

# Projected shifts in loggerhead sea turtle habitat in the Northwest Atlantic Ocean due to climate change

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## Research Article

**Keywords:** Sea Surface Temperature, Depth, Foraging, Satellite Telemetry, Middle Atlantic 27 Bight, Southern New England, Georges Bank, Gulf of Maine

**Posted Date:** January 8th, 2021

**DOI:** <https://doi.org/10.21203/rs.3.rs-135577/v1>

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**Version of Record:** A version of this preprint was published at Scientific Reports on April 23rd, 2021. See the published version at <https://doi.org/10.1038/s41598-021-88290-9>.

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2 **climate change**

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28 Bight, Southern New England, Georges Bank, Gulf of Maine

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31

## 32 **Abstract**

33           It is well established that sea turtles are vulnerable to atmospheric and oceanographic  
34 shifts associated with climate change. However, few studies have formally projected how their  
35 seasonal marine habitat may shift in response to warming ocean temperatures. Here we used a  
36 high-resolution global climate model and a large satellite tagging dataset to project changes in  
37 the future distribution of suitable thermal habitat for loggerheads along the northeastern  
38 continental shelf of the United States. Between 2009 and 2018, we deployed 196 satellite tags on  
39 loggerheads within the Middle Atlantic Bight (MAB) of the Northwest Atlantic continental shelf  
40 region, a seasonal foraging area. Tag location data combined with depth and remotely sensed sea  
41 surface temperature (SST) were used to characterize the species' current thermal range in the  
42 MAB. The best-fitting model indicated that the habitat envelope for tagged loggerheads  
43 consisted of SST ranging from 11.0° - 29.7° C and depths between 0 – 105.0 m. The calculated  
44 core habitat consisted of temperatures between 15.0° – 28.0° C and at depths between 8.0 – 92.0  
45 m and the highest probability of presence occurred in regions with SST between 17.7° – 25.3° C  
46 and at depths between 26.1 – 74.2 m. The habitat suitability model was then forced by a high-  
47 resolution global climate model under a doubling of atmospheric CO<sub>2</sub> to project loggerhead  
48 probability of presence over the next 80 years. Our results suggest that loggerhead thermal  
49 habitat and seasonal duration will likely increase in northern regions of the NW Atlantic shelf.  
50 This change in spatiotemporal range for sea turtles in a region of high anthropogenic use may  
51 prompt adjustments to the localized protected species conservation measures.

## 52 **Introduction**

53           Warming ocean temperatures due to climate change are already having a measurable  
54 impact on ecological processes (IPCC 2014). An emerging body of research has documented

55 distribution shifts (Pinsky et al. 2020), phenological changes to seasonal migration and  
56 reproduction (Poloczanska et al. 2013), and trophic mismatch (Edwards and Richardson 2004) in  
57 a wide variety of marine taxa. All of these changes increase the difficulty to manage  
58 commercially valuable marine species (Weatherdon et al. 2016) and protect endangered and  
59 threatened animals (Mawdsley et al. 2009).

60         Understanding species distribution and habitat preferences is becoming a fundamental  
61 component to developing effective resource management and conservation strategies (Cañadas  
62 and Hammond 2008, Franco et al. 2004). Fisheries bycatch is one of the most serious threats to  
63 sea turtles around the world (Spotila et al. 2000, Wallace et al. 2013). Attempts to mitigate  
64 bycatch levels are often based on an understanding of when and where a species occurs over time  
65 and how interactions occur with the fishing gear (Dunn et al. 2011, Senko et al. 2014). With the  
66 advent of time/area closures in fisheries management, more research is being conducted to  
67 understand the spatio-temporal nature of by-catch species (Dunn et al. 2011). In the Pacific,  
68 fisheries interactions with loggerhead sea turtles (*Caretta caretta*) have resulted in temporary  
69 area closures, and vessels must comply with stringent regulations to prevent the incidental  
70 capture of this species (Howell et al. 2008). While these types of regulations have resulted in  
71 reduced bycatch of both loggerhead and leatherback sea turtles (*Dermochelys coriacea*)  
72 (Swimmer et al. 2017), they will need to be continuously modified to account for climate change.

73         Sea turtles, including the loggerhead, are susceptible to climate and ecosystem changes,  
74 particularly those associated with temperature. This is most commonly documented with regards  
75 to sea turtle reproductive biology such that nesting phenology, hatchling sex ratios, and various  
76 metrics of nesting success can all be affected by slight changes (< 3° C) in ocean and air  
77 temperature (e.g. Saba et al, 2012, Santidrián Tomillo et al. 2015, Patel et al. 2016a). In terms of

78 marine distribution, habitable temperature ranges are broad, with loggerhead sea turtles observed  
79 throughout the NW Atlantic shelf region in waters with sea surface temperature (SST) ranging  
80 from 7° - 30° C (Shoop and Kenney 1992). In a smaller study on loggerheads at the southern  
81 edge of the NW Atlantic shelf region, Coles and Musick (2000) found that the available thermal  
82 range (4.9° – 32.2° C) was larger than the occupied range (13.3° – 28.0° C), indicating that  
83 loggerheads at-sea likely stay within a preferred temperature envelope. Many marine species  
84 within the region are expected to shift their distribution to remain in preferred thermal habitat  
85 (Kleisner et al. 2017). We hypothesize that loggerheads will do so similarly as the climate  
86 warms.

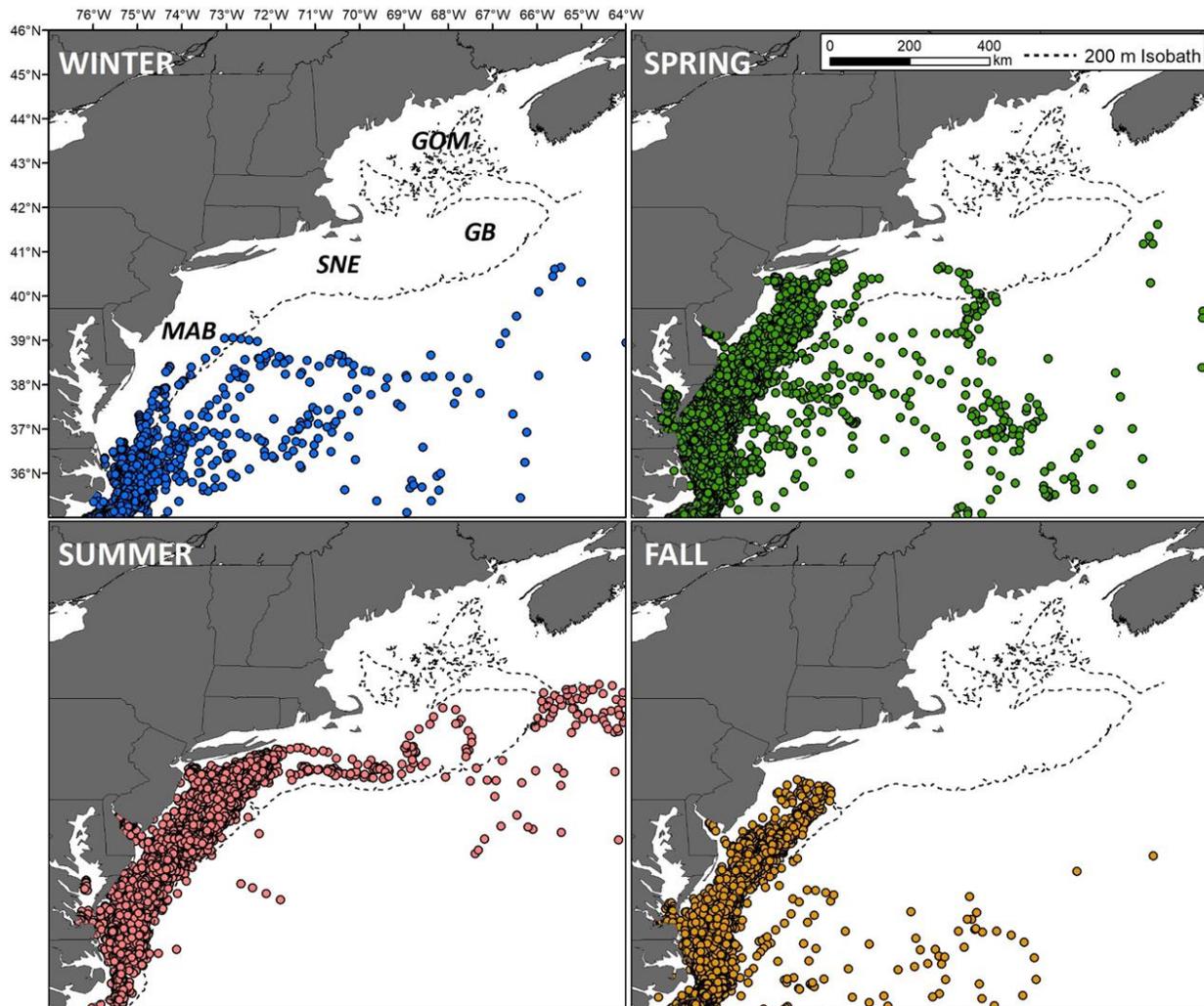
87         In the marine realm, animal distribution modelling has been limited by the availability of  
88 species occurrence data and relevant environmental data (Tyberghein et al 2012). Satellite  
89 telemetry has been occurring for over 35 years (Stoneburner 1982). However, due to the cost  
90 prohibitive nature of these technologies, it is rare for a single population to be studied over many  
91 consecutive years (Hart and Hyrenbach 2009, Hebblewhite and Haydon 2010, Hays and Hawkes  
92 2018). As a result, developing correlations between sea turtle marine habitat and oceanographic  
93 variables have been limited to short term telemetry studies or dependent on opportunistic data  
94 sources, like fisheries bycatch (e.g. Hawkes et al. 2007, Howell et al. 2008, Swimmer et al.  
95 2017). The increasing abundance and availability of information collected by remote sensing  
96 tools such as satellite relayed data loggers and long-term high-resolution environmental data  
97 means that animal distribution models can now more easily be compared with oceanographic  
98 variables (Hazen et al. 2013, Roe et al. 2014, Winton et al 2018).

99         Projections from global climate models are regularly used to estimate long-term shifts in  
100 the distribution of marine species (Poloczanska et a. 2013). However, only a few studies have

101 attempted to project, over a long-term, the climate change induced shift in available marine  
102 habitat for sea turtles. Using a thermal range previously established by Hawkes et al. (2007) for  
103 loggerheads in the NW Atlantic, Witt et al. (2010) projected the change in available thermal  
104 habitat in the Atlantic through 2089 using the global climate model HadGEM1 (Hadley Centre  
105 Global Environmental Model, version 1). Witt et al. (2010) calculated annualized northern and  
106 southern extents at which 90% of SST in the Atlantic Ocean will remain above 15° C as a  
107 threshold for loggerhead distribution. In the Pacific, Hazen et al. (2013) used a generalized  
108 additive model to estimate the relationships between sea turtle distribution and several  
109 oceanographic variables. SST and chlorophyll-a values from Earth system model GFDL ESM  
110 2.1 (Geophysical Fluid Dynamics Lab Earth System Model 2.1) were used to project the  
111 potential change in available ocean habitat through 2100 (Hazen et al. 2013). This bioclimatic  
112 envelope model provided a more direct correlation between the species' distribution and the  
113 projected available habitat (Araujo and Townsend Peterson 2012).

114         The Middle Atlantic Bight (MAB), Southern New England (SNE), Georges Bank (GB),  
115 and the Gulf of Maine (GOM) are adjacent continental shelf regions of the NW Atlantic Ocean  
116 (**Figure 1**) that support a number of large commercial fisheries, a high amount of commercial  
117 and recreational vessel traffic, and the majority of the United States (US) federal wind energy  
118 lease areas (Gilman et al. 2016). Based on aerial surveys, the MAB is also a seasonal foraging  
119 ground for ~40,000 – ~60,000 juvenile and adult loggerheads. The South Atlantic Bight (SAB)  
120 region of the US, between North Carolina and central Florida, is home to ~500,000 – ~1,000,000  
121 loggerheads during the summer months (NEFSC and SEFSC 2011). The population values for  
122 the MAB may be an underestimate as stable isotope analysis and satellite telemetry data indicate

123 that potentially 30 – 50% of loggerheads that nest and reside along the US eastern seaboard  
 124 seasonally forage within the MAB (Griffin et al. 2013, Ceriani et al. 2017, Winton et al 2018).



**Figure 1:** Reconstructed tracks from 196 loggerhead sea turtles satellite tagged between 2009 – 2018 within the northwest Atlantic. Dashed lines denote the 200 m bathymetric contour. GOM = Gulf of Maine, GB = Georges Bank, SNE = Southern New England, and MAB = Middle Atlantic Bight.

125 We calculated the probability of turtle presence in the NW Atlantic shelf region based on  
 126 loggerhead distribution data collected from 2009 – 2018 and the observed relationships with SST  
 127 and depth. We projected the changes to loggerhead bathythermal habitat over the next 80-years  
 128 of climate change using projections of SST for the region from a high-resolution global climate  
 129 model. Overall, we suggest that the loggerhead marine bathythermal habitat will likely expand to

130 the northern regions and increase in seasonal duration to earlier in the spring and later into the  
131 fall.

## 132 **Methods**

### 133 *Loggerhead Satellite Tagging*

134 All loggerheads were tagged between 2009 – 2018. The majority of turtles (n = 190) were  
135 tagged within the MAB between May and September. Two turtles were tagged in eastern GB and  
136 one was tagged off-shelf near southern MAB at the edge of the Gulf Stream, a fast moving  
137 current that flows along the US eastern seaboard pulling warm water from the Gulf of Mexico  
138 northward and eastward (Fofonoff 1981). Three additional turtles were tagged off of the North  
139 Carolina coast in February. Curved carapace length, from anterior notch to posterior tip, was  
140 measured for each captured turtle. Most (n = 186) turtles were equipped with a satellite relay  
141 data logger manufactured by the Sea Mammal Research Unit at the University of St Andrews;  
142 and ten turtles were equipped with a Wildlife Computers SPOT tag. See Patel et al. (2016b,  
143 2018), Winton et al. (2018) and Crowe et al. (2020) for full details on capture and handling  
144 protocols, satellite tag parameterizations, and more details from the satellite tag data.

145 All fieldwork was approved by the US National Marine Fisheries Service (NMFS)  
146 Atlantic Institutional Animal Care and Use Committee and the US Endangered Species Act  
147 (ESA) Section 10(a)(1)(a). Work was conducted under ESA permits #14249 and #18526 issued  
148 to Coonamessett Farm Foundation, Inc., ESA permits #1576 and #16556 issued to the Northeast  
149 Fisheries Science Center and ESA permit #1551 issued to the Southeast Fisheries Science  
150 Center. All methods were carried out in accordance with approved guidelines and regulations.

### 151 *Data Processing*

152 Telemetered location data were processed following standard guidelines for sea turtle  
153 tracking data. Tracks of individual turtles were filtered using a Continuous Time Correlated  
154 Random Walk movement model fitted to location data using the software ‘Template Model  
155 Builder’ (Kristensen et al. 2016) in R (R Core Team 2017). Daily position estimates were  
156 interpolated from each movement model’s output (Johnson et al. 2008; Albertsen et al. 2015;  
157 Winton et al. 2018) to correct for the different transmission rates of each tag and to reduce  
158 autocorrelation in position estimates. Prior to fitting the movement model, all location  
159 coordinates were re-projected into the oblique Mercator center projection centered on 35.0°N,  
160 75.0°W using R package ‘rgdal’ (Bivand et al. 2015) and a speed filter with a max speed of 5  
161 km/hr was applied to remove errant telemetered locations (TEWG 2009).

162 For analysis, loggerhead locations were aggregated over the NMFS Atlantic Marine  
163 Assessment Program for Protected Species (AMAPPS) spatial grid that has a 10-km resolution  
164 (Winton et al. 2018). For model fitting, we used loggerhead location estimates from continental  
165 shelf waters (depths < 200 m) between 33.5°N - 41.6°N, which encompasses the MAB as  
166 delineated by NMFS statistical areas (Clay 1996) and corresponds to the highest density area  
167 traversed by satellite-tagged loggerheads (Winton et al. 2018). Although on occasion  
168 loggerheads were tracked in waters deeper than 200 m and north of 41.6°N latitude, those  
169 locations were removed from this study due to the low sample size and the higher incidences of  
170 fisheries interactions occurring on shelf waters (Murray and Orphanides 2013). Locations were  
171 binned by month to match the resolution of climate projections (Saba et al. 2016) and aggregated  
172 over the 10-km resolution AMAPPS spatial grid. The AMAPPS grid was bounded by the  
173 coastline to constrain loggerhead space use to the ocean. All maps were prepared with ESRI  
174 ArcMAP 10.8.

175 *Characterizing the Bathothermal Range of Loggerheads*

176           Although we understand loggerheads are likely influenced by a large range of  
 177 environmental variables, our goal was to investigate how the distribution of loggerheads may  
 178 change in response to warming water temperatures associated with climate change. To model  
 179 spatial variation in the occurrence of tagged loggerheads and identify the proportion of the  
 180 observed variation related to water temperature, generalized linear models (GLMs) were applied  
 181 to estimate the relationship between the probability of loggerhead presence, SST, and depth. We  
 182 modeled the occurrence,  $y_{it}$ , (0 = absent, 1 = present), of tagged turtles in grid cell  $i$  during time  
 183 step  $t$  as the outcome of a Bernoulli random variable:

$$184 \qquad y_{it} \sim \text{Bernoulli}(p_{it}),$$

185 where  $p_{it}$  is the probability that a tagged turtle was present. We modeled  $p_{it}$  as a function of SST  
 186 and depth as:

$$187 \qquad \text{logit}(p_{it}) = \beta_0 + \beta_1 \text{SST}_{it} + \beta_2 \text{SST}_{it}^2 + \beta_3 \text{Depth}_{it} + \beta_4 \text{Depth}_{it}^2,$$

188 where the logit link function constrains  $p_{it}$  from 0 to 1,  $\beta_0$  is an intercept term;  $\beta_1$  and  $\beta_2$  represent  
 189 a quadratic effect of SST (which allows for a non-linear relationship);  $\beta_3$  and  $\beta_4$  a quadratic effect  
 190 of bottom depth. Depth data for daily loggerhead locations were obtained from the gridded  
 191 ETOPO1 bathymetry data set (Amante and Eakins 2009). For observed ocean temperature data  
 192 (2009 – 2018), we used the Optimum Interpolation Sea Surface Temperature (OISST) product  
 193 for the same time period of the turtle tracking. OISST is a combination of observations from  
 194 different platforms (satellites, ships, buoys) and is produced at a  $1/4^\circ$  spatial resolution  
 195 (Reynolds and Smith 1994). Daily OISST satellite composites were obtained from the NOAA  
 196 CoastWatch Program (<http://coastwatch.pfeg.noaa.gov/erddap/griddap/>) using functions

197 available in the R package ‘rerrdap’ (Chamberlain 2016) and averaged together to create  
198 monthly climatologies to match the output of the climate model projections. These composites  
199 were then up-sampled to align with the AMAPPS grid by simple averaging.

200 All model variants were fit via maximum likelihood methods using the package  
201 ‘Template Model Builder’ (Kristensen et al. 2016). All parameters were treated as fixed effects;  
202 the final gradient value for parameters and the hessian matrix were inspected for each model fit  
203 to confirm convergence. We used the Akaike information criterion (AIC; Akaike 1973) and the  
204 percent deviance explained (Maunder and Punt 2004) for model selection. Given the small  
205 number of explanatory variables considered, we used a forward, step-wise selection approach to  
206 identify the most parsimonious combination of regression terms (Zuur et al. 2009). Individual  
207 terms were retained in the model if their inclusion resulted in lower AIC values and increased the  
208 proportion of the deviance explained relative to the best-fitting model from the previous step. To  
209 assess the fit of the selected model and examine potential model misspecification, we examined  
210 standard residual diagnostic plots using normalized, randomized residuals (Benjamin et al.  
211 2003). Visualizations of model results were produced using functions available in the R package  
212 ‘tidyverse’ (Wickham et al. 2019).

### 213 *Forecasting the distribution of loggerheads under climate change*

214 To investigate how the distribution of loggerheads may shift under climate change, the  
215 selected model was fitted to long-term (80 year) projections of SST in the MAB, SNE, GB, and  
216 GOM, the most northern portion of loggerhead range within the western Atlantic Ocean (Winton  
217 et al. 2018). Projections were based on a climate change scenario from the National Oceanic and  
218 Atmospheric Association’s (NOAA’s) high-resolution global climate model (CM2.6) as  
219 described by Saba et al. (2016). Unlike most global climate models that have a warm bias due to

220 misrepresentation of the position of the Gulf Stream, CM2.6 resolves the Gulf Stream, regional  
221 ocean circulation, and bathymetry of the Northwest Atlantic shelf (Saba et al. 2016) much more  
222 realistically. Many studies that have projected marine species habitat shifts in the Northwest  
223 Atlantic have relied on this climate model (Tanaka et al. 2020, McHenry et al. 2019, Selden et al.  
224 2018, Kleisner et al. 2017)

225         The SST output from CM2.6 represents a monthly deviation from a historical average  
226 derived from control simulations (deltas). The CM2.6 output was rasterized onto a  $0.1^\circ \times 0.1^\circ$   
227 mesh and then synced to the AMAPPS grid. The SST deltas were then added to the mean  
228 monthly SST values for the observed time period. Along with depth, which we assumed  
229 remained constant, the projected monthly SST was used to project the probability of loggerhead  
230 presence from the MAB north to GOM within the continental shelf for 80 years conditioning on  
231 the fitted model. For visualization purposes, observed and projected data were grouped into  
232 seasons based on both general climate trends for the region and turtle habitat usage patterns  
233 (Griffin et al. 2013, Patel et al. 2018, Winton et al. 2018). The projected probability of presence  
234 was then averaged across years (10 and 20 year bins). January through March were grouped into  
235 winter, April through June to spring, July through September to summer; and October through  
236 December to fall.

### 237 *Quantifying climate-related shifts in distribution*

238         To better understand and visualize the predicted changes in loggerhead occupancy  
239 (presence / absence) under climate change, we developed a binary classifier using the Index of  
240 Union (IU) approach to determine whether a cell would be occupied by a loggerhead turtle given  
241 the relationships identified (Unal 2017). This analysis was done using the R package “ROCR”  
242 (Sing et al. 2005). In short, the IU approach attempts to find an optimal cut point ( $c$ ) that

243 correctly classifies the fitted continuous probabilities of loggerhead presence as a 1 (present) or 0  
 244 (absent). The optimal value of  $c$  is that which minimizes the  $IU$  criterion:

$$245 \quad IU(c) = (|Se(c) - AUC| + |Sp(c) - AUC|),$$

246 where  $Se$  is the sensitivity (true positive rate),  $Sp$  is the specificity (true negative rate), and  $AUC$   
 247 is the Area Under the Receiver Operating Curve. Using the optimal cut-point ( $c = 0.08$ ), we  
 248 classified seasonal averages of presence probabilities by projected decade to identify cells that  
 249 could be occupied by loggerhead turtles based on the bathythermal habitat associated with  
 250 observed loggerhead occupancy patterns. We labeled the IU region classified to have loggerhead  
 251 presence as the ‘core habitat’. The fraction of cells in the study area that could be occupied by  
 252 loggerheads was then calculated to explore trends in projected occupancy over time. We also  
 253 calculated the region with the highest probability of presence by taking the top 25% of the  
 254 predicated habitat values (Hazen et al. 2013).

## 255 Results

**Table 1:** Summary information for satellite tag deployments  
 (CCL = curved carapace length from notch to tip).

256 In total, 196  
 257 loggerheads from 2009 – 2018  
 258 were fitted with satellite tags  
 259 (**Table 1**). Turtles were either  
 260 late-stage juveniles or adults  
 261 with a mean ( $\pm$ SD) curved  
 262 carapace length of  $80.0 \pm 9.7$   
 263 cm. Filtering the location  
 264 estimates from these tags

	Tags	Mean	SD
Year	Deployed	CCL	CCL
2009	2	71.8	7.4
2010	14	77.8	9.2
2011	26	79.1	7.8
2012	30	81.9	8.7
2013	16	79.2	13.4
2014	18	78.2	9.8
2015	10	78.7	12.4
2016	21	80.4	8.6
2017	24	78.5	12.0
2018	35	82.7	8.4
Mean	19.6	80.0	9.7

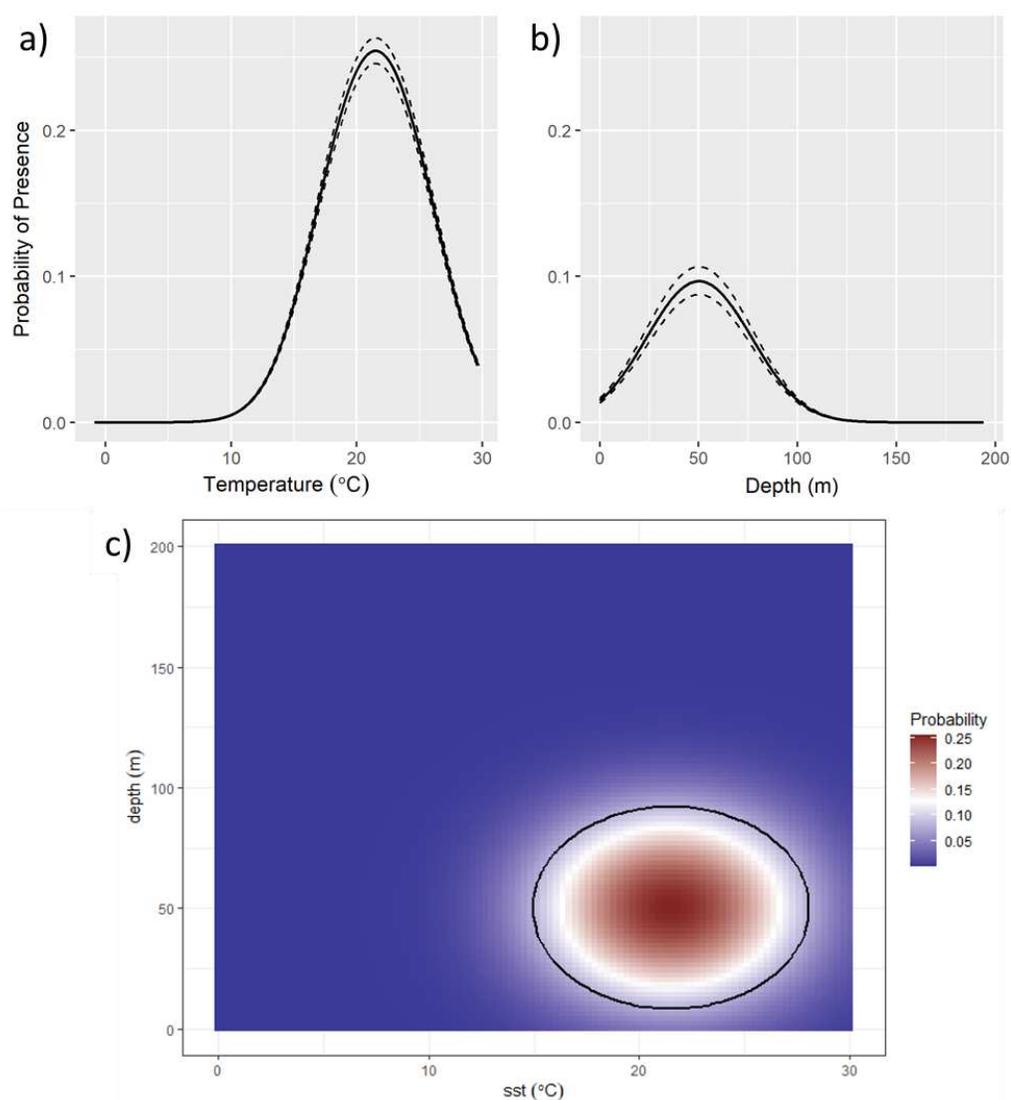
265 yielded 45,840 daily locations within the NW Atlantic (north of 33.5°N and west of 64°W, the  
266 approximate southern and western boundaries of the US northeast continental shelf Large Marine  
267 Ecosystem, Link et al. 2008), of which 44,865 daily locations occurred on the continental shelf  
268 in the MAB and were used for model fitting.

269 Model selection for explanatory variables supported a relationship between loggerhead  
270 presence, SST, and depth. SST alone explained 15.4% of the variability in loggerhead presence,  
271 while including only bottom depth explained 4.1%. SST and depth combined explained 20.1% of  
272 the variability in loggerhead presence. Based on the fitted model, loggerheads occur at SST  
273 between 11.0° - 29.7° C and at depths between 0 – 105.0 m (**Figures 2a and 2b**). The overall  
274 predicted distribution for each month was consistent with the reconstructed tracks and indicated  
275 that the probability of loggerhead presence in the NW Atlantic shelf waters is highest from May  
276 through October. Portions of SNE and GB were estimated to have a higher probability of  
277 presence than the MAB during summer months.

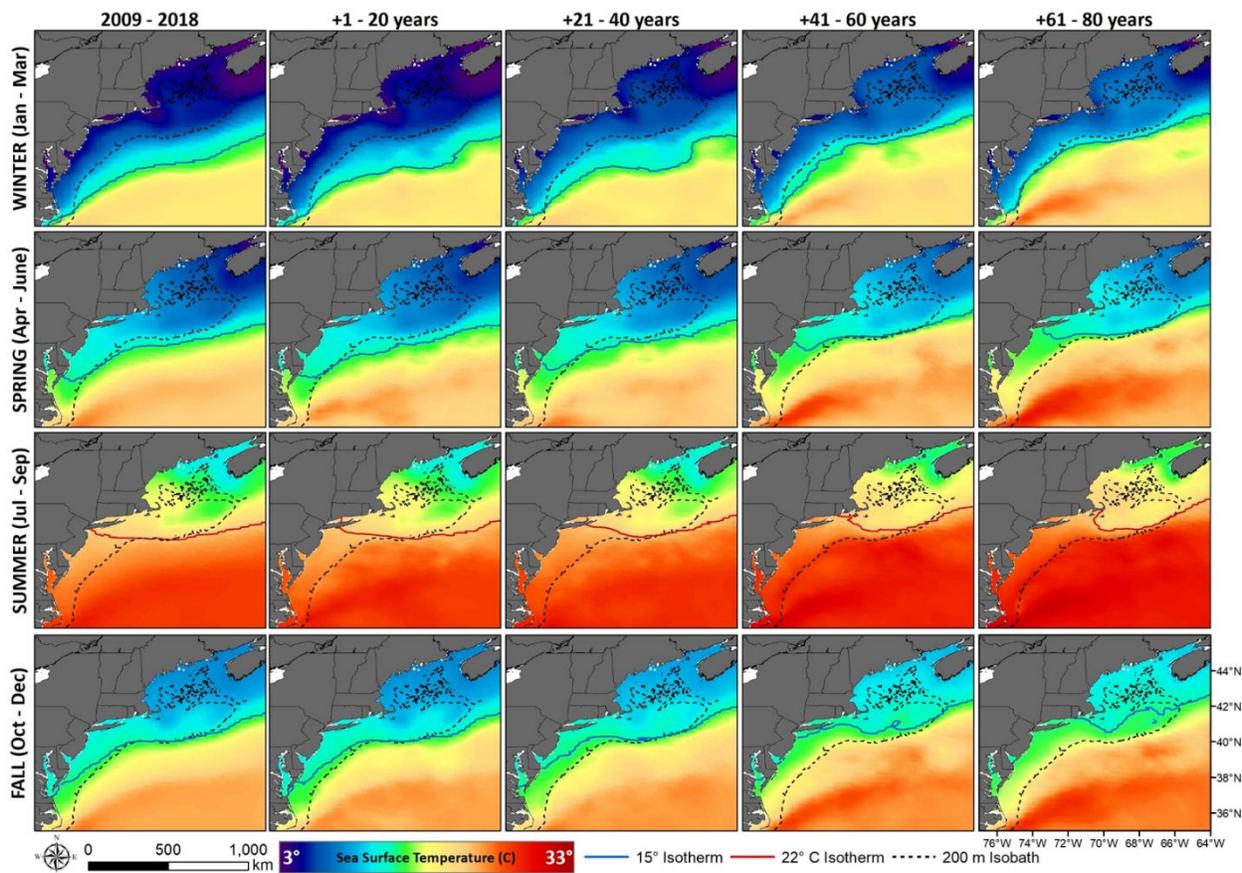
278 Using the binary classifier resulted in a core habitat that consisted of temperatures  
279 between 15.0° – 28.0° C and depths between 8.0 – 92.0 m (**Figure 2c**). The highest probability  
280 of presence occurred in regions with SST between 17.7° – 25.3° C, and depths between 26.1 –  
281 74.2 m. More specifically, turtles tended to occupy regions of the MAB with SST closest to  
282 21.5° C at depths closest to 50 m.

283 The CM2.6 model projected that warmer SST would push farther inshore and north  
284 through all seasons (**Figure 3**). Mean SST within the shelf region is expected to gradually  
285 increase in the first 40 years, then intensify over the following 40 years. The probability of  
286 presence for loggerheads in the NW Atlantic is projected to increase from the observed May –  
287 October season, to an April – December season within 20 – 60 years, encompassing the entire

288 spring, summer and fall seasons (**Figure 4**). In particular, fall is expected to have the largest  
 289 increase in available thermal habitat for loggerheads, followed by spring (**Figure 5**). Minimal  
 290 changes in winter are expected in terms of available suitable habitat over the next 80 years, while  
 291 available habitat in the summer is expected to slightly decline as the most southern portions of  
 292 the MAB warm beyond the range of our established temperature envelop.



**Figure 2:** Probability of presence of loggerheads in relation to (a) sea surface temperature (SST) and (b) bottom depth. Dashed lines indicate 95% confidence intervals. The resulting core habitat as identified using the 'Index of union' is illustrated in (c), where the graph identifying the probability of loggerhead presence from observed data associated with the combined SST and depth ranges and the calculated core habitat (black circle).



**Figure 3:** Seasonal maps of historical and projected sea surface temperature in the northwest Atlantic. The north- and shore-ward movement of the Gulf Stream is expected to increasing warming within shelf waters.

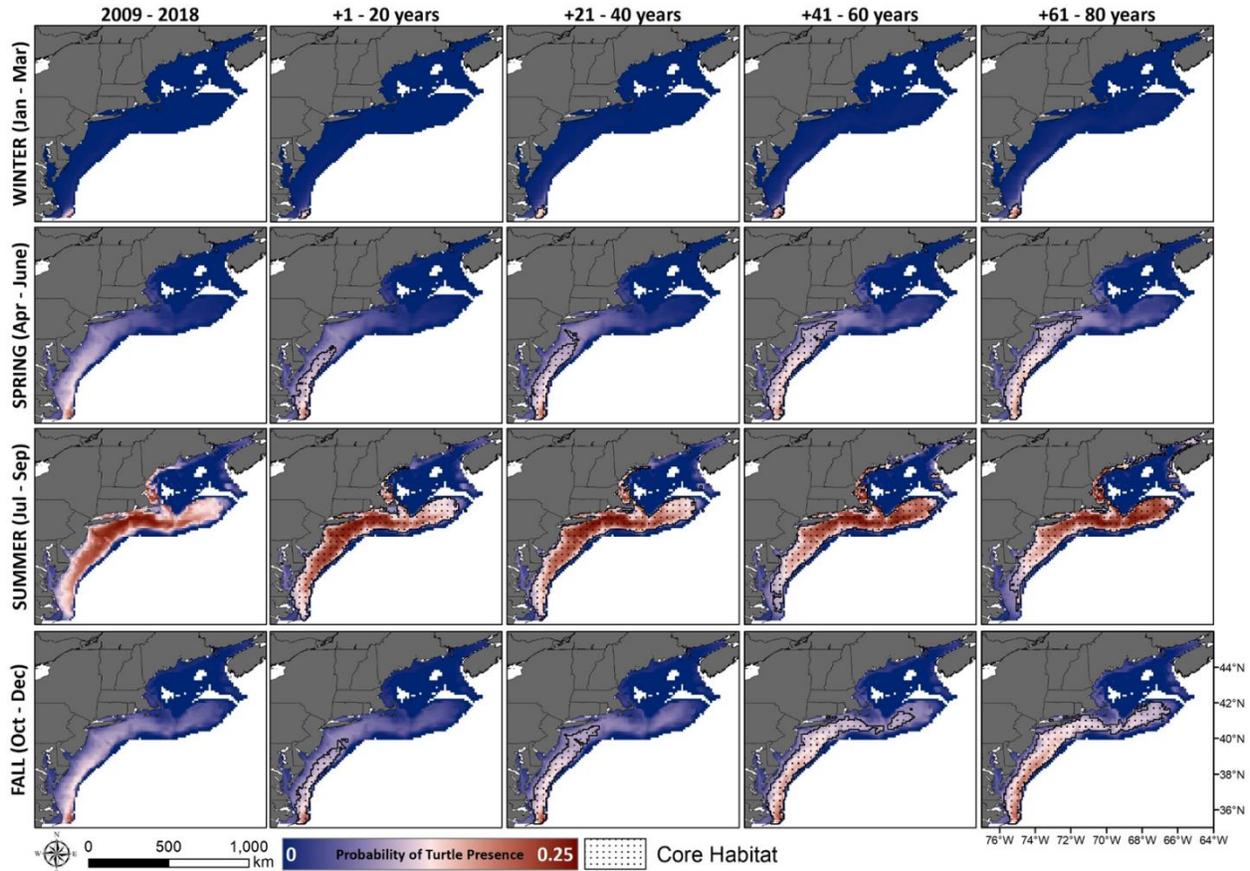
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**Figure 4:** Seasonal maps of probability of turtle presence and core habitat based on observed and projected sea surface temperatures (SST) using the CM2.6 model. Color ramp matches Figure 2c and indicates the probability of presence based on SST and depth.

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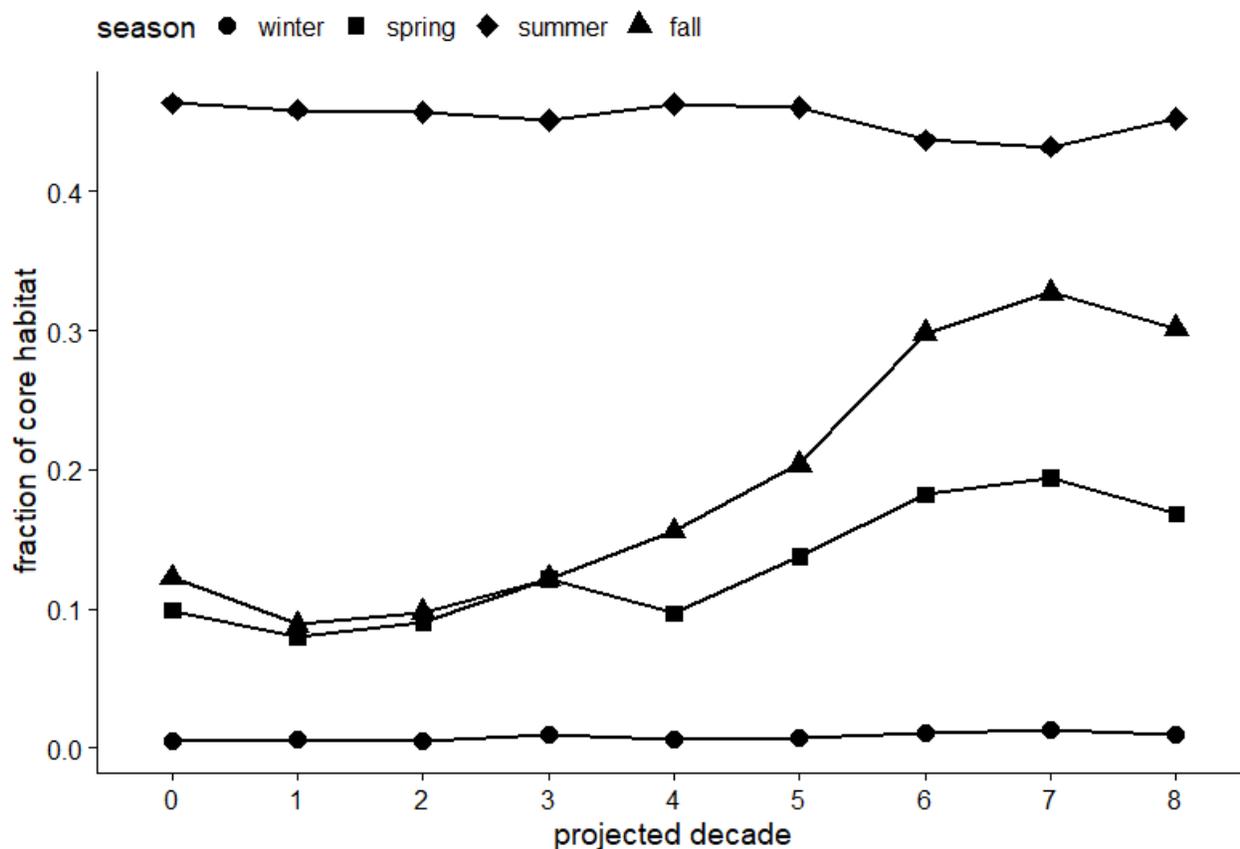
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**Figure 5:** Change in fraction of the NW Atlantic shelf region identified as core habitat for loggerheads across the projected 80 years binned by decade. Spring and fall are projected to have the largest change. Decade ‘0’ refers to the observed data.

306

## 307 Discussion

308 Based on projected increases in SST and the large bathythermal habitat envelope for  
 309 loggerheads, we project an expansion of potentially suitable loggerhead habitat. The CM2.6  
 310 model projects a  $\sim 3^{\circ}\text{C}$  increase in mean SST for all seasons within 60 – 80 years. This increase  
 311 in SST is not uniform – the GOM is expected to warm faster than the more southern regions  
 312 (Saba et al. 2016). Shelf waters in the MAB are typically cooler than offshore waters with these  
 313 two water masses bounded by the Slope Sea, a narrow band of ocean along the continental slope  
 314 between the shelf and the Gulf Stream (Bane et al. 1988). However, CM2.6 projects a change to

315 these conditions with the southward flowing cool Arctic waters of the Labrador Current  
316 retreating along with a north- and shore-ward shift of the Gulf Stream, allowing warmer water to  
317 enter the NW Atlantic shelf region (Saba et al. 2016). These oceanographic features are expected  
318 to provide continued suitable bathythermal habitat for loggerheads in the MAB, SNE and GB,  
319 and may also expand the range of suitable habitat beyond the observed season (Griffin et al.  
320 2013, Winton et al. 2018) and northward into GOM.

321 Our results matched well with previous research attempting to establish the bathythermal  
322 habitat envelope for loggerheads in the region. Hawkes et al. (2007) found that satellite tagged  
323 loggerheads had a similar range for temperature ( $12.8^{\circ} - 30.4^{\circ} \text{C}$ ) and depth (0.25 – 104.4 m),  
324 while Hawkes et al. (2011) updated these results and narrowed the ranges to SST =  $18.2^{\circ} - 29.2^{\circ}$   
325 C and depth = 3 – 89 m. Additionally, Mansfield et al. (2009) identified that tracked loggerheads  
326 that stayed within the neritic zone of the NW Atlantic experienced an SST range between  $9.0^{\circ} -$   
327  $29.3^{\circ} \text{C}$  across all seasons. During the summer the MAB experiences a sharp thermocline with  
328 the formation of a Cold Pool water mass that is  $\sim 15^{\circ} \text{C}$  cooler than surface waters and occurring  
329 near-bottom at depths between 30 – 70 m (Lentz et al. 2017). Loggerheads regularly forage  
330 within the Cold Pool (Patel et al. 2018) and those that remain in the shelf waters of the southern  
331 US regularly inhabit environments with a much higher SST than observed in the MAB (Iverson  
332 et al. 2019). As a result, the habitat envelope estimated from our model may be slightly narrower  
333 than the temperature and depth ranges in which loggerheads are able to thrive. Thus loggerhead  
334 response may not match the rate of the projected northward shift in their available thermal  
335 envelope or their movements may be driven by other correlating factors (Hazen et al. 2013).

336 Despite our use of only two explanatory variables (SST and depth), our model results  
337 showed similar patterns of loggerhead distribution to sightings, strandings, and bycatch data,

338 with slight variations due to the unique practices of each fishery. Braun-McNeill et al. (2008)  
339 found that 11° C was a conservative minimum SST threshold that aligned well with sea turtle  
340 distribution in the NW Atlantic shelf region from ten years of sightings, strandings and bycatch  
341 data. Swimmer et al. (2017) identified that loggerheads were most often caught in long lines  
342 when SST ranged between 18° - 24° C and hook depth was 22 m or shallower; however, these  
343 results included a much larger portion of the greater Atlantic Ocean. Gillnet bycatch between  
344 Massachusetts and North Carolina, having occurred nearly year-round, was within a broader  
345 range of SST and depth (SST = 8.6° – 27.8°C; depth = 1.8 – 76.8 m; Murray 2009). Observed  
346 bycatch in scallop dredges was limited to SST between 18° - 25° C and depths of 36 – 68 m  
347 (Murray 2011). These values from the scallop fishery aligned closest to our ranges for highest  
348 probability of turtle presence (SST = 17.7° – 25.3° C and depths = 26.1 – 74.2 m) because by-  
349 caught turtles in the scallop fishery had a high spatiotemporal overlap with when and where we  
350 caught the majority of our tracked loggerheads. Simultaneous integration of multiple data  
351 streams during statistical model estimation could help with more robust characterization of  
352 habitat envelopes for marine species in addition to this corroborative evidence, particularly for  
353 cases with incomplete and imperfect data resolution and could be a target for future research.

354 We built upon projections calculated by Witt et al. (2010) of suitable loggerhead habitat  
355 by developing probability models with monthly projections. Witt et al. (2010) created annual  
356 projections using a 15° C threshold and suggested that for 90% of the year the MAB and areas  
357 north are unsuitable habitats for loggerheads, even as ocean temperatures warm. However, Witt  
358 et al. (2010) added that during summer months, loggerheads would regularly forage farther north  
359 than their annualized habitat suitability contours. Results of our winter projections matched well  
360 with annualized contours from Witt et al. (2010), indicating that loggerheads would have a very

361 low probability of entering the MAB during this season, remaining farther south, or potentially in  
362 warmer offshore waters. However, throughout the remainder of the year, we projected that the  
363 loggerhead thermal habitat envelope would expand into MAB shelf waters earlier in the spring,  
364 continue moving north beyond the observed range in SNE and GB, and retreat south later in the  
365 fall. This corresponds closely with the trend of the spring and fall 15° C SST threshold, with this  
366 isotherm continuing north in both shelf and offshore waters throughout the next 80 years.

367         Although observed data from aerial surveys indicates that SNE and GB are already  
368 suitable for loggerheads during portions of the year (Shoop and Kenney 1992), turtles are likely  
369 not as abundant farther north due to the shorter thermal window and the existing availability of  
370 prey resources in the MAB. With the projected increased thermal window, loggerheads may  
371 have more time to explore and actively forage within the northern shelf regions, while reducing  
372 competition for prey resources in the more heavily populated MAB, creating higher value to the  
373 longer distance migration (Alerstam et al. 2003).

374         In addition to moving northward, there are likely other loggerhead distribution shifts that  
375 could occur as environmental conditions in the NW Atlantic change. For example, if conditions  
376 become unsuitable for loggerhead foraging in the MAB, turtles may forage off-shelf for extended  
377 periods of time. Our tagging data do indicate that loggerheads venture off-shelf on rare occasions  
378 and adult and sub-adult loggerheads have been tracked foraging in pelagic environments  
379 throughout the world (Hatase et al. 2002, 2007, Hawkes et al. 2006, Reich et al. 2010). In the  
380 MAB, we have documented loggerheads foraging pelagically on jellyfish (Smolowitz et al. 2015,  
381 Patel et al. 2016c), and these off-shelf regions adjacent to GOM, GB, MAB and SNE in the  
382 Northwest Atlantic are known migratory corridors and feeding grounds for leatherback turtles,  
383 obligate jellyfish foragers (James et al. 2005, Dodge et al. 2014).

384           As the thermal habitat in the MAB through GB shelf region changes, this will also likely  
385 cause shifts in prey densities and species compositions. Using the same CM2.6 global climate  
386 model, Kleisner et al. (2017) described the shifts in available thermal habitats for over 30  
387 commercially valuable marine species within the NW Atlantic continental shelf. In general,  
388 Kleisner et al. (2017) projected the expansion of available thermal habitats for southern species,  
389 and the reduction in available thermal habitats for northern species during the spring and fall  
390 seasons. Amongst these species, the vulnerability of the Atlantic sea scallop (*Placopecten*  
391 *magellanicus*) to climate change may provide an indication of how turtles may shift habitat  
392 usage. Atlantic sea scallops are a known prey item for loggerheads (Smolowitz et al. 2015) and  
393 there is generally a high overlap between loggerhead and sea scallop distribution in the MAB  
394 based on preferred depth range (Hart and Chute 2004). Recent research by Rheuban et al. (2018)  
395 has found that each scallop stock (MAB and GB) may react differently to climate change and  
396 that the more northern GB population may be slightly buffered from negative impacts due to the  
397 different mechanisms for larval recruitment between the stocks. However, using changes in  
398 ocean temperature and salinity from CM2.6, Tanaka et al. 2020 projected substantial scallop  
399 habitat declines throughout the MAB and GB but increased habitat in coastal GOM. As a result,  
400 the benthic community could substantially change in the MAB, potentially causing loggerheads  
401 to seek other, perhaps more northerly, shelf habitats for prey resources.

402           Shifting distribution as a result of climate change may change interactions of marine  
403 species with human uses. Changes to loggerhead range and seasonality may create spatial  
404 overlap with fisheries that have not previously needed sea turtle conservation measures. For  
405 example, the Atlantic sea scallop fishery developed gear modifications, (Turtle Deflector Dredge  
406 and Turtle Chains; Smolowitz et al. 2012) to reduce the bycatch of sea turtles and mandated

407 these measures for boats fishing in the MAB, specifically west of  $-71^{\circ}\text{W}$ , from May through  
408 November (Framework Adjustment 23 NOAA-NMFS-2011-0255). Our model projects that  
409 within the next 20 – 40 years, loggerheads could forage within the NW Atlantic shelf region  
410 outside the spatial and temporal range these scallop gear modifications are required. Because the  
411 scallop industry has already developed a dredge effective at reducing turtle bycatch, adjusting the  
412 gear to remain efficient in the more northern scalloping grounds and expanding its usage could  
413 be an effective solution with minimal economic impacts to the fishery (Smolowitz et al. 2012).  
414 However, northern fisheries that use pelagic long lines, trawls and gillnets have the potential to  
415 see increases in turtle bycatch if fisheries management does not adapt to projected environmental  
416 changes. For example, the bottom trawl fishery operating in the MAB, SNE and GB, from 2014  
417 – 2018, had the highest number of estimated sea turtle interactions occur north of  $39^{\circ}\text{N}$ , which is  
418 farther north than in previous years (Murray 2020).

419 Overall, sea turtle seasonal habitat usage and distribution is likely linked to a broader  
420 range of environmental variables beyond SST and depth (Hazen et al. 2013), as well as  
421 biological factors like the availability of prey resources (Houghton et al. 2006), reproductive  
422 cycles and life stage (Nelson 1988). However, given the availability of data and what is known  
423 about loggerhead ecology in general, the type of bioclimatic envelope model we present provides  
424 a reasonable assessment of the potential drivers for the distribution of this cohort of loggerheads  
425 (Araujo and Townsend Peterson 2012). To truly determine how climate change will impact these  
426 turtles will require continuous monitoring, particularly in the MAB, SNE and GB. Our results  
427 can guide expectations for likely future turtle distributions and inform discussions to plan for  
428 reasonable conservation measures. If future conservation measures and turtle distribution need to  
429 be synchronized, additional monitoring and adaptive management will be necessary.

**430 Author Contributions**

431 S.P. wrote the main text and prepared figures 1, 3 and 4. M.W and J.H. performed the  
432 analyses, created content for figures 1 and 4 and prepared figures 2 and 5. H.H. and R.S.  
433 provided research funding to obtain turtle telemetry data. V.S. provided climate model data. G.F.  
434 provided analytical expertise. All authors developed and reviewed the manuscript.

**435 Acknowledgements**

436 We thank James Gutowski of Viking Village Fisheries and the captains, crew and  
437 scientists on the F/V Kathy Ann and F/V Ms Many for their expert field work. Leah Crowe,  
438 Kathryn Goetting, Eric Matzen, Henry Milliken, Shea Miller, Liese Siemann, Brianna Valenti,  
439 Daniel Ward, and Matthew Weeks were integral to the success of this project. This study was  
440 funded in part by the scallop industry Sea Scallop Research Set Aside program administered by  
441 the Northeast Fisheries Science Center, by the U.S. Department of the Interior, Bureau of Ocean  
442 Energy Management through Interagency Agreement M19PG00007 with the U.S. Department of  
443 the Commerce, National Oceanic and Atmospheric Administration (NOAA), National Marine  
444 Fisheries Service (NMFS), Northeast Fisheries Science Center (NEFSC), by the National  
445 Oceanic and Atmospheric Administration Saltonstall-Kennedy Grant Program, and by the  
446 National Marine Fisheries Protected Species Toolbox Initiative.

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# Figures

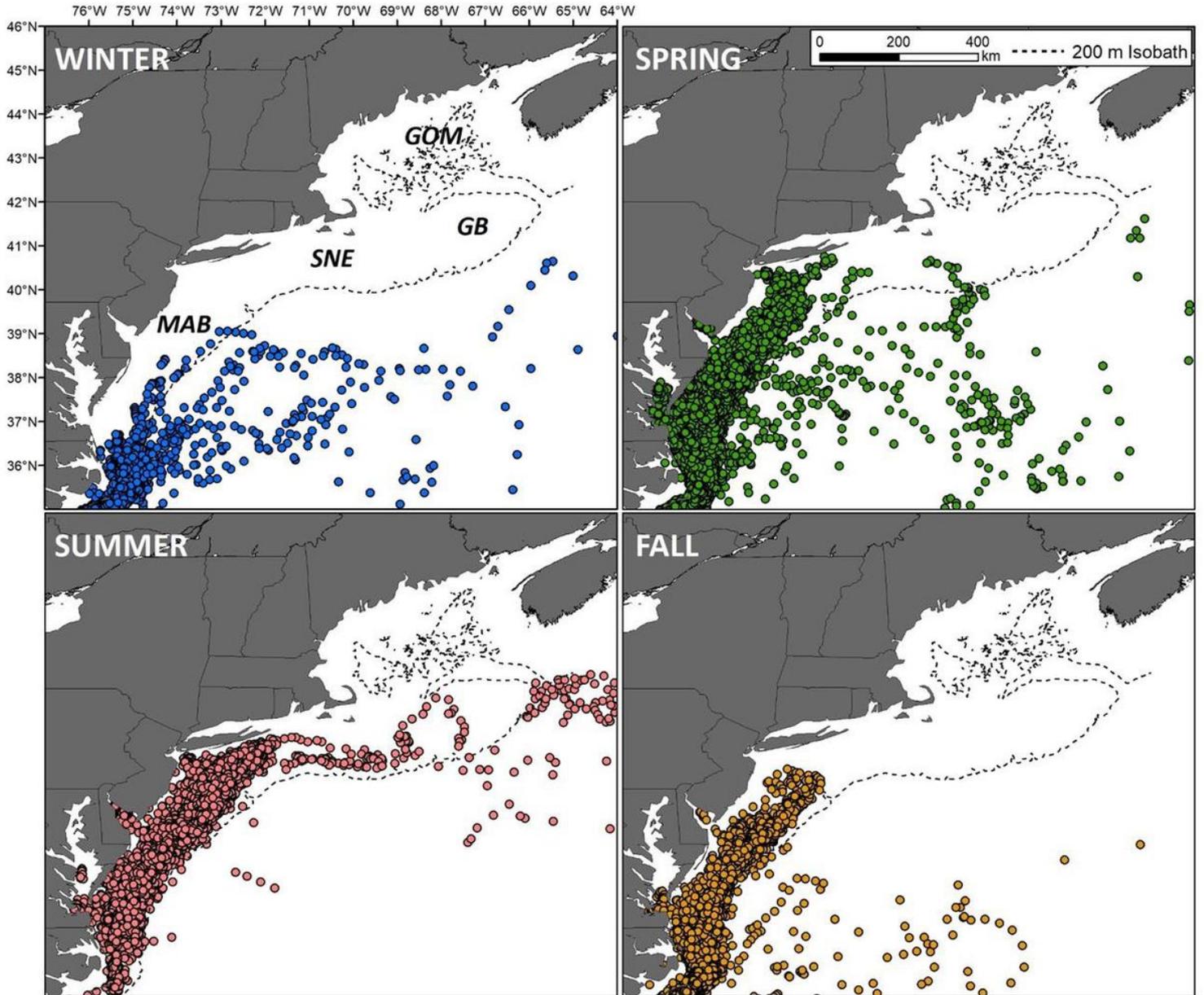
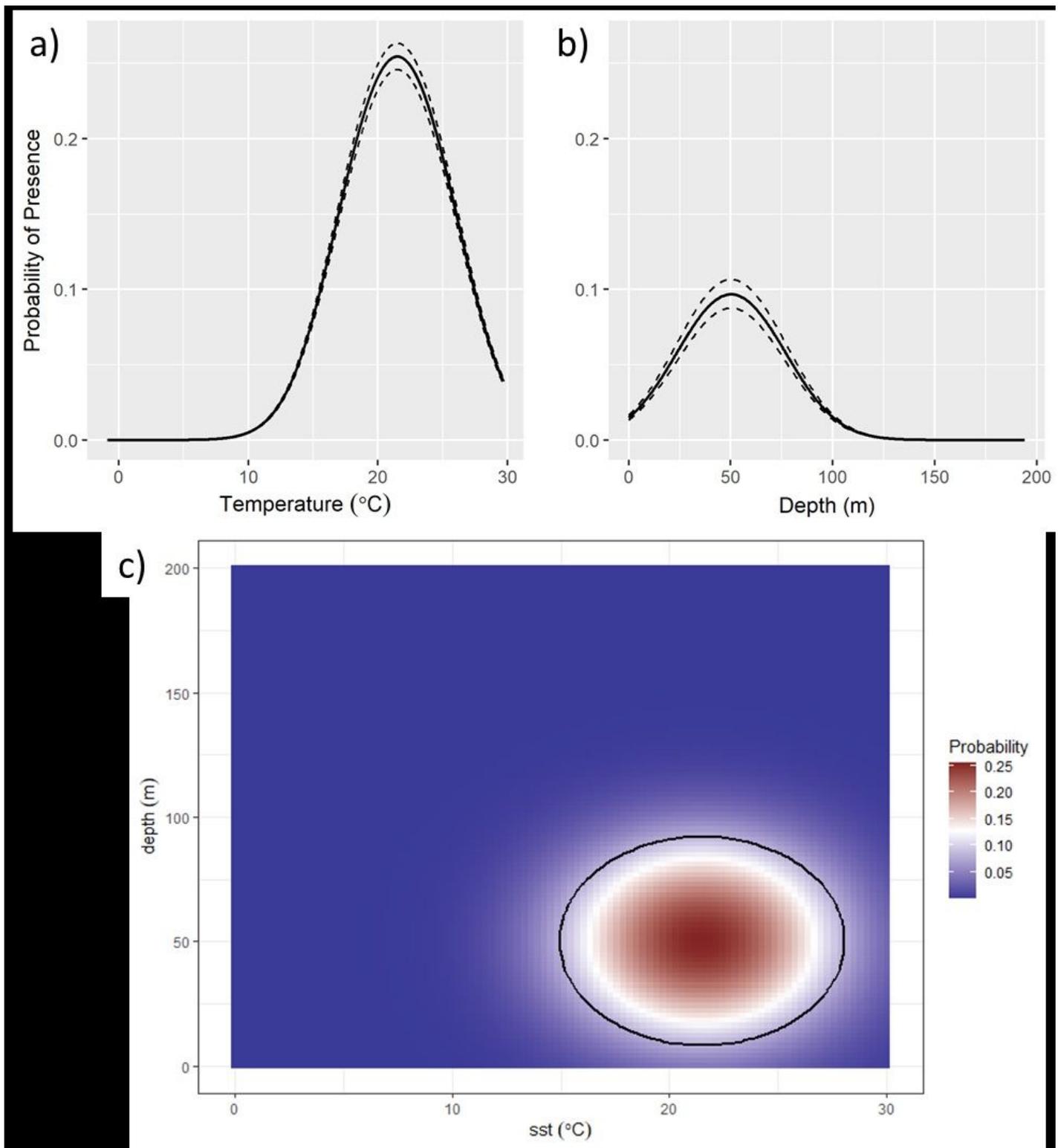


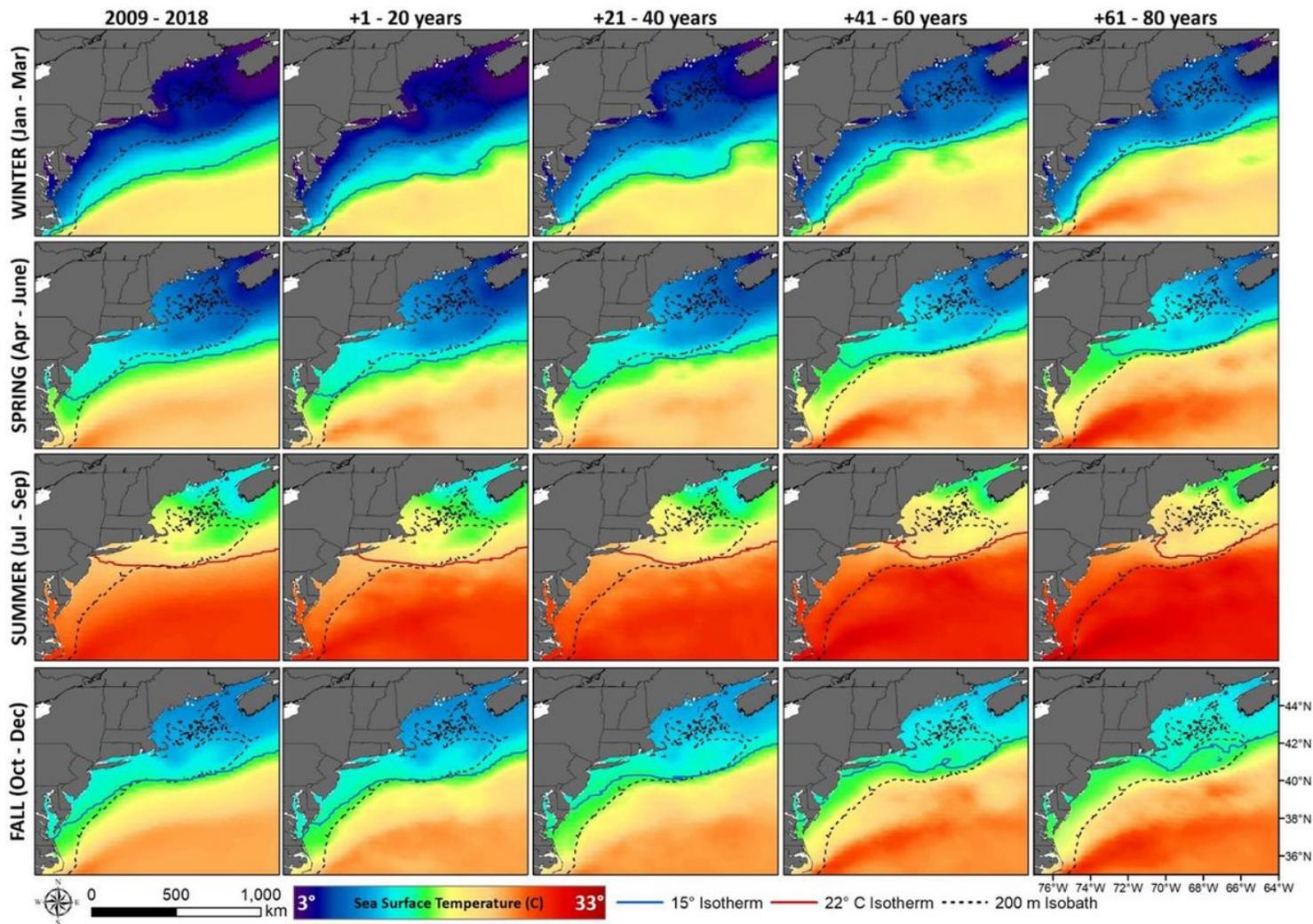
Figure 1

Reconstructed tracks from 196 loggerhead sea turtles satellite tagged between 2009-2018 within the northwest Atlantic. Dashed lines denote the 200 m bathymetric contour. GOM = Gulf of Maine, GB = Georges Bank, SNE = Southern New England, and MAB = Middle Atlantic Bight.



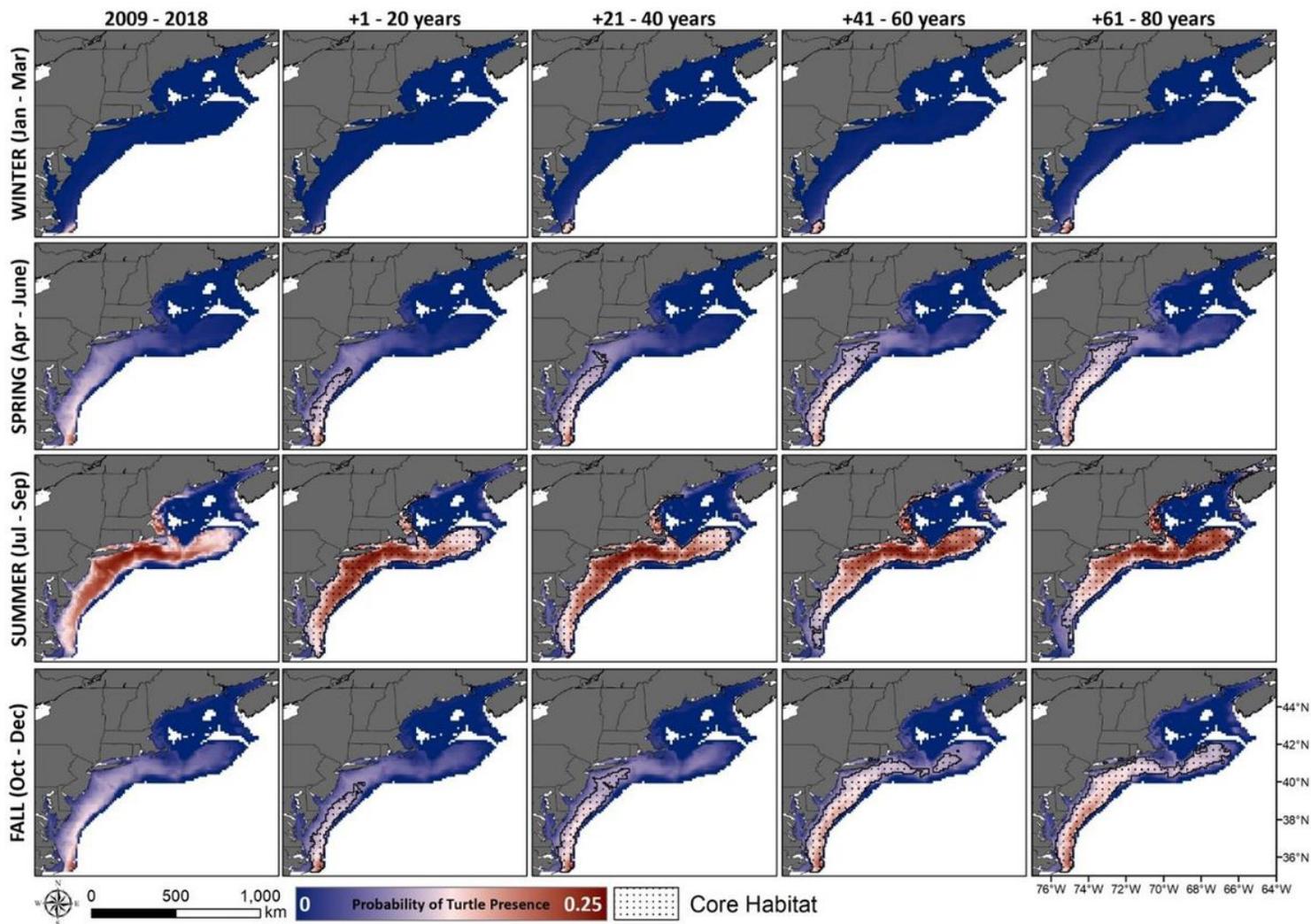
**Figure 2**

Probability of presence of loggerheads in relation to (a) sea surface temperature (SST) and (b) bottom depth. Dashed lines indicate 95% confidence intervals. The resulting core habitat as identified using the 'Index of union' is illustrated in (c), where the graph identifies the probability of loggerhead presence from observed data associated with the combined SST and depth ranges and the calculated core habitat (black circle).



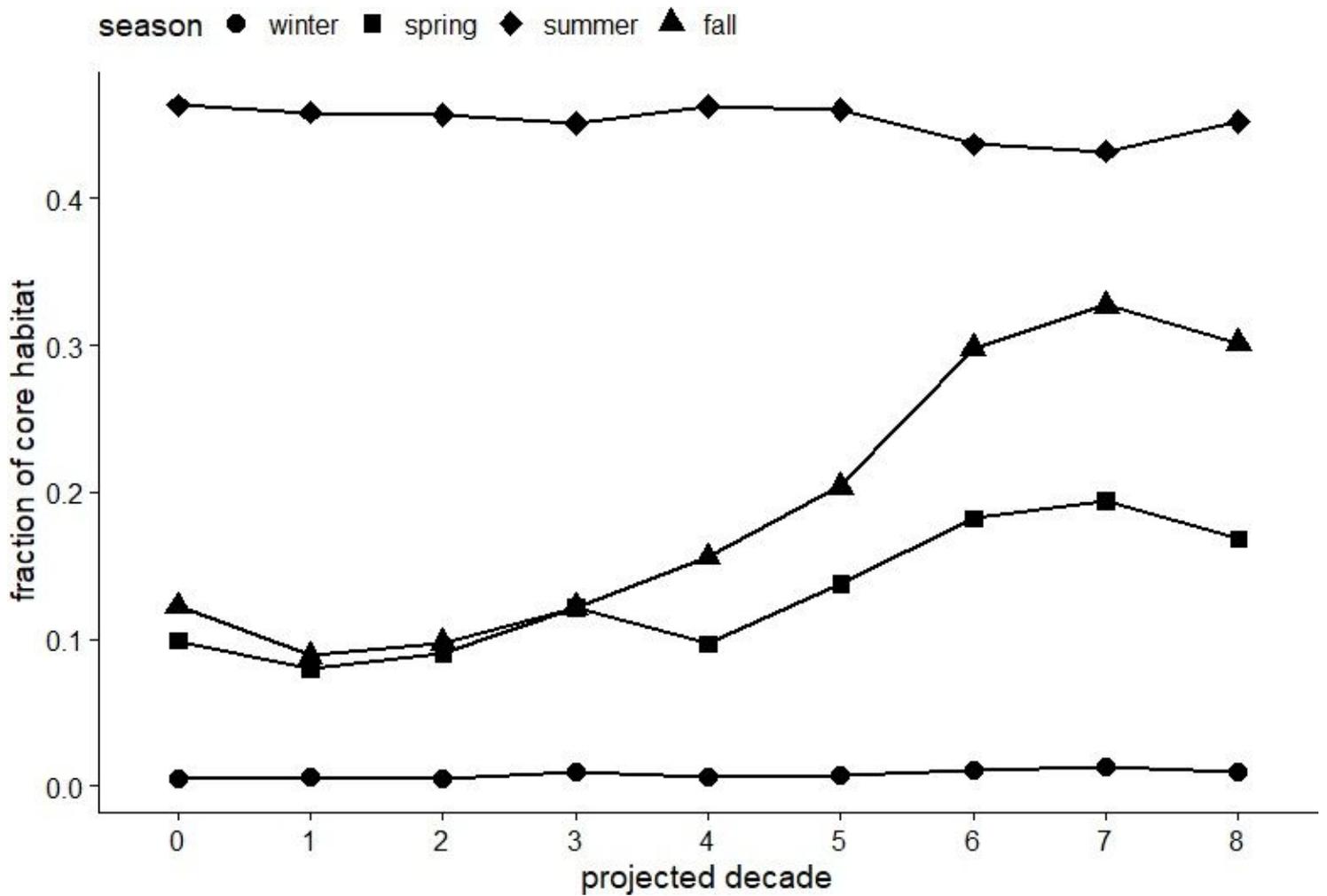
**Figure 3**

Seasonal maps of historical and projected sea surface temperature in the northwest Atlantic. The north and shoreward movement of the Gulf Stream is expected to increase with warming within shelf waters.



**Figure 4**

Seasonal maps of probability of turtle presence and core habitat based on observed and projected sea surface temperatures (SST) using the CM2.6 model. Color ramp matches Figure 2c and indicates the probability of presence based on SST and depth.



**Figure 5**

Change in fraction of the NW Atlantic shelf region identified as core habitat for loggerheads across the projected 80 years binned by decade. Spring and fall are projected to have the largest change. Decade '0' refers to the observed data.