

1 **Abstract**

2 Ample evidence has indicated shifts in distribution of fish populations in response
3 to environmental stress. However, most studies focused at the whole population scale. This
4 neglects the spatial dynamics between groups of different body size (body size groups), that
5 fundamentally shapes the spatial structure of a population. Here, we explored the
6 mechanisms that modulate spatial dynamics of body size groups, and applied our analyses
7 to three North Sea fish populations which experienced severe declines in biomass from 1977
8 to 2019: Atlantic cod (*Gadus morhua*), haddock (*Melanogrammus aeglefinus*), and whiting
9 (*Merlangius merlangius*). All three populations exhibited strong declines in the overlapped
10 area between body size groups in winter over 43 years, yet their mechanisms differed. These
11 declines were either due to (1) different magnitudes of contraction of the distribution area
12 of body size groups; and/or (2) different speeds and directions of spatial shift among various
13 body size groups, both increasing spatial segregation within populations. These patterns
14 were either associated with ocean warming, and/or declining population biomass, and such
15 associations often varied according to distinct body size groups. Our analytical approach
16 provides a powerful tool for identifying vulnerable populations under environmental stress
17 and can be generalized to study a variety of size/age structured populations at various
18 ecosystem types.

19 **Keywords**

20 biogeography, marine ecology, ocean warming, population spatial structure, population
21 spatial shift.

22 **Introduction**

23 Many marine fish populations have undergone significant shifts in their spatial
24 distributions over the past decades, largely related to ocean warming and declining
25 population size (Perry et al. 2005, Sunday et al. 2015). Most of these studies focused at the
26 whole population scale; however, several lines of evidence have suggested that the spatial
27 shift varies in magnitude and direction for different body size groups within a population
28 (hereafter, body size groups) (Bell et al. 2015, Barbeaux and Hollowed 2018, Frank et al.
29 2018, Yang et al. 2019, Li et al. 2022). For instance, the distribution of the middle size
30 groups of some fish populations in the Eastern Bering Sea shifted at a greater speed in
31 warm seasons, compared to groups of smaller or larger body sizes (Barbeaux and Hollowed
32 2018). Another study across North Pacific, North Atlantic, and South Atlantic suggested
33 that the distribution of large size groups within some fish populations shifted deeper, as a
34 result of size-selective fishing at shallower water (Frank et al. 2018). These size-dependent
35 shifts in distribution are likely to reduce overlapped areas between body size groups, that is,
36 increase the spatial segregation within populations. However, temporal changes in spatial
37 segregation (i.e., overlapped area) between body size groups have not been quantified for
38 real-world populations, despite earlier efforts from theoretical approaches (Hughes and
39 Grand 2000).

40 Changes in spatial segregation between body size groups of a population have
41 various consequences on population dynamics. On one hand, a population with high spatial
42 segregation between body size groups can reduce the stress from predation and competition.
43 On the other hand, a population with highly segregated size group is more vulnerable to
44 local perturbations. These perturbations include size-selective fishing, size-selective
45 predation, or unfavorable habitat conditions for certain body size groups (Hsieh et al. 2010b).
46 These perturbations can change the abundance of certain body size groups, which in turn
47 alter the demographic structure and spatial structure of the population (Tao et al. 2021).

More generally, changes in the spatial structure of a marine population can influence life history and demographic variations, which potentially affect its resilience to perturbations (Ciannelli et al. 2013).

What are the potential mechanisms shaping spatial segregation between body size groups of a population? Within a population, the distribution area of each body size group, and the distance between their abundance-weighted centers of distribution area (hereafter, centers of abundance), determine the overlapped area between them. On the one hand, when the distribution areas of two body size groups contract, their overlapped area declines, provided that their centers of abundance are fixed. On the other hand, elongated distance between the centers of abundance reduces the overlapped area between body size groups, provided that their areas of distribution are fixed.

Ocean warming and population decline potentially impact the area of distribution and the center of abundance of body size groups (Barnett et al. 2017, Orio et al. 2017). These impacts are likely size-specific. For example, earlier studies showed that ocean warming and fishing altered the abundance of body size groups at various extents (Barnett et al. 2017, Orio et al. 2017). Such size-specific changes in abundance could lead to differential changes in their area of distribution, based on abundance-distribution relationships and density-dependent habitat selection (MacCall 1990, Fisher and Frank 2004, Thorson et al. 2016). In addition, previous findings suggest that ocean warming and fishing contributed to size-specific shift in spatial distribution (Barbeaux and Hollowed 2018, Frank et al. 2018). This is due to thermal tolerance, food requirements, spatial constraints, and mobility that vary with body sizes within a population (Dahlke et al. 2020, Ciannelli et al. 2022). Depending on the original positions of the center of abundance, the size-specific shift could increase the distance between their distribution. Linking body size-specific distribution response to ocean warming and population decline is key to understanding the

mechanisms behind the changes in spatial segregation between body size groups of a population.

In this study, we quantified and explored the mechanisms of changes in spatial segregation over time between body size groups within fish populations. We asked the following question: did the overlapped area of body size groups within populations decline over time, and what are the mechanisms behind? We studied fish populations in the North Sea, a global warming hotspot that has experienced rising sea surface temperature over the past decades (Hobday and Pecl 2014). Particularly, we focused on those fish populations that are ecologically and economically important and experienced large geographical redistribution over the past century (Huserbråten et al. 2018), including Atlantic cod (*Gadus morhua*), haddock (*Melanogrammus aeglefinus*), and whiting (*Merlangius merlangius*). The total biomass of these populations has declined since 1980s with slow recovery in recent years (Engelhard et al. 2014). Therefore, these populations are prone to distribution area contraction and fragmentation. We analyzed their spatial dynamics using 43-year (1977-2019) winter survey data. We hypothesized that the body size groups of these populations became spatially more segregated over time, which was associated with contracted distribution area of body size groups, and/or elongated distance between centers of abundance of these groups. In addition, these changes were caused by body size-specific responses to environmental stress, including ocean warming and population decline.

Materials and Methods

Fish populations and survey data

The North Sea is a European epicontinental sea connected to the Atlantic Ocean. The north part of the North Sea is deeper, colder with higher salinity, while the south part is warmer, shallower with higher primary productivity. The North Sea has experienced

rising sea temperature and intensive fishing activities over the past decades (Murgier et al. 2021). Fishing has been more intensive in the south part of the North Sea (Engelhard et al. 2014).

We focused on three fish populations in the North Sea: Atlantic cod, haddock and whiting. They belong to the *Gadidae* family and are demersal populations which live just above the bottom of the sea (for life histories of three populations see **Table S1**). They have spawning migration in winter (Tobin et al. 2010, González-Irusta and Wright 2016). Evidence have shown that North Sea Atlantic cod is a metapopulation composed of three subpopulations: South, Northwest, and Viking (ICES 2020).

We obtained the survey data of three target populations from the online database of the International Bottom Trawl Survey (IBTS) of International Council for Exploitation of the Sea (ICES) (<https://data.ices.dk/>). This survey follows a stratified sampling on survey rectangles of 1° longitude \times 0.5° latitude. The dataset is in the form of catch per unit effort (CPUE) per body size (in 10mm unit) for each rectangle and year-quarter. We extracted the winter data (January to February) between year 1977 and 2019 as our study period, because fishing gear was not standardized until 1977. We did not analyze the summer data, because the survey period is relatively short (starting from 1991), and that seasonal differences in the spatial structure is out of the scope of our study.

Define body size groups within populations

We examined the spatial dynamics at the body size level. We followed the most common approach for body size grouping through dividing a population into equal body size bins (Barbeaux and Hollowed 2018, Li et al. 2019, Yang et al. 2019). We first summed the CPUE for each body size bin (in 1mm unit) over time and survey rectangle, to derive body size distribution. As the distribution was right-skewed, we removed individuals below

5% and above 85% quantile to avoid extremely low abundance at both ends. Then, we divided the body size distribution into equal-interval body size groups.

We tested different body size group number (10, 15, and 20 groups) to see how it influenced the value of spatial dynamics. While higher size group number gave higher precision, the spatial dynamics did not differ with group number (**Table S2**). We thus reported the results with 20 body size groups in the main text. We did not use group number higher than 20, otherwise would leads to too few individuals for largest and smallest body size groups; this could raise uncertainty of the results.

Deriving fixed number of body size group for each population leads to wider body size bins for larger populations, and narrower body size bins for smaller populations. To confirm the temporal dynamics of spatial overlap within populations, we alternatively derived body size groups by using fixed bin width for all three populations (e.g., 5 cm).

We also examined the changes in the overlapped area over time between life stages within populations as a preliminary test. To do so, we grouped each population into juvenile and adult, based on the body size at 50% maturity (**Table S1**).

We did not analyze the spatial structure using age groups because existing age-specific data did not distinguish age groups older than six years. Thus, spatial dynamics calculated using this dataset would neglect the dynamics between older groups. In addition, body size interval differed from one age to another due to non-linear age-size relationships. Because the results from age group or size group are not comparable, we reported only spatial structure between body size groups in this work.

Spatial structure indices

To explore the temporal changes in the spatial distribution of body size groups within populations, we calculated the following indices for each survey year: 1) area of

distribution of each body size group, 2) center of abundance of each body size group, 3) overlapped area between pairs of body size groups, and 4) distance between centers of abundance of pairs of body size groups. There is a total of 190 (C_2^{20}) pairs of body size groups between 20 body size groups within a population.

The area of distribution of each body size group is the proportion of occupied area at any given year, over the maximum occupied area of the same body size group over the study period. This standardized measure accounts for variations in the occupied area between different body size groups. Thus, this measure allows us to directly comparing distribution area between different body size groups. The occupied area of a body size group at a given year is defined as the number of survey rectangles where the CPUE of this group is greater than zero. Therefore, the area of distribution of body size group i at year t is $N_{i,t}/Max(N_i)$, where $N_{i,t}$ is the number of rectangles with the non-zero CPUE of body size group i at year t , and $Max(N_i)$ is the maximum number of rectangles of body size group i over the study period.

Center of abundance is CPUE-weighted center of occupied area for each body size group. For body size group i at year t , the center of abundance in longitude is $CEN_{i,t,lon} = \sum_{r=1}^N CPUE_{i,r,t} \times lon_r / \sum_{r=1}^N CPUE_{i,r,t}$, where lon_r is the longitudinal center of rectangle r , and N is the number of survey rectangles where the CPUE of the whole population is greater than zero. Similarly, the center of abundance in latitude $CEN_{i,t,lat} = \sum_{r=1}^N CPUE_{i,r} \times lat_r / \sum_{r=1}^N CPUE_{i,r}$, where lat_r is the latitudinal center of rectangle r .

The overlapped area for a given pair of body size groups are indicated by union overlapped area, and partial overlapped area. Union overlapped area is the proportion of co-occupied area, over the area where either of the body size group occupies. For body size group i and j at year t , the union overlapped area is $N_{intersect,i,j,t}/N_{union,i,j,t}$, where $N_{intersect,i,j,t}$ is the number of rectangles where both body size group i and j have CPUE

greater than zero at year t , and $N_{union,i,j,t}$ is the number of rectangles where either body size group i or j has CPUE greater than zero at year t . Partial overlapped area is proportion of co-occupied area over the occupied area of each body size group of the pair and then taken average. For body size group i and j at time t , partial overlapped area is $0.5 \times (N_{intersect,i,j,t}/N_{i,t} + N_{intersect,i,j,t}/N_{j,t})$. The concepts of distributional overlap has been used in inter-species co-occurrence at the community level (Griffith et al. 2018, Carroll et al. 2019), but not at the body size level.

The distance between centers of abundance is the longitudinal or latitudinal distance between centers of occupied area of a pair of body size groups. For body size group i and j at year t , the distance between centers of abundance in longitude is $|CEN_{i,t,lon} - CEN_{j,t,lon}|$, while the distance between centers of abundance in latitude is $|CEN_{i,t,lat} - CEN_{j,t,lat}|$.

Atlantic cod has three subpopulations in the North Sea (ICES 2020). Thus, we calculated the spatial structure of Atlantic cod at both the regional scale, as well as at the spatial scale concerning each subpopulation.

Population biomass decline

We used the estimates of yearly total stock biomass from the ICES stock assessment (ICES 2016, 2018) as a proxy for population depletion level (Zhou et al. 2017). Total stock biomass showed a declining trend from 1977 to 2019 for all three populations (**Fig. S1**).

Ocean warming

We used sea bottom temperature as an indicator of ocean warming, because all three target populations are demersal species. We obtained the sea bottom temperature of the CTD stations across the North Sea region from the ICES online database. We obtained the yearly winter sea bottom temperature at the North Sea region by averaging the

measurements from all CTD stations at each year. Sea bottom temperature in the North Sea exhibited a temporal increase from 1977 to 2019 (**Fig. S1**).

Statistical Analysis

The statistical models were constructed separately for each target population. We applied a four-step analysis are as follows:

- (1) We used linear mixed-effects models to examine the temporal trends in the overlapped area of pairs of body size groups. Overlapped area of 190 paired groups was included as the response variable (not averaged but as 190 measures). Overlapped area is count-based percentage data. Therefore, it was logit-transformed before model fitting for better homoscedasticity. Survey year was normalized and fitted as a fixed effect. The id of paired groups nested within the survey year was fitted as a random effect. This allows for random intercept and slope for each pair of body size group. We repeated the same analysis to examine the temporal trends in the distance between centers of abundance for pairs of body size groups, without transforming the response variable. Then, we repeated the analysis to examine the temporal trends in the area of distribution of body size groups. Area of distribution is continuous proportional data. Thus, it was logit-transformed before fitting. The id of body size group nested within survey year was fitted as random effect.
- (2) Then, we constructed a multiple regression model to test the relative importance of area of distribution and distance between centers of abundance on overlapped area. We regressed yearly mean of overlapped area (mean of 190 paired groups) against yearly mean of area of distribution (mean of 20 body size groups), and yearly mean of distance between centers of abundance in latitude and longitude (mean of 190 paired groups). This resulted in 43 data points (43 years) in each model. To account for serial correlation in time series data, we included the temporal autocorrelation of one-step time lag (AR1).

For the initial model, we included an interaction term between area of distribution and the distance between centers of abundance. As none of the interaction term was significant for neither population, we removed the interaction term from the initial model. The final model wrote:

Yearly mean of overlapped area across paired groups $\sim \beta_1$ Yearly mean of area of distribution across body size groups + β_2 Yearly mean of distance between centers of abundance in longitude across paired groups + β_3 Yearly mean of distance between centers of abundance in latitude across paired groups + AR1,

where β represents the fixed effects coefficients. All explanatory variables were normalized before fitting. All the explanatory variables had variance inflation factors < 6, suggesting no noticeable multicollinearity. We extracted the fixed effects coefficients with 95% confidence intervals to represent the relative importance of each explanatory variable.

(3) For each body size group, we evaluated the temporal trends in the area of distribution and center of abundance. To do so, we fitted a simple linear regression model for each body size group separately. We included the area of distribution (logit-transformed), or center of abundance of a body size group, as the response variable. We included survey year as the explanatory variable. We used the slope coefficient to indicate the rate of change in the area of distribution or center of abundance. Then, we examined how the rate of change varied with body size. To do so, we used nonparametric loess regression models. We included the rate of change in area of distribution, or center of abundance, as the response variable. We included body size group as a continuous explanatory variable.

(4) Finally, we examined whether the overlapped area was influenced by sea bottom temperature (Temperature) and total stock biomass (Biomass). In addition, we examined

how the effects differed between body size groups within each fish population. We hypothesized that the overlapped area is shaped by the area of distribution, and center of abundance of each body size group. Therefore, we examined the effects of Temperature and Biomass on these two variables. Temperature and Biomass are highly colinear for three populations. Thus, we tested their effects using separate models. In addition, we included AR1 in the model to account for the temporal autocorrelation. The four full models were:

- i. Yearly area of distribution of each body size group $\sim \beta_1$ body size group id \times yearly Temperature + β_2 CPUE + AR1, and
- ii. Yearly area of distribution of each body size group $\sim \beta_1$ body size group id \times Biomass + β_2 CPUE + AR1, and
- iii. Yearly center of abundance of each body size group in longitude or latitude $\sim \beta_1$ body size group id \times Yearly Temperature + AR1, and
- iv. Yearly center of abundance of each body size group in longitude or latitude $\sim \beta_1$ body size group id \times Yearly Biomass + AR1,

where log-transformed CPUE of each body size group was included as a covariate to account for abundance-distribution relationships. From each full model, we performed a backward stepwise model selection. We derived the most parsimonious model based on AIC and R^2 values.

We performed linear mixed-effects models using the function *lmer* from the *lme4* package. *P-values* were extracted using *lmerTest* package. We extracted Conditional R^2 (variance explained by both fixed and random effects) from the function *r.squaredGLMM* of *MuMIn* package. We performed the loess regression model with the *geom_smooth* function of *ggplot2* package. We further used heatmaps to visualize the differences in the temporal trends of overlapped area between each pair of size groups.

Results and Discussion

Temporal decline in overlapped area between body size groups

For all three populations between 1977 and 2019, the overlapped area between pairs of body size groups declined; that is, the spatial segregation increased between 20 body size groups (**Fig. 1**). The declining trends were significant regardless of the number of size groups we defined for each population (from 10 to 20 size groups, see **Table S2**), or fixed size bin width (e.g., 5cm, **Fig S2**). In addition, the declining patterns were observed for each subpopulation of Atlantic cod (South, Northwest, and Viking) (**Table S3**), suggesting a universal declining spatial overlap for the Atlantic cod metapopulation.

For Atlantic cod, the decline in spatial overlap was strong between small groups, between large groups, and between small and large groups (**Fig. S3**). Supporting these results, we observed clear declines over time in the number of co-occupied survey rectangles between juvenile and adult stages (**Fig. S4**). In contrast, for haddock, the decline in spatial overlap occurred only between small size groups (**Fig. S3**). Similarly, whiting showed declining spatial overlap between smaller groups, but increasing spatial overlap between larger groups (**Fig. S3**). The lack of changes in the spatial overlap between small and large groups, for both haddock and whiting, explained why the changes in co-occupied survey rectangles between juvenile and adult stages are less drastic compared to Atlantic cod (**Fig. S5 – S6**).

Contraction of the area of distribution of body size groups

One mechanism of spatial segregation between body size groups over time was related to the contraction of their distribution area, driven by rising sea temperature and/or population biomass decline. This mechanism was strongest in Atlantic cod and haddock.

The mechanism was weaker in whiting, which exhibited contraction of distribution area for smaller groups but expansion for larger groups over time.

Particularly, for Atlantic cod and haddock, the mean distribution area across body size groups declined over time (**Fig. 2a-b**). The mean distribution area was positively associated with the mean overlapped area across pairs of body size groups (**Fig. 3a-b**). In addition, total stock biomass positively contributed to the distribution area of each body size group (**Fig. 4a-b**). These results suggest that declining total stock biomass over time for these two populations (**Fig. S1**) led to contracted distribution area of body size groups, which in turn decreased their overlapped area. Particularly, during the latter years with lower stock biomass, larger size groups of Atlantic cod contracted their distribution areas in a greater rate than smaller groups ($P < 0.0001$ for an interactive term of total stock biomass \times body size group, **Fig. 4a, Table S4**). This pattern implies a greater removal of larger groups under intensive fishing exploitation time period (Horwood et al. 2006, Hsieh et al. 2010a). The positive association between population biomass and the area of distribution of body size groups agrees with the positive relationship observed at the whole population level of many fish species, as a result of density-dependent habitat selection (Fretwell and Lucas 1970, MacCall 1990, Fisher and Frank 2004, Thorson et al. 2016). Our finding is also supported by earlier evidence that during low abundance years, the area of distribution of age-1 and age-2 North Sea cod contracted to less than half of that available, towards habitats that have near-optimal bottom temperatures (Blanchard et al. 2005).

Overfishing is a potential main reason for biomass decline and spatial segregation within the Atlantic cod population. However, we did not examine the direct impact of fishing activity (i.e., fishing mortality) on spatial dynamics of Atlantic cod. It is because Atlantic cod is categorized as overexploited species, and its biomass recovers very slowly even after relaxing the fishing pressure since 1990s (Köster et al. 2014). Thus, instantaneous

fishing mortality measured at each year does not reflect the long-term impacts of fishing on the biomass and spatial structure of the population. Thus, in this study, we used estimated total stock biomass as the indicator of population depletion level (Froese et al. 2017) rather than fishing mortality, as a proxy to examine long-term fishing impacts on population spatial dynamics.

In contrast to Atlantic cod and haddock, whiting did not have a significant decline in the mean distribution area across body size groups (**Fig. 2c**). It was because larger groups expanded their distribution area while smaller groups contracted their distribution area over time (**Fig. 4c**). However, the mean area of distribution across body size groups was still positively related to their overlapped area (**Fig. 3c**).

In addition to the effect of population biomass decline, ocean warming also impacted the distribution area of body size groups, and the impacts varied among populations. For haddock, sea bottom temperature negatively explained the area of distribution of all body size groups (slope coefficient \pm standard error = -0.012 ± 0.004 , $P < 0.005$, **Fig. 4b**, **Table S4**). That is, the rising temperature over the study period contributed to the contraction of the distribution area of all body size groups, which then reduced the overlapped area between them (**Fig. 3b**). In contrast, for whiting, rising sea temperature resulted in the contraction of the distribution area of smaller size groups, but expansion of distribution area of large sizes groups ($P < 0.01$ for the interactive term of sea bottom temperature \times body size group, **Fig. 4c**). The differential responses between smaller and larger groups explains the lack of temporal patterns in the mean area of distribution across body size groups of whiting (**Fig. 2c**). In contrast to haddock and whiting, the distribution area of body size groups of Atlantic cod was determined by the population biomass but not by the sea bottom temperature (**Fig 4a**).

We speculate that the differences between haddock and whiting, in their distribution area response to ocean warming, may be due to their prey types. Haddock, regardless of body size, mainly feeds on benthic organisms which are spatially restricted under environmental changes (Schückel et al. 2010). In contrast, whiting is one of the top marine predators feeding on fishes, such as Norway pout, sandeel and sprat (Hislop et al. 1991). These fish prey have higher dispersal potential than benthic organisms under environmental changes, and thus could lead to the expansion of distribution for adult whiting that followed their prey. This is supported by otolith microchemistry analysis, showing that adult whiting can travel long distances (>500 km) to faraway spawning areas (Tobin et al. 2010). Whereas, contrary to larger size whiting, the distribution area of small size whiting contracted over time (**Fig 4c**). These observations support the notion that larger groups of some fish populations can be resistant to adverse conditions related to warming, and could have better knowledge and higher mobility moving to the optimal foraging and spawning grounds (Hsieh et al. 2010a).

Haddock and whiting have shifted northward since 1977, and the shift of whiting was correlated with warming (Perry et al. 2005). If some fishes have shifted outside of the North Sea, then the population biomass within the North Sea may reduce, leading to contraction in the distribution area and then spatial segregation between body size groups. Nevertheless, the spatial overlap indices used in our study is not sensitive to the spatial boundary of populations. This is because the indices are calculated based on the ratio of co-occupied area over occupied area by each body size group. Thus, these indices reveal the temporal variations in the degree of spatial overlap within the region analyzed in this study.

Distance increased between the centers of abundance between body size groups

In addition to the area of distribution, we hypothesized that the overlapped area between body size groups were negatively associated with their distance between the centers of abundance. In addition, such pattern was due to body size-specific shift in the centers of abundance, responding to rising sea temperature or population biomass decline. We found that whiting was the only population that exhibited this mechanism. In contrast, the spatial overlap within Atlantic cod and haddock was mainly determined by the area of distribution of body size groups.

Particularly, whiting showed temporal increases in the mean distance between centers of abundance across pairs of body size groups in longitude and latitude (**Fig. 2 f, i**). In addition, the mean distance was negatively associated with the overlapped area across paired groups (**Fig. 3c**). These results suggest that an increase in the distance contributed to a decline in the overlap between body size groups. The center of abundance of larger whiting shifted westward, while smaller groups shifted eastward ($P < 0.005$ for an interactive term of sea bottom temperature \times body size group id, **Table S4, Fig. 4f**). Therefore, depending on the original position of distributions, the body size-varying shift in the centers of abundance may have increased the distance between body size groups, hence reducing their overlapped area.

In contrast to whiting, for Atlantic cod and haddock, the distance between the centers of abundance across paired groups did not significantly explain the overlapped area (**Fig. 3a-b**). However, both populations showed an increase in the distance between centers of abundance (except for Atlantic cod at the longitudinal distance) (**Fig 2d-e, 2g-h**). These results suggest that the changes in the distance were too weak to influence the overlapped area between body size groups. Instead, the contraction of area of distribution of body size groups was the main driver for the spatial segregation under lower population biomass for Atlantic cod and haddock (**Fig 3a-b**). Interestingly, for haddock, the center of abundance of

larger groups at latitudinal direction was more negatively associated with the sea bottom temperature, compared to smaller haddock ($P < 0.01$ for an interactive term of sea bottom temperature \times body size group id, **Fig. 4h, Table S4**). Consequently, the centers of abundance of all size groups shifted northward in response to higher temperature, but larger size groups shifted faster than smaller size groups. Such different magnitudes of shift of the centers of abundance of body size groups in response to warming may have led to increased distance between their distributions for haddock.

Implications

While all the populations examined in this study demonstrated increased spatial segregation between body size groups over time, the underlying spatial dynamics of body size groups (i.e., area of distribution and center of abundance) and driving forces (i.e., ocean warming and population biomass decline) differed among the three studied populations (**Fig. 5**). These results have important implications for exploring the differences between populations in their physiological and biogeographic traits at the body size level. For example, body size groups within a population can exhibit different niches (e.g., thermal tolerance, food requirements (Ciannelli et al. 2013)). What drives different spatial responses among populations depends on the extent to which the niches of body size groups overlap. For example, populations with stronger or weaker niche preferences between body size groups may respond differently to disturbances such as climatic or anthropogenic stress (Tao et al. 2021).

For Atlantic cod and haddock, the contraction of the distribution area of body size groups was the main driver for the spatial segregation among body size groups over time. This finding has important implications to identify populations at risk of increased spatial segregation at body size group level. For example, both highly migratory pelagic predators (e.g., tuna, billfish) (Worm and Tittensor 2011) and species living in regional seas (e.g.,

Monterey Spanish mackerel at the coast of California (Collette and Russo 1984) and yellowtail flounder around Newfoundland (Brodie et al. 1998)) have shown contraction of their distribution area over the past decades. Although the contractions of distribution area were observed at the population level, these patterns may apply to the finer level of body size group. Furthermore, global projections estimated that the biomass of 77% of exploited fishes and invertebrates will decrease when high-temperature extreme will occur (Cheung et al. 2021). These pieces of evidence imply that many fish populations may have exhibited spatial segregation between body size groups, especially for those underwent reduced population biomass and contracted area of distribution, and for those living in climate-unstable regions. Large distributional shift may also reduce population biomass and distributional area at the original habitats. For example, in the North Sea, nearly two-third of fish species have shifted northward or deeper between 1977 and 2001 (Perry et al. 2005). Further investigations on these species which are “on the move” in the North Sea and beyond can help identify the state of the art of spatial dynamics within these populations, and to examine the spatial mechanisms and drivers for these vulnerable populations. These results are helpful to prioritize management and conservation efforts.

The ecological consequences of spatial segregation between body size groups of a population needs further investigation. While population growth may increase due to weakened cannibalism and competition, spatial segregation of size groups may increase the vulnerability of demographic structure to local perturbations. This merits future research to investigate the net effects of within-population spatial segregation on population dynamics and stability.

Conclusion

Recently, increasing evidence on aquatic and terrestrial populations has shown that the shift in spatial distribution varied between life stages or body size under environmental change (Bell et al. 2015, Máliš et al. 2016, Fei et al. 2017, Barbeaux and Hollowed 2018, Frank et al. 2018, Yang et al. 2019). However, it remains unclear to what extent different size groups within populations has segregated from each other over time. We develop a new analytical approach to deepen the understanding of spatial dynamics within populations under global environmental stress. This approach can be applied to populations at various terrestrial and aquatic ecosystems globally, to identify vulnerable populations under environmental stress. This approach also allows us to uncover the mechanisms of spatial segregation within populations, which have profound consequences in demographic connectivity and population stability.

Data and availability statements:

All raw data that support the findings of this study are publicly available. Fish survey data and sea bottom temperature are available at the ICES data portal <https://data.ices.dk/>. Total stock biomass is available from ICES stock assessment reports (ICES 2016, 2018). The codes needed to replicate the analyses presented in this paper will be available at online repository.

References

- Barbeaux, S. J., and A. B. Hollowed. 2018. Ontogeny matters: Climate variability and effects on fish distribution in the eastern Bering Sea. *Fisheries Oceanography* 27:1–15.
- Barnett, L. A. K., T. A. Branch, R. A. Ranasinghe, and T. E. Essington. 2017. Old-Growth Fishes Become Scarce under Fishing. *Current Biology* 27:2843-2848.e2.

465 Bell, R. J., D. E. Richardson, J. A. Hare, P. D. Lynch, and P. S. Fratantoni. 2015. Disentangling
 466 the effects of climate, abundance, and size on the distribution of marine fish: An example
 467 based on four stocks from the Northeast US shelf. *ICES Journal of Marine Science*
 468 72:1311–1322.

469 Blanchard, J. L., C. Mills, S. Jennings, C. J. Fox, B. D. Rackham, P. D. Eastwood, and C. M.
 470 O’Brien. 2005. Distribution-abundance relationships for North Sea Atlantic cod (*Gadus*
 471 *morhua*): observation versus theory. *Canadian Journal of Fisheries and Aquatic Sciences*
 472 62:2001–2009.

473 Brodie, W. B., S. J. Walsh, and D. B. Atkinson. 1998. The effect of stock abundance on range
 474 contraction of yellowtail flounder (*Pleuronectes ferruginea*) on the Grand Bank of
 475 Newfoundland in the Northwest Atlantic from 1975 to 1995. *Journal of Sea Research*
 476 39:139–152.

477 Carroll, G., K. K. Holsman, S. Brodie, J. T. Thorson, E. L. Hazen, S. J. Bograd, M. A. Haltuch, S.
 478 Kotwicki, J. Samhouri, P. Spencer, E. Willis-Norton, and R. L. Selden. 2019. A review of
 479 methods for quantifying spatial predator–prey overlap. *Global Ecology and Biogeography*
 480 28:1561–1577.

481 Cheung, W. W. L., T. L. Frölicher, V. W. Y. Lam, M. A. Oyinlola, G. Reygondeau, U. Rashid
 482 Sumaila, T. C. Tai, L. C. L. Teh, and C. C. C. Wabnitz. 2021. Marine high temperature
 483 extremes amplify the impacts of climate change on fish and fisheries. *Science Advances* 7.

484 Ciannelli, L., J. A. D. Fisher, M. Skern-Mauritzen, M. E. Hunsicker, M. Hidalgo, K. T. Frank, and
 485 K. M. Bailey. 2013. Theory, consequences and evidence of eroding population spatial
 486 structure in harvested marine fishes: A review. *Marine Ecology Progress Series* 480:227–
 487 243.

488 Ciannelli, L., A. B. Neuheimer, L. C. Stige, K. T. Frank, J. M. Durant, M. Hunsicker, L. A.
 489 Rogers, S. Porter, G. Ottersen, and N. A. Yaragina. 2022. Ontogenetic spatial constraints
 490 of sub-arctic marine fish species. *Fish and Fisheries* 23:342–357.

491 Collette, B. B., and J. L. Russo. 1984. Morphology, systematics, and biology of the Spanish
 492 mackerels (*Scomberomorus*, Scombridae). *Fishery Bulletin* 82:545–692.

493 Dahlke, F. T., S. Wohlrab, M. Butzin, and H. O. Pörtner. 2020. Thermal bottlenecks in the life
 494 cycle define climate vulnerability of fish. *Science* 369:65–70.

495 Engelhard, G. H., D. A. Righton, and J. K. Pinnegar. 2014. Climate change and fishing: A century
 496 of shifting distribution in North Sea cod. *Global Change Biology* 20:2473–2483.

497 Fei, S., J. M. Desprez, K. M. Potter, I. Jo, J. A. Knott, and C. M. Oswalt. 2017. Divergence of
 498 species responses to climate change. *Science Advances* 3.

499 Fisher, J. A. D., and K. T. Frank. 2004. Abundance-distribution relationships and conservation of
 500 exploited marine fishes. *Marine Ecology Progress Series* 279:201–213.

501 Frank, K. T., B. Petrie, W. C. Leggett, and D. G. Boyce. 2018. Exploitation drives an ontogenetic-
 502 like deepening in marine fish. *Proceedings of the National Academy of Sciences of the*
 503 *United States of America* 115:6422–6427.

504 Fretwell, S. D., and H. L. Jr. Lucas. 1970. On territorial behavior and other factors influencing
 505 habitat distribution in birds. I. Theoretical development. *Acta Biotheor.* 19:16–36.

506 Froese, R., N. Demirel, G. Coro, K. M. Kleisner, and H. Winker. 2017. Estimating fisheries
 507 reference points from catch and resilience. *Fish and Fisheries* 18:506–526.

508 González-Irusta, J. M., and P. J. Wright. 2016. Spawning grounds of Atlantic cod (*Gadus*
 509 *morhua*) in the North Sea. *ICES Journal of Marine Science: Journal du Conseil* 73:304–
 510 315.

511 Griffith, G. P., P. G. Strutton, and J. M. Semmens. 2018. Climate change alters stability and
 512 species potential interactions in a large marine ecosystem. *Global Change Biology*
 513 24:e90–e100.

514 Hislop, J. R. G., A. P. Robb, M. A. Bell, and D. W. Armstrong. 1991. The diet and food
 515 consumption of whiting (*Merlangius merlangus*) in the north sea. *ICES Journal of Marine*
 516 *Science* 48:139–156.

517 Hobday, A. J., and G. T. Pecl. 2014. Identification of global marine hotspots: Sentinels for change
 518 and vanguards for adaptation action. *Reviews in Fish Biology and Fisheries* 24:415–425.

519 Horwood, J., C. O’Brien, and C. Darby. 2006. North Sea cod recovery? *ICES Journal of Marine*
 520 *Science* 63:961–968.

521 Hsieh, C. hao, A. Yamauchi, T. Nakazawa, and W. F. Wang. 2010a. Fishing effects on age and
 522 spatial structures undermine population stability of fishes. *Aquatic Sciences* 72:165–178.

523 Hsieh, C., A. Yamauchi, T. Nakazawa, and W. F. Wang. 2010b. Fishing effects on age and spatial
 524 structures undermine population stability of fishes. *Aquatic Sciences* 72:165–178.

525 Hughes, N. F., and T. C. Grand. 2000. Physiological ecology meets the ideal-free distribution:
 526 Predicting the distribution of size-structured fish populations across temperature gradients.
 527 *Environmental Biology of Fishes* 59:285–298.

528 Huserbråten, M. B. O., E. Moland, and J. Albretsen. 2018. Cod at drift in the North Sea. *Progress*
 529 *in Oceanography* 167:116–124.

530 ICES. 2016. Report of the Inter-Benchmark Protocol for Whiting in the North Sea (IBP Whiting).

531 ICES. 2018. Report of the Working Group on Assessment of Demersal Stocks in the North Sea
 532 and Skagerrak. Page ICES CM.

533 ICES. 2020. Workshop on Stock Identification of North Sea Cod.

534 Köster, F. W., R. L. Stephenson, and E. A. Trippel. 2014. Gadoid fisheries: The ecology and
 535 management of rebuilding. *ICES Journal of Marine Science* 71:1311–1316.

536 Li, L., A. B. Hollowed, E. D. Cokelet, S. J. Barbeaux, N. A. Bond, A. A. Keller, J. R. King, M.
537 M. McClure, W. A. Palsson, P. J. Stabeno, and Q. Yang. 2019. Subregional differences in
538 groundfish distributional responses to anomalous ocean bottom temperatures in the
539 northeast Pacific. *Global Change Biology* 25:2560–2575.

540 Li, L., A. B. Hollowed, E. D. Cokelet, M. M. McClure, A. A. Keller, S. J. Barbeaux, and W. A.
541 Palsson. 2022. Three-dimensional ontogenetic shifts of groundfish in the Northeast
542 Pacific. *Fish and Fisheries* 23:1221–1239.

543 MacCall, A. D. 1990. Dynamic geography of marine fish populations. Washington Sea Grant
544 Program, Seattle, Wash.

545 Máliš, F., M. Kopecký, P. Petřík, J. Vladovič, J. Merganič, and T. Vida. 2016. Life stage, not
546 climate change, explains observed tree range shifts. *Global Change Biology* 22:1904–
547 1914.

548 Marandel, F., P. Lorance, M. Andreello, G. Charrier, S. Le Cam, S. Lehuta, and V. M. Trenkel.
549 2018. Insights from genetic and demographic connectivity for the management of rays and
550 skates. *Canadian Journal of Fisheries and Aquatic Sciences* 75:1291–1302.

551 Murgier, J., M. Mclean, A. Maire, D. Mouillot, N. Loiseau, F. Munoz, C. Violle, and A. Auber.
552 2021. Rebound in functional distinctiveness following warming and reduced fishing in the
553 North Sea.

554 Orio, A., A. B. Florin, U. Bergström, I. Šics, T. Baranova, and M. Casini. 2017. Modelling
555 indices of abundance and size-based indicators of cod and flounder stocks in the Baltic
556 Sea using newly standardized trawl survey data. *ICES Journal of Marine Science*
557 74:1322–1333.

558 Perry, A. L., P. J. Low, J. R. Ellis, and J. D. Reynolds. 2005. Climate change and distribution
559 shifts in marine fishes. *Science* 308:1912–1915.

Schückel, S., S. Ehrich, I. Kröncke, and H. Reiss. 2010. Linking prey composition of haddock *Melanogrammus aeglefinus* to benthic prey availability in three different areas of the northern North Sea. *Journal of Fish Biology* 77:98–118.

Sunday, J. M., G. T. Pecl, S. Frusher, A. J. Hobday, N. Hill, N. J. Holbrook, G. J. Edgar, R. Stuart-Smith, N. Barrett, T. Wernberg, R. A. Watson, D. A. Smale, E. A. Fulton, D. Slawinski, M. Feng, B. T. Radford, P. A. Thompson, and A. E. Bates. 2015. Species traits and climate velocity explain geographic range shifts in an ocean-warming hotspot. *Ecology Letters* 18:944–953.

Tao, H. H., G. Dur, P. J. Ke, S. Souissi, and C. hao Hsieh. 2021. Age-specific habitat preference, carrying capacity, and landscape structure determine the response of population spatial variability to fishing-driven age truncation. *Ecology and Evolution* 11:6358–6370.

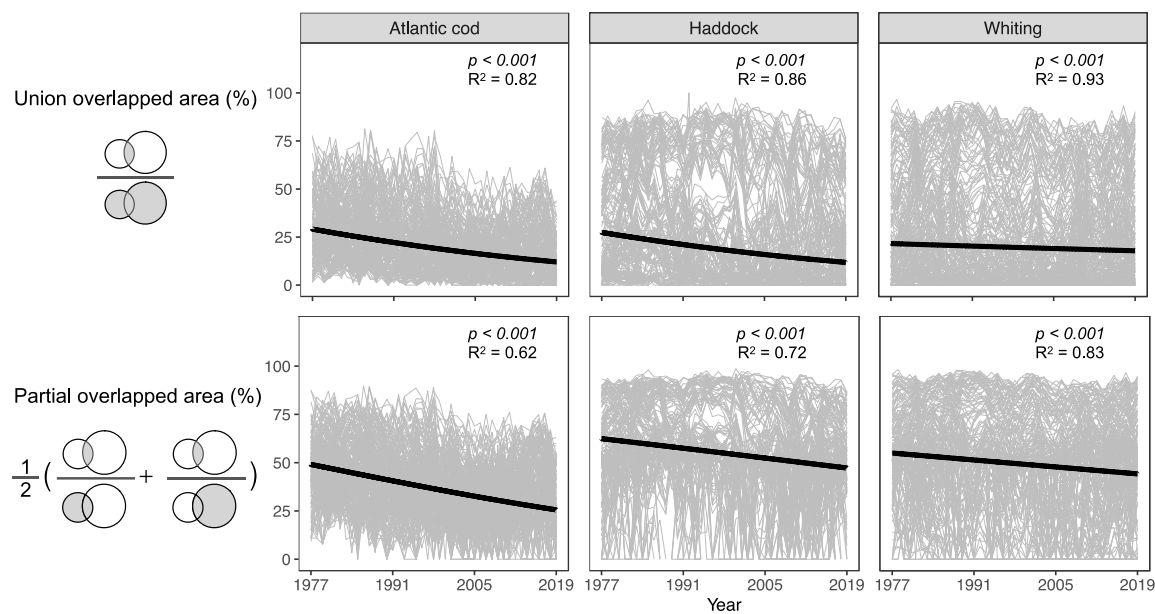
Thorson, J. T., A. Rindorf, J. Gao, D. H. Hanselman, and H. Winker. 2016. Density-dependent changes in effective area occupied for sea-bottom-associated marine fishes. *Proceedings of the Royal Society B: Biological Sciences* 283:20161853.

Tobin, D., P. J. Wright, F. M. Gibb, and I. M. Gibb. 2010. The importance of life stage to population connectivity in whiting (*Merlangius merlangus*) from the northern European shelf. *Marine Biology* 157:1063–1073.

Worm, B., and D. P. Tittensor. 2011. Range contraction in large pelagic predators. *Proceedings of the National Academy of Sciences of the United States of America* 108:11942–11947.

Yang, Q., E. D. Cokelet, P. J. Staben, L. Li, A. B. Hollowed, W. A. Palsson, N. A. Bond, and S. J. Barbeaux. 2019. How “The Blob” affected groundfish distributions in the Gulf of Alaska. *Fisheries Oceanography* 28:434–453.

Zhou, S., A. E. Punt, Y. Ye, N. Ellis, C. M. Dichmont, M. Haddon, D. C. Smith, and A. D. M. Smith. 2017. Estimating stock depletion level from patterns of catch history. *Fish and Fisheries* 18:742–751.



587

588 **Fig. 1 Decline in the overlapped area between pairs of body size groups of three fish**

589 **populations between 1977 and 2019.** Each population is divided into 20 body size groups,

590 resulting in a total of 190 pairs of body size groups. Overlapped area was calculated as the

591 proportion of co-occupied area over the area where either body size group occupies (union

592 overlapped area, upper panel), and as the proportion of co-occupied area over the averaged area of

593 distribution of each body size group of the pair (partial overlapped area, lower panel). Yearly

594 overlapped area for each of the 190 body size group pairs is shown in grey thin line. The temporal

595 trend of overlapped area was examined using linear mixed-effects model, with logit-transformed

596 overlapped area as the response variable, survey year as the fixed effect, and pairs of body size

597 groups within year as the random effect. Black thick lines indicate the regression lines with slope

598 coefficients that were significantly different from zero according to F-test. Conditional R^2 (which

599 considers variances of both fixed and random effects) were reported on the graphs. Marginal R^2

600 (which considers only variances of fixed effects) for Atlantic cod, haddock, and whiting were 0.11,

601 0.03, 0.002 (union overlapped area) and 0.11, 0.04, and 0.006 (partial overlapped area), respectively.

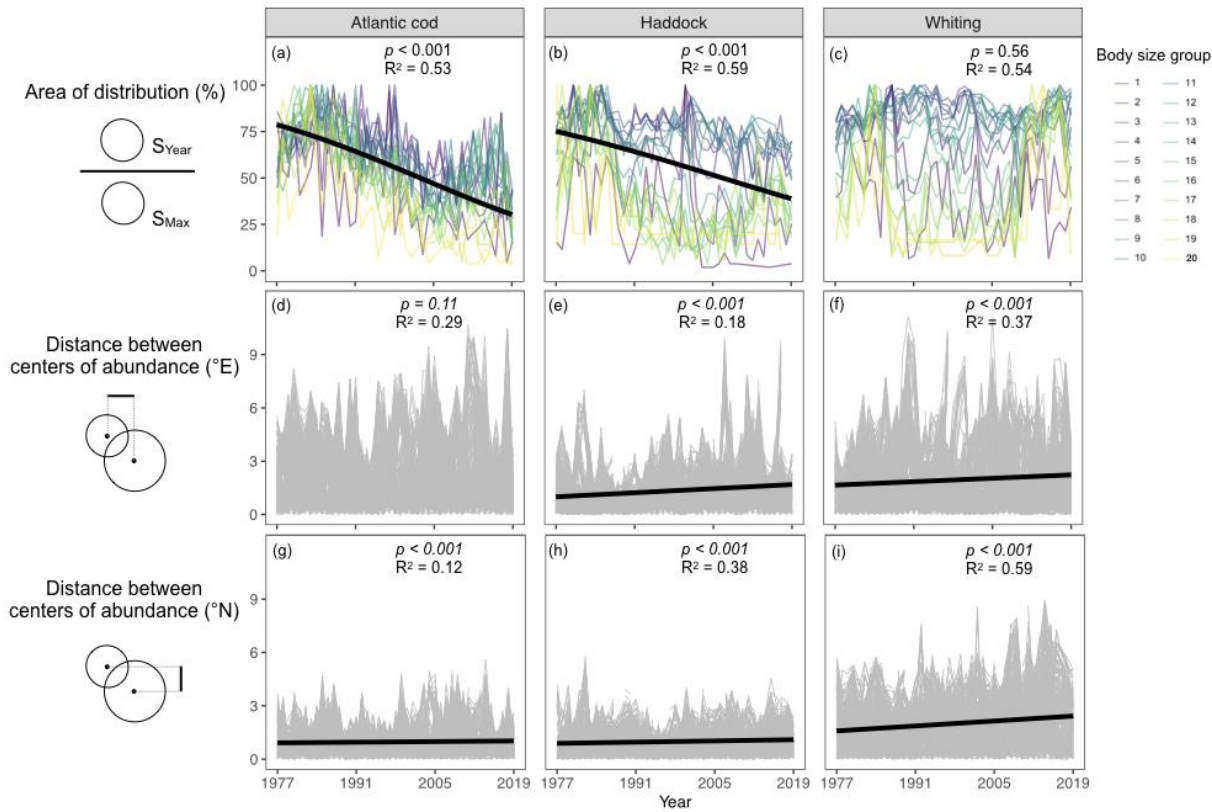


Fig. 2 Area of distribution (a-c) and distance between centers of abundance in longitude (d-f) and latitude (g-i) between 1977 and 2019. (a-c) Area of distribution of each body size group for each year is calculated as the proportion of occupied area (S_{Year}) over the maximum occupied area over the study period (S_{Max}). Colored lines indicate time series of 20 body size groups. We performed linear mixed-effects models, with logit-transformed area of distribution as the response variable, year as the explanatory variable, and body size group nested within year as the random effect. (d-i) Each grey line indicates the time series of one of 190 pairs of body size groups. We performed linear mixed-effects models, with the distance between centers of distribution as the fixed effect, survey year as the fixed effect, and pairs of body size groups nested within year as the random effect. Black thick lines indicate significant models ($p < 0.05$), while nonsignificant results are now shown.

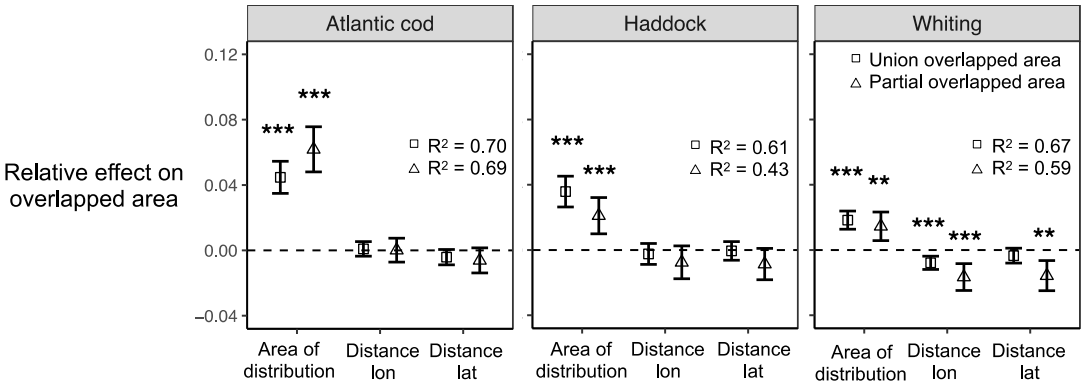
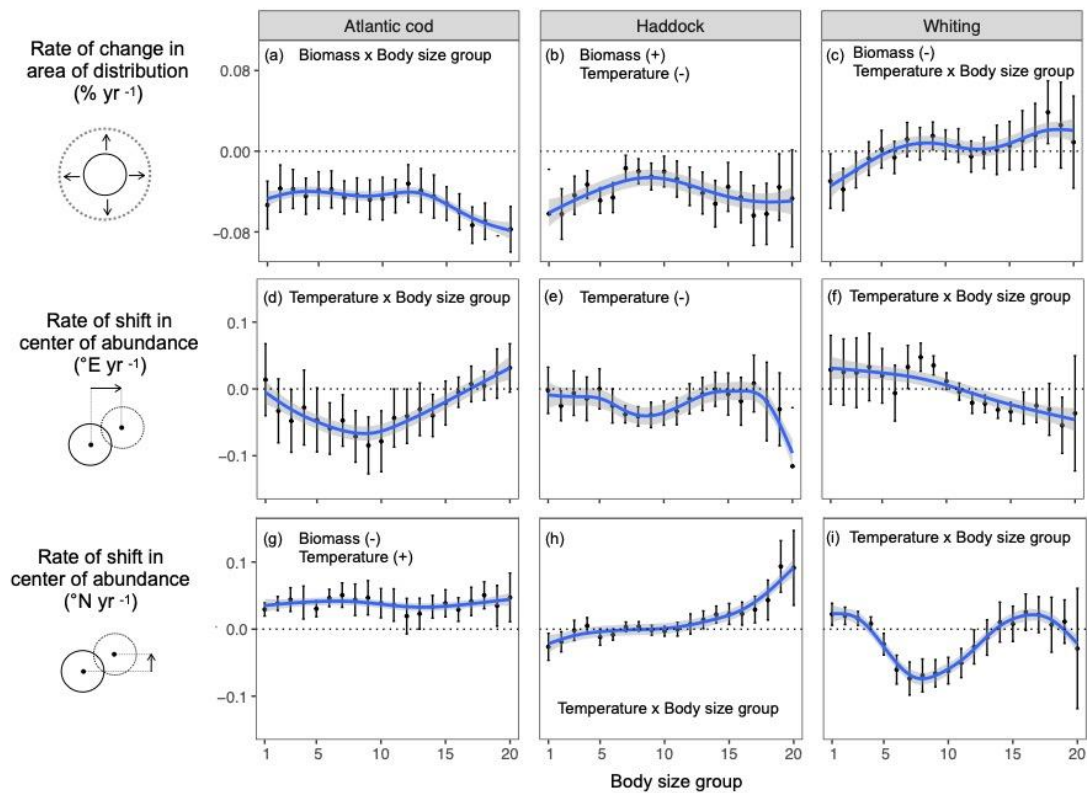


Fig. 3 Effects of area of distribution and distance between centers of abundance on overlapped area. Relative effects were derived from slope coefficients of multiple linear regression models, including overlapped area as the response variable, area of distribution of each body size group and distance between centers of abundance of pairs of body size groups (in longitude and latitude) as explanatory variables, and an AR1 term. Response and explanatory variables were yearly mean values across all body size groups (for area of distribution) or across all pairs of body size groups (for overlapped area and distance between centers of abundance). Bars indicate 95% confidence intervals. Dotted horizontal line indicates zero slope coefficient. *** $p < 0.001$, ** $p < 0.01$, and * $p < 0.05$ indicate that the slope coefficient is significantly different from zero according to F-test.



627 **Fig. 4 Rate of change in the area of distribution and center of abundance between 1977 and**
628 **2019 in response to ocean warming and population biomass decline.** The rate of change for each
629 body size group was indicated by the temporal slope (point) of linear regression model, including
630 area of distribution or center of abundance (in longitude or latitude) as a response variable, and
631 survey year as the explanatory variable. Black bars indicate the 95% confidence intervals of
632 temporal slopes. Differences in the rate of change between body size groups were visualized using
633 loess regression prediction (blue line) and 95% confidence intervals (grey shade), including the
634 temporal slope of body size groups as the response variable and body size group as the continuous
635 explanatory variable. Size-dependent effects of sea bottom temperature (Temperature) and total
636 stock biomass (Biomass) on the area of distribution and the center of abundance were examined in
637 linear regression, by including interaction terms of Temperature or Biomass with body size group

638 in the initial models (further details in Methods). Significant explanatory variables ($p < 0.05$) from
639 the most parsimonious models are shown, with positive effects (+), negative effects (-), or size-
640 dependent effect (\times body size).

