeld, E. (2020). Climate effects on temporal and spatial dynamics of phytoplankton and zooplankton in the Barents Sea. *Progress in Oceanography*, *185*, 102320. https://doi.org/10.1016/j.pocean.2020.102320
- Duffy, G. A., & Chown, S. L. (2017). Explicitly integrating a third dimension in marine species distribution modelling. *Marine Ecology Progress Series*, 564, 1–8. https://doi.org/10.3354/meps12011
- Duffy-Anderson, J. T., Stabeno, P., Andrews, A. G., Cieciel, K., Deary, A., Farley, E., Fugate, C., Harpold, C., Heintz, R., Kimmel, D., Kuletz, K., Lamb, J., Paquin, M., Porter, S., Rogers, L., Spear, A., & Yasumiishi, E. (2019). Responses of the northern Bering Sea and southeastern Bering Sea Pelagic ecosystems following record-breaking low winter sea ice. *Geophysical Research Letters*, 46(16), 9833–9842. https://doi. org/10.1029/2019gl083396
- Eriksen, E., Skjoldal, H. R., Gjosaeter, H., & Primicerio, R. (2017). Spatial and temporal changes in the Barents Sea pelagic compartment during the recent warming. *Progress in Oceanography*, 151, 206– 226. https://doi.org/10.1016/j.pocean.2016.12.009
- Ershova, E. A., Kosobokova, K. N., Banas, N. S., Ellingsen, I., Niehoff, B., Hildebrandt, N., & Hirche, H.-J. (2021). Sea ice decline drives biogeographical shifts of key *Calanus* species in the central Arctic Ocean. *Global Change Biology*, 27(10), 2128–2143. https://doi.org/10.1111/gcb.15562
- Falk-Petersen, S., Mayzaud, P., Kattner, G., & Sargent, J. R. (2009). Lipids and life strategy of Arctic Calanus. Marine Biology Research, 5(1), 18–39. https://doi.org/10.1080/17451000802512267
- Feng, Z. X., Ji, R. B., Ashjian, C., Campbell, R., & Zhang, J. L. (2018). Biogeographic responses of the copepod *Calanus glacialis* to a changing Arctic marine environment. *Global Change Biology*, 24(1), E159–E170. https://doi.org/10.1111/gcb.13890
- Feng, Z. X., Ji, R. B., Campbell, R. G., Ashjian, C. J., & Zhang, J. L. (2016). Early ice retreat and ocean warming may induce copepod biogeographic boundary shifts in the Arctic Ocean. *Journal* of Geophysical Research: Oceans, 121(8), 6137–6158. https://doi. org/10.1002/2016jc011784
- Fossheim, M., Primicerio, R., Johannesen, E., Ingvaldsen, R. B., Aschan, M. M., & Dolgov, A. V. (2015). Recent warming leads to a rapid borealization of fish communities in the Arctic. *Nature Climate Change*, 5(7), 673–677. https://doi.org/10.1038/nclimate2647
- Fourcade, Y., Engler, J. O., Roedder, D., & Secondi, J. (2014). Mapping species distributions with Maxent using a geographically biased sample of presence data: A performance assessment of methods for correcting sampling bias. *PLoS ONE*, 9(5), e97122. https://doi. org/10.1371/journal.pone.0097122
- Frainer, A., Primicerio, R., Kortsch, S., Aune, M., Dolgov, A. V., Fossheim, M., & Aschan, M. M. (2017). Climate-driven changes in functional biogeography of Arctic marine fish communities. *Proceedings of the National Academy of Sciences of the United States of America*, 114(46), 12202–12207. https://doi.org/10.1073/pnas.1706080114
- Hastings, R. A., Rutterford, L. A., Freer, J. J., Collins, R. A., Simpson, S. D., & Genner, M. J. (2020). Climate change drives poleward increases and equatorward declines in marine species. *Current Biology*, 30(8), 1572–1577. https://doi.org/10.1016/j.cub.2020.02.043
- Helaouet, P., & Beaugrand, G. (2007). Macroecology of Calanus finmarchicus and C-helgolandicus in the North Atlantic Ocean and adjacent seas. Marine Ecology Progress Series, 345, 147–165. https://doi. org/10.3354/meps06775
- Helaouet, P., & Beaugrand, G. (2009). Physiology, ecological niches and species distribution. *Ecosystems*, 12(8), 1235–1245. https://doi. org/10.1007/s10021-009-9261-5
- Helaouet, P., Beaugrand, G., & Reid, P. C. (2011). Macrophysiology of Calanus finmarchicus in the North Atlantic Ocean. Progress in Oceanography, 91(3), 217-228. https://doi.org/10.1016/j. pocean.2010.11.003

- Hirche, H. J., & Kosobokova, K. (2007). Distribution of Calanus finmarchicus in the northern North Atlantic and Arctic Ocean–Expatriation and potential colonization. Deep-Sea Research Part II: Topical Studies in Oceanography, 54(23–26), 2729–2747. https://doi.org/10.1016/ j.dsr2.2007.08.006
- Hirche, H. J., Meyer, U., & Niehoff, B. (1997). Egg production of Calanus finmarchicus: Effect of temperature, food and season. Marine Biology, 127(4), 609–620. https://doi.org/10.1007/s002270050051
- Hop, H., Vihtakari, M., Daase, M., Kwasniewski, S., Gluchowska, M., Lischka, S., Buchholz, F., & Falk-Petersen, S. (2019). Zooplankton in Kongsfjorden (1996–2016) in relation to climate change. In H. Hop & C. Wiencke (Eds.), *The ecosystem of Kongsfjorden, Svalbard* (pp. 229–300). Springer.
- Jarnevich, C. S., Stohlgren, T. J., Kumar, S., Morisette, J. T., & Holcombe, T. R. (2015). Caveats for correlative species distribution modeling. *Ecological Informatics*, 29, 6–15. https://doi.org/10.1016/ j.ecoinf.2015.06.007
- Ji, R., Ashjian, C. J., Campbell, R. G., Chen, C., Gao, G., Davis, C. S., Cowles, G. W., & Beardsley, R. C. (2012). Life history and biogeography of *Calanus* copepods in the Arctic Ocean: An individual-based modeling study. *Progress in Oceanography*, 96(1), 40–56. https://doi. org/10.1016/j.pocean.2011.10.001
- Kahru, M., Brotas, V., Manzano-Sarabia, M., & Mitchell, B. G. (2011). Are phytoplankton blooms occurring earlier in the Arctic? *Global Change Biology*, 17(4), 1733–1739. https://doi. org/10.1111/j.1365-2486.2010.02312.x
- Kimmel, D. G., Eisner, L. B., Wilson, M. T., & Duffy-Anderson, J. T. (2018). Copepod dynamics across warm and cold periods in the eastern Bering Sea: Implications for walleye pollock (*Gadus chalcogrammus*) and the Oscillating Control Hypothesis. *Fisheries Oceanography*, 27(2), 143–158. https://doi.org/10.1111/fog.12241
- Kortsch, S., Primicerio, R., Beuchel, F., Renaud, P. E., Rodrigues, J., Lonne, O. J., & Gulliksen, B. (2012). Climate-driven regime shifts in Arctic marine benthos. Proceedings of the National Academy of Sciences of the United States of America, 109(35), 14052–14057. https://doi. org/10.1073/pnas.1207509109
- Kortsch, S., Primicerio, R., Fossheim, M., Dolgov, A. V., & Aschan, M. (2015). Climate change alters the structure of arctic marine food webs due to poleward shifts of boreal generalists. *Proceedings of the Royal Society B: Biological Sciences*, 282(1814), 31–39. https:// doi.org/10.1098/rspb.2015.1546
- Kwasniewski, S., Gluchowska, M., Jakubas, D., Wojczulanis-Jakubas, K., Walkusz, W., Karnovsky, N., Blachowiak-Samolyk, K., Cisek, M., & Stempniewicz, L. (2010). The impact of different hydrographic conditions and zooplankton communities on provisioning Little Auks along the West coast of Spitsbergen. *Progress in Oceanography*, 87(1-4), 72-82. https://doi.org/10.1016/j.pocean.2010.06.004
- Ljungstrom, G., Langbehn, T. J., & Jorgensen, C. (2021). Light and energetics at seasonal extremes limit poleward range shifts. *Nature Climate Change*, 11(6), 530–536. https://doi.org/10.1038/s41558-021-01045-2

Mauchline, J. (1998). The biology of calanoid copepods. Academic Press.

- Melle, W., Runge, J., Head, E., Plourde, S., Castellani, C., Licandro, P., Pierson, J., Jonasdottir, S., Johnson, C., Broms, C., Debes, H., Falkenhaug, T., Gaard, E., Gislason, A., Heath, M., Niehoff, B., Nielsen, T. G., Pepin, P., Stenevik, E. K., & Chust, G. (2014). The North Atlantic Ocean as habitat for *Calanus finmarchicus*: Environmental factors and life history traits. *Progress in Oceanography*, 129, 244–284. https://doi.org/10.1016/j. pocean.2014.04.026
- Merow, C., Smith, M. J., & Silander Jr., J. A. (2013). A practical guide to MaxEnt for modeling species' distributions: What it does, and why inputs and settings matter. *Ecography*, 36(10), 1058–1069. https:// doi.org/10.1111/j.1600-0587.2013.07872.x
- Moller, E. F., & Nielsen, T. G. (2020). Borealization of Arctic zooplanktonsmaller and less fat zooplankton species in Disko Bay, Western

Greenland. Limnology and Oceanography, 65(6), 1175–1188. https://doi.org/10.1002/lno.11380

- Niehoff, B., Madsen, S. D., Hansen, B. W., & Nielsen, T. G. (2002). Reproductive cycles of three dominant Calanus species in Disko Bay, West Greenland. *Marine Biology*, 140(3), 567–576. https://doi. org/10.1007/s00227-001-0731-3
- Phillips, S., & Dudik, M. (2008). Modeling of species distributions with Maxent: New extensions and a comprehensive evaluation. *Ecography*, 31(2), 161–175. https://doi.org/10.1111/j.0906-7590. 2008.5203.x
- Phillips, S., Dudik, M., Elith, J., Graham, C. H., Lehmann, A., Leathwick, J., & Ferrier, S. (2009). Sample selection bias and presence-only distribution models: Implications for background and pseudoabsence data. *Ecological Applications*, 19(1), 181–197. https://doi. org/10.1890/07-2153.1
- Polyakov, I. V., Alkire, M. B., Bluhm, B. A., Brown, K. A., Carmack, E. C., Chierici, M., Danielson, S. L., Ellingsen, I., Ershova, E. A., Gårdfeldt, K., Ingvaldsen, R. B., Pnyushkov, A. V., Slagstad, D., & Wassmann, P. (2020). Borealization of the Arctic Ocean in response to anomalous advection from sub-Arctic seas. *Frontiers in Marine Science*, 7. https://doi.org/10.3389/fmars.2020.00491
- Polyakov, I. V., Pnyushkov, A. V., Alkire, M. B., Ashik, I. M., Baumann, T. M., Carmack, E. C., Goszczko, I., Guthrie, J., Ivanov, V. V., Kanzow, T., Krishfield, R., Kwok, R., Sundfjord, A., Morison, J., Rember, R., & Yulin, A. (2017). Greater role for Atlantic inflows on sea-ice loss in the Eurasian Basin of the Arctic Ocean. *Science*, *356*(6335), 285– 291. https://doi.org/10.1126/science.aai8204
- Rayner, N. A., Parker, D. E., Horton, E. B., Folland, C. K., Alexander, L. V., Rowell, D. P., & Kaplan, A. (2003). Global analyses of sea surface temperature, sea ice, and night marine air temperature since the late nineteenth century. *Journal of Geophysical Research: Atmospheres*, 108(D14). https://doi.org/10.1029/2002jd002670
- Record, N. R., Ji, R. B., Maps, F., Varpe, O., Runge, J. A., Petrik, C. M., & Johns, D. (2018). Copepod diapause and the biogeography of the marine lipidscape. *Journal of Biogeography*, 45(10), 2238–2251. https://doi.org/10.1111/jbi.13414
- Renaud, P. E., Daase, M., Banas, N. S., Gabrielsen, T. M., Søreide, J. E., Varpe, Ø., Cottier, F., Falk-Petersen, S., Halsband, C., Vogedes, D., Heggland, K., & Berge, J. (2018). Pelagic food-webs in a changing Arctic: A trait-based perspective suggests a mode of resilience. *ICES Journal of Marine Science*, 75(6), 1871–1881. https://doi. org/10.1093/icesjms/fsy063
- Reygondeau, G., & Beaugrand, G. (2011). Future climate-driven shifts in distribution of *Calanus finmarchicus*. *Global Change Biology*, 17(2), 756–766. https://doi.org/10.1111/j.1365-2486.2010.02310.x
- Ringuette, M., Fortier, L., Fortier, M., Runge, J. A., Bélanger, S., Larouche, P., Weslawski, J.-M., & Kwasniewski, S. (2002). Advanced recruitment and accelerated population development in Arctic calanoid copepods of the North Water. *Deep-Sea Research Part II: Topical Studies in Oceanography*, 49(22–23), 5081–5099. https://doi. org/10.1016/s0967-0645(02)00179-0
- Sainmont, J., Andersen, K. H., Varpe, O., & Visser, A. W. (2014). Capital versus income breeding in a seasonal environment. *The American Naturalist*, 184(4), 466–476. https://doi.org/10.1086/677926
- Slagstad, D., Ellingsen, I. H., & Wassmann, P. (2011). Evaluating primary and secondary production in an Arctic Ocean void of summer sea ice: An experimental simulation approach. *Progress in Oceanography*, 90(1–4), 117–131. https://doi.org/10.1016/j.pocean.2011.02.009
- Soberon, J., & Nakamura, M. (2009). Niches and distributional areas: Concepts, methods, and assumptions. Proceedings of the National Academy of Sciences of the United States of America, 106, 19644– 19650. https://doi.org/10.1073/pnas.0901637106
- Song, H. J., Ji, R. B., Jin, M. B., Li, Y., Feng, Z. X., Varpe, O., & Davis, C. S. (2021). Strong and regionally distinct links between ice-retreat timing and phytoplankton production in the Arctic Ocean. *Limnology*

and Oceanography, 66(6), 2498-2508. https://doi.org/10.1002/ Ino.11768

Global Change Biology -WILF

- Soreide, J. E., Leu, E., Berge, J., Graeve, M., & Falk-Petersen, S. (2010). Timing of blooms, algal food quality and *Calanus* glacialis reproduction and growth in a changing Arctic. *Global Change Biology*, 16(11), 3154-3163. https://doi. org/10.1111/j.1365-2486.2010.02175.x
- Strand, E., Bagoien, E., Edwards, M., Broms, C., & Klevjer, T. (2020). Spatial distributions and seasonality of four *Calanus* species in the Northeast Atlantic. *Progress in Oceanography*, 185, 102344. https:// doi.org/10.1016/j.pocean.2020.102344
- Tarling, G. A., Freer, J. J., Banas, N. S., Belcher, A., Blackwell, M., Castellani, C., Cook, K., Cottier, F., Daase, M., Johnson, M., Last, K., Lindeque, P., Mayor, D., Mitchell, E., Parry, H., Stowasser, G., & Wootton, M. (in press). Has a key boreal *Calanus* copepod species become an Arctic resident? Evidence and implications for Arctic food-webs. *Ambio*.
- Tsubouchi, T., Våge, K., Hansen, B., Larsen, K. M. H., Østerhus, S., Johnson, C., Jónsson, S., & Valdimarsson, H. (2021). Increased ocean heat transport into the Nordic Seas and Arctic Ocean over the period 1993–2016. *Nature Climate Change*, 11(1), 21–26. https:// doi.org/10.1038/s41558-020-00941-3
- Valavi, R., Elith, J., Lahoz-Monfort, J. J., & Guillera-Arroita, G. (2019). blockCV: An r package for generating spatially or environmentally separated folds for k-fold cross-validation of species distribution models. *Methods in Ecology and Evolution*, 10(2), 225–232. https:// doi.org/10.1111/2041-210x.13107
- Vignali, S., Barras, A. G., Arlettaz, R., & Braunisch, V. (2020). SDMtune: An R package to tune and evaluate species distribution models. *Ecology and Evolution*, 10(20), 11488–11506. https://doi.org/10.1002/ece3.6786
- Vihtakari, M., Welcker, J., Moe, B., Chastel, O., Tartu, S., Hop, H., Bech, C., Descamps, S., & Gabrielsen, G. W. (2018). Black-legged kittiwakes as messengers of Atlantification in the Arctic. *Scientific Reports*, 8. https://doi.org/10.1038/s41598-017-19118-8
- Vilgrain, L., Maps, F., Picheral, M., Babin, M., Aubry, C., Irisson, J. O., & Ayata, S. D. (2021). Trait-based approach using in situ copepod images reveals contrasting ecological patterns across an Arctic ice melt zone. *Limnology and Oceanography*, 66(4), 1155–1167. https:// doi.org/10.1002/lno.11672
- Villarino, E., Chust, G., Licandro, P., Butenschon, M., Ibaibarriaga, L., Larranaga, A., & Irigoien, X. (2015). Modelling the future biogeography of North Atlantic zooplankton communities in response to climate change. *Marine Ecology Progress Series*, 531, 121–142. https:// doi.org/10.3354/meps11299
- Walsh, J. E., Chapman, W. L., Fetterer, F., & Stewart, J. S. (2019). Gridded monthly sea ice extent and concentration, 1850 onward, version 2. NSIDC: National Snow and Ice Data Centre.
- Wassmann, P., Duarte, C. M., Agusti, S., & Sejr, M. K. (2011). Footprints of climate change in the Arctic marine ecosystem. *Global Change Biology*, 17(2), 1235–1249. https://doi. org/10.1111/j.1365-2486.2010.02311.x
- Wassmann, P., Reigstad, M., Haug, T., Rudels, B., Carroll, M. L., Hop, H., Gabrielsen, G. W., Falk-Petersen, S., Denisenko, S. G., Arashkevich, E., Slagstad, D., & Pavlova, O. (2006). Food webs and carbon flux in the Barents Sea. *Progress in Oceanography*, 71(2–4), 232–287. https://doi.org/10.1016/j.pocean.2006.10.003
- Weydmann, A., Carstensen, J., Goszczko, I., Dmoch, K., Olszewska, A., & Kwasniewski, S. (2014). Shift towards the dominance of boreal species in the Arctic: Inter-annual and spatial zooplankton variability in the West Spitsbergen Current. *Marine Ecology Progress Series*, 501, 41–52. https://doi.org/10.3354/meps10694
- Weydmann, A., Walczowski, W., Carstensen, J., & Kwasniewski, S. (2018). Warming of Subarctic waters accelerates development of a key marine zooplankton Calanus finmarchicus. Global Change Biology, 24(1), 172–183. https://doi.org/10.1111/gcb.13864

-WILEY- 🚍 Global Change Biology

Wilson, R. J., Heath, M. R., & Speirs, D. C. (2016). Spatial Modeling of Calanus finmarchicus and Calanus helgolandicus: Parameter differences explain differences in biogeography. Frontiers in Marine Science, 3. https://doi.org/10.3389/fmars.2016.00157

SUPPORTING INFORMATION

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