

# Applying species distribution modelling to a data poor, pelagic fish complex: the ocean sunfishes

Phillips, N. D., Reid, N., Thys, T., Harrod, C., Payne, N. L., Morgan, C. A., ... Houghton, J. D. R. (2017). Applying species distribution modelling to a data poor, pelagic fish complex: the ocean sunfishes. DOI: 10.1111/jbi.13033

Published in: Journal of Biogeography

**Document Version:** Peer reviewed version

**Queen's University Belfast - Research Portal:** Link to publication record in Queen's University Belfast Research Portal

#### Publisher rights

© 2017 John Wiley & Sons Ltd. This work is made available online in accordance with the publisher's policies. Please refer to any applicable terms of use of the publisher.

#### General rights

Copyright for the publications made accessible via the Queen's University Belfast Research Portal is retained by the author(s) and / or other copyright owners and it is a condition of accessing these publications that users recognise and abide by the legal requirements associated with these rights.

#### Take down policy

The Research Portal is Queen's institutional repository that provides access to Queen's research output. Every effort has been made to ensure that content in the Research Portal does not infringe any person's rights, or applicable UK laws. If you discover content in the Research Portal that you believe breaches copyright or violates any law, please contact openaccess@qub.ac.uk.

1	Original Article				
2	Applying Species Distribution Modelling to a Data Poor, Pelagic Fish Complex:				
3	The Ocean Sunfishes				
4					
5	N.D. Phillips* <sup>1</sup> , N. Reid <sup>1,2</sup> , T. Thys <sup>3</sup> , C. Harrod <sup>4</sup> , N. Payne <sup>5</sup> , C. Morgan <sup>6</sup> , H.J. White <sup>1</sup> ,				
6	S. Porter <sup>1</sup> , J.D.R. Houghton <sup>1,2</sup>				
7					
8	<sup>1</sup> School of Biological Sciences, Queen's University Belfast MBC Building, 97 Lisburn Road, Belfast, BT9 7BL,				
9	U.K., <sup>2</sup> Institute of Global Food Security, Queen's University Belfast, 18-30 Malone Road, Belfast, BT9 5BN				
10	U.K., <sup>3</sup> California Academy of Science, 55 Music Concourse Drive, Golden Gate Park, San Francisco, CA				
11	94118 U.S.A., <sup>4</sup> Instituto de Ciencias Naturales Alexander von Humboldt, Universidad de Antofagasta, Avenida				
12	Angamos 601, Antofagasta, Chile, $5$ University of Roehampton, Holybourne Avenue, London, SW15 4JD				
13	Cooperative Institute for Marine Resources Studies, Oregon State University, 2030 S. Marine Science Center				
14	Newport, OR 97365				
15					
16	*Author to whom correspondence should be addressed: Natasha Phillips, Queen's University				
17	Belfast, Medical Biology Centre, 97 Lisburn Road, Belfast, BT9 7AE,				
18	nphillips01@qub.ac.uk				
19					
20	Running head: Distribution and seasonal movements of ocean sunfishes				
21					
22	Keywords: Biogeography, Environmental Niche Models, Marine, Migration, Mola,				
23	sunfishes, Spatial Ecology				

- 24 Abstract
- 25

#### 26 Aim

Conservation management of vulnerable species requires detailed knowledge of their spatial 27 28 and temporal distribution patterns. Within this context species distribution modelling (SDM) can provide insights into the spatial ecology of rarely encountered species and is used here to 29 explore the distribution pattern of ocean sunfishes (Mola mola and M. ramsayi). Both species 30 are prone to high levels of bycatch and are classified respectively as Globally Vulnerable and 31 32 Not Assessed by the IUCN; although their overall range and drivers of distribution remain poorly defined. Here, we constructed suitable habitat models for *Mola* spp. on a global scale 33 and considered how these change seasonally to provide a much needed baseline for future 34 35 management.

36

## 37 Location

38 Global.

39

## 40 Methods

Sighting records collected between 2000 and 2015 were used to build SDMs and provided the first global overview of sunfish seasonal distribution. *Post-hoc* analyses provided a quantitative assessment of seasonal changes in total range extent and latitudinal shifts in suitable habitat.

45

#### 46 **Results**

47 Mola is a widely distributed genus; however, sightings exhibited significant spatial clustering 48 most notably in coastal regions. SDMs suggested that *Mola* presence was strongly dependant 49 on sea surface temperatures with highest probability of presence between 16 and 23°C. The 50 models identified significant variation in seasonal range extent with latitudinal shifts 51 throughout the year; although large areas of suitable year-round habitat exist globally.

52

### 53 Main conclusions

We provided the first assessment of *Mola* distribution on a global scale, with evidence of a wide latitudinal range and significant clustering in localised 'hotspots' (notably between 40-56 50°N). By assessing the results of SDMs alongside evidence from published satellite tagging studies, we suggest that the species within the genus *Mola* are highly mobile, acting as facultative seasonal migrants. By identifying key suitable habitat alongside potential
movement paths, this study provides a baseline that can be used in active conservation
management of the genus.

#### 61 Introduction

Conservation management efforts are dependent on a detailed understanding of the spatial 62 distribution, biogeography and ecology of target species (Ferrier et al., 2002; Ricklefs, 2004; 63 Rushton et al., 2004). For widespread or cryptic species this can pose significant challenges 64 (Pearson et al., 2007; Rissler & Apodaca, 2007). Species distribution models (SDMs, also 65 known as ecological niche models, species-habitat models or predictive habitat models) 66 assess the complex relationship between species occurrence records and environmental 67 variation, even from limited datasets, and offers insight into habitat suitability both spatially 68 69 and temporally (Elith & Leathwick, 2009; Franklin, 2009). For little known oceanic species, such methods can provide a key starting point in understanding complex, wide-ranging 70 distribution patterns and the mechanisms driving environmental tolerances (Elith et al., 71 72 2006).

One such family of oceanic taxa, the ocean sunfishes (or Molidae), are often described as 73 rare, inactive drifters (Pope et al., 2010), however recent studies have revealed high density 74 aggregations in coastal waters (e.g. Silvani et al., 1999; Pope et al., 2010; Syväranta et al., 75 2012), sustained long distance swimming of ~48 km per day (e.g. Cartamil & Lowe, 2004; 76 Nakamura et al., 2015; Thys et al., 2015) and repeated deep-diving to mesopelagic depths 77 foraging for gelatinous prey (e.g. Cartamil & Lowe, 2004; Nakamura et al., 2015). Such 78 79 observations suggest that this is an active, highly motile taxon (Cartamil & Lowe, 2004), with 80 a broad trophic niche (e.g. Harrod et al., 2013; Nakamura & Sato, 2014; Sousa et al., 2016a) and capable of travelling significant distances in a directed manner (see review, Pope et al., 81 2010). This suggests that Mola may have more complex ecology than previously thought 82 83 (Syväranta et al., 2012), which poses broader implications for sustainable management. Such insight is important in light of current bycatch levels (Silvani et al., 1999; Cartamil & Lowe, 84 2004; Pope et al., 2010), such as the reported capture of  $> 36\ 000$  individuals per annum in 85

Mediterranean drift gillnets (Petersen & McDonell, 2007). Bycatch numbers coupled with impacts of large-scale target fisheries, led to a recent IUCN Red List classification of *Mola mola* (L. 1758) as globally Vulnerable (Jing et al., 2010) and Data Deficient in Europe (see Table 1, Appendices). This Red Listing represents a tentative first step towards future management strategies and highlights areas of sunfish ecology that require further research, such as knowledge of their distribution and movements, which currently restricts management and conservation efforts.

Anecdotal evidence collated in a review by Pope et al. (2010) suggested that the Molidae 93 94 (see Table 1. Appendices) have a pan-global distribution within temperate and tropical latitudes, although limited sighting records and inherent difficulties in species identification 95 have led to problems in delineating species-specific ranges and seasonal movement patterns. 96 97 Recent high-profile reports of ocean sunfishes at high latitudes, such as in Alaska (Dobbyn, 2015), have led many media outlets to speculate as to why these species are "suddenly" 98 appearing so far north. However, without baseline data on the range extent of ocean 99 100 sunfishes, it is difficult to know whether they have undergone recent expansion and, if so, what might be driving such changes. Although taken to be widespread (Cartamil & Lowe, 101 2004), it is not yet known if ocean sunfishes adhere consistently to a migratory paradigm 102 (whether obligate or facultative). Evidence from multiple studies, using satellite tags and 103 accelerometer derived dead-reckoning (e.g. Sims et al., 2009; Dewar et al., 2010; Nakamura 104 et al., 2015; Thys et al., 2015), suggests that *Mola* in temperate and subtropical regions may 105 move to equatorial latitudes during autumn, for example, into UK and Japanese waters. 106 However, other studies using satellite tracking (Hays et al., 2009) and dietary analysis 107 (Harrod et al., 2013) suggest year-round, or at least long-term, residence in some regions, 108 including in Mediterranean and South African waters. The results from these studies support 109

suggestions of distinct, local populations with differing drivers of distribution; however, thereis a paucity of evidence across wide spatio-temporal scales.

From a broader conservation perspective, the IUCN states that creating a "comprehensive, 112 objective global approach for evaluating the conservation status of [all] species [is important 113 in order to] inform and catalyse action for biodiversity conservation" (IUCN, 2016). In line 114 with this statement, this study uses SDM to provide an initial assessment of the global 115 distribution pattern of a vulnerable marine genus that is plagued with species-specific 116 identification problems. We present basic life history information for the genus *Mola* and its 117 118 seasonal range extent in relation to key predictive environmental parameters. This study provides an objective evidence base critical to providing a full IUCN Red Listing, upon 119 which international management decisions can be founded. 120

121

#### 122 Materials and Methods

#### 123 Data sources and manipulation

Global sightings of Mola were collected from public databases, published papers and 124 fisheries logs (see Appendix S1). A total of 14 953 sightings, recorded between the years 125 1758 and 2015, were compiled before specific criteria were set for standardising the dataset. 126 This study aimed to assess the distribution of the genus Mola which currently contains two 127 species. Mola is easily distinguishable from other genera in the Molidae (Ranzania and 128 129 *Masturus*, see Table 1. Appendices), due to its differing morphology, and therefore potential for confusion is limited. We accept that misidentification is possible, but by maintaining a 130 conservative approach to data acquisition (i.e. by removing records not identified to genus), 131 we have tried to mitigate this risk. Any incomplete records (missing location or date of 132 observation) were removed. All sighting locations were converted to decimal degrees, and 133 mapped using ARCGIS 10.3.1 (ESRI, California, USA) and all locations that erroneously fell 134

on land were removed. Although the sightings dataset extended over 257 years, 79% of
sightings occurred between 2000 and 2015. Therefore only this subset of 5 419 sightings was
retained for further analysis. These sightings were divided into each quarter of the year (JanMar, Apr-May, Jun-Aug and Sep-Dec) and matched with recent climate data available
through online data sharing platforms.

140

### 141 Environmental parameters

Climate data with near global oceanic coverage described surface oceanography at a 142 143 resolution of one decimal degree delineated as a cellular matrix. The most recently collected dynamic parameters were selected and of these, sea surface temperature, nitrate, oxygen and 144 chlorophyll concentration were averaged over three month periods suited to generating 145 146 seasonal summaries (Jan-Mar, Apr-May, Jun-Aug and Sep-Dec). The datasets included sea surface temperature averaged from 2005 to 2012 (NOAA, 2015), nitrate and oxygen 147 concentrations averaged from 1955 to 2012 (NOAA, 2015) and chlorophyll concentration 148 averaged from 2002 to 2012 (NASA, 2012). Despite the extensive coverage provided by 149 satellite data, the limitations of this dataset must be acknowledged; such as the lower quality 150 data from nearshore or frequently clouded environments (Smith et al., 2013). Of all the 151 parameters included, bathymetry was the only static variable recorded from a 2002-2003 152 global survey (NASA, 2003). If climatic data were missing from the decimal degree cell in 153 which a sighting was recorded, it was removed from the analysis (leaving n = 4 985 154 sightings). 155

156

#### 157 Data validation

Since all *Mola* data collected were 'presence only' sightings, we implemented a bias file as a proxy of survey effort to indicate the likelihood of being encountered and recorded, as

presence-absence models perform better than presence only models (Elith et al., 2006). Since 160 true absence data were not available, we followed established methods to construct a 'bias 161 file' (e.g. Phillips et al., 2009; Aguirre-Gutierrez et al., 2013: Pokharel et al., 2016). This 162 process requires the identification of a suitable proxy species (termed a target group) for 163 which further presence data were available (e.g. Ponder et al., 2001: Anderson, 2003). We 164 chose to use the leatherback turtle, Dermochelys coriacea (Vandelli, 1761) as it is suggested 165 to inhabit similar environments to ocean sunfishes (Hays et al., 2009). Moreover, the species 166 is an active predator of gelatinous zooplankton and conforms to the seasonal migration 167 168 paradigm suggested for sunfishes (see Pope review, 2009), while being subject to similar sea surface and coastal observation biases (Houghton et al., 2006; Hays et al., 2009). Leatherback 169 turtle sightings data were downloaded from the Global Biodiversity Information Facility 170 171 sightings database (GBIF, 2015). The use of target group data has been reported to provide a considerable improvement in model performance, providing more realistic data than taking 172 pseudo-absences from sites that have not been sampled at all (e.g. Phillips, 2009; Mateo et 173 al., 2010; Aguirre-Gutierrez et al., 2013). The rationale here is that leatherback sightings 174 provided a proxy for recorder presence with the inference that ocean sunfish sightings would 175 have been recorded concurrently if present. Correspondingly, these locations were used to 176 generate ocean sunfish pseudo-absence data (n = 434) to train SDMs. 177

178

#### 179 Statistical Analysis & SDMs

180 The distribution of *Mola* was mapped globally and a minimum convex hull containing all 181 sightings created to satisfy the IUCN Red List range map requirements. Owing to the cryptic 182 speciation within *Mola*, such range mapping was constrained to genus level.

183 A cluster analysis of sightings was performed using a Clark-Evans nearest neighbour test
184 (Clark & Evans, 1954) using the R x64 3.2.2 (R Development Core Team, 2008) package

'spatstat' (Baddeley et al., 2015). The degree of grouping was determined using a correction
cumulative distribution function and a Monte Carlo test to provide a probability value.

Climatic data were tested for collinearity using Pearson's correlation, before SDMs were 187 produced using the R package 'Biomod2' (Thuiller et al., 2015). Seven SDM types were 188 assessed including: surface range envelopes (SRE, quant = 0.025), classification tree analysis 189 (CTA, CV.tree = 50), random forest (RF), multiple adaptive regression splines (MARS), 190 flexible discriminant analysis (FDA), generalised linear models (GLM, type = simple) and 191 generalised additive models (GAM, spline = 3). The models were designed with an 80:20192 193 data split for training and testing and run with a 5 000 fold cross validation. All models used in Biomod2 were run using the default settings recommended by Thuiller et al. (2010). Using 194 this model design, the seasonal distribution of Mola was predicted using matched sightings 195 196 and environmental data from each quarter of the year.

Model evaluation statistics were calculated including the Kappa value (k), true skill 197 statistic (TSS) and area under the curve (AUC) of the receiver operating characteristic 198 (ROC). These evaluation metrics are frequently used to evaluate SDM performance, although 199 AUC values have recently been criticised for overestimating performance by including large 200 areas of absence data (Lobo et al. 2008; Leach et al. 2015). Popular alternatives also have 201 limitations, such as TSS which is calculated from sensitivity and specificity, which 202 themselves can contain misleading commission errors (Leach et al. 2015). The Kappa value 203 provides a more objective measure of prediction accuracy, although this can also produce 204 commission errors (Leach et al. 2015), but it provides accepted thresholds used in model 205 evaluation. Here, we present each evaluation metric for all models however, the final 206 207 evaluation of model accuracy used Kappa.

The optimal SDM was selected from those with a Kappa > 0.4 (see Table 2), as this threshold has been widely used in a range of published work (Landis et al., 1977; Altman, 1990; Allouche et al., 2006; Leach et al. 2015). The random forest model was the single best
approximating model selected for further analysis and re-run with 100% of the sightings data
to predict the seasonal probability of *Mola* presence globally.

To assess the seasonal range extent of Mola, the proportion of cells predicted with a 213 probability of presence > 0.7 was calculated and tested with a 4-sample test for equality of 214 proportions without continuity correction. As the distribution data were strongly skewed, 215 non-parametric tests were used. Due to uneven sampling, data were divided into Northern 216 and Southern Hemispheres and the predicted range extent of *Mola* examined by plotting box 217 218 and whisker diagrams of the latitudinal range divided by season and compared statistically using a Kruskal-Wallis test. To assess if individual *Mola* move seasonally in accordance with 219 the model predictions, the latitude of all sightings were plotted against the Julian day of the 220 221 year on which they were recorded and fitted with a locally weighted scatterplot smoothing curve (LOESS). 222

223

#### 224 **Results**

*Mola* observations were distributed globally (Fig. 1a and b) but with significant clustering (z = 0.335, p < 0.05), with aggregations in North American and European coastal waters predominately between 20-60°N, and peaking at 50°N (Fig. 2).

Nitrate and oxygen concentrations were significantly correlated (r = 0.88, p < 0.001), and since nitrate is used here as a proxy for productivity, it was removed to avoid leverage in statistical models. The random forest model had the highest model evaluation statistic values (mean values of 5 model runs: Kappa = 0.63, TSS: 0.72, ROC: 0.93) and were thus chosen as the optimal SDM technique.

Random forest LOESS curves suggested *Mola* presence was associated with shallow,
 temperate (7-23°C), relatively low productivity (chlorophyll < 125mg/m<sup>3</sup>), oxygen rich (>

4ml/L) coastal waters (Fig. 3a-d). However, cells predicted to have a probability of presence 235 > 0.7 were widespread in all seasons resulting in a pan-global distribution in surface waters 236 (Fig. 4a and b); but with lowest occurrence in polar and equatorial waters. The extent of 237 suitable habitat (defined as the percentage suitable ocean surface) varied significantly 238 between seasons ( $\chi^2_{df=3}$  = 591.2, p < 0.001; Table 3). The latitudinal range of *Mola* also 239 varied significantly in both Northern (tested individually) across all seasons ( $\chi^2_{df=3} = 1690.5$ , 240 tabulated  $\chi^2_{df=3} = 8.81e^{-11}$ , p < 0.001) and Southern Hemispheres ( $\chi^2_{df=3} = 3121.2$ , tabulated 241  $\chi^2_{df=3} = 8.81e^{-11}$ , p < 0.001). Seasonal differences in latitudinal range reflected movement 242 243 patterns, with the latitude of individual sightings varying temporally in both the Northern and Southern Hemispheres with animals shifting to more northerly latitudes in both hemispheres 244 between April and October (Figs. 5a and b). 245

246

#### 247 **Discussion**

This study used detailed records from public sightings databases, alongside fisheries surveys 248 and museum archives which provided global coverage of a Data Deficient genus (IUCN, 249 2016). Although public sightings are widely used in broad-scale ecological studies, such data 250 come with caveats, such as potential misidentification of cryptic species, incorrect data entry 251 or regions of limited data availability. Despite such restrictions, such citizen science 252 initiatives offer extensive coverage well beyond the budget and feasibility of most research 253 254 projects. One of the best known examples, the North American Christmas Bird Count, has been running for over 100 years, with millions of person hours contributed to survey effort 255 (Bibby, 2003; Audubon, 2008). With careful interpretation and strict data processing, 256 substantial quantities of data can be collated over wide spatial and temporal scales, to the 257 same quality as those collected by experts (Danielsen et al., 2014). 258

When applying SDM to sightings data, we must be aware of the limitations of the dataset in 259 question, choose ecologically relevant variables (Mac Nally, 2000) and use appropriate 260 methods (Elith & Leathwick, 2009). However, despite potential pitfalls and limitations, SDM 261 have become important tools for predicting species distribution patterns and subsequent 262 conservation management (Kremen et al., 2007; Wiens & Graham, 2005; Evans et al., 2015). 263 In this study, SDM enabled us to delineate the range extent of ocean sunfishes, quantify 264 distinct local clustering and describe seasonal changes in range extent accompanied with 265 intra-annual movement patterns consistent with being a facultative seasonal migrant. 266

267

#### 268 Distribution patterns

To date, there are two recognised species within Mola: Mola mola (L. 1758) and Mola 269 ramsayi (Giglioli, 1883). Alongside these two species, recent papers have reported 270 differences between the Atlantic and Pacific M. mola populations based on genetic and 271 morphological studies (e.g. Bass et al., 2005; Yoshita et al., 2009; Gaither et al., 2016). 272 Despite these discoveries, a formal classification of cryptic species is yet to be published, and 273 the species taxonomy of *Mola* remains in flux (see review by Pope, 2009). In light of the 274 current pressures faced by the ocean sunfishes, this study provides baseline information on 275 *Mola* spatial ecology, which can be further refined to species-specific level as discrepancies 276 over speciation resolve themselves over time. 277

Our study revealed that the genus *Mola* has a wide habitat range (see Fig. 1b) with confirmed sightings records extending 128° of latitude from approximately 70°N near Altenfjord, Norway to -58°S in the Beagle Canal, Chile (sightings contributed by Lukas Kubicek, *pers. comm.*). When compared to the latitudinal range extents of > 10 000 other marine species (Strona et al., 2012), this range would appear in the top 15 range extents (maximum reported range 150° latitude). However, within this range, our analysis suggests that *Mola* frequently aggregate and cluster in specific regions rather than being distributed randomly. Such clustering may be partly an artefact of sighting bias in coastal regions and known hotspots, particularly in North American and Europe. Nonetheless, the findings presented here align well with anecdotal evidence that *Mola* occur in patchily distributed, high density aggregations, particularly in coastal waters (e.g. Silvani et al., 1999; Sims & Southall, 2002; Houghton et al., 2006).

Several regions globally have already been identified as hosting annual aggregations of Mola 290 *mola*, suggested to be shoals of juveniles (< 1 m); for example in Camogli, Italy (Syväranta et 291 al., 2012) and California, USA (Cartamil & Lowe, 2004; Thys et al., 2015), whilst our 292 analysis may help predict other areas with high density populations. We are aware that 293 limited data availability such as sparse information from equatorial regions, may have a 294 partial effect on our habitat suitability predictions, but this is likely reduced by our 295 implementation of a bias file. To the best of our knowledge, we have defined the full range 296 extent of Mola (Fig. 1b), however, as sightings were likely subject to significant observer 297 bias. Indeed, the predicted presence from SDMs (Fig. 4b) may be of greater use to 298 characterise the actual range extent *Mola* populations whilst predicted probability values are 299 likely correlated with density. 300

301

## 302 Environmental drivers of *Mola* distribution

The Random Forest model provided the most reliable approximation of *Mola* distribution. Sea surface temperature and an indicator of regional productivity (chlorophyll *a* concentration) have been proposed as primary drivers of *Mola* movements (e.g. Thys et al., 2015; Sims et al., 2009). *Mola* habitat suitability increased gradually with chlorophyll *a* concentration until reaching a threshold of approximately 140 mg m<sup>-3</sup> with habitat suitability declining rapidly at higher concentrations. Many studies comment on *Mola* range limitation

in terms of minimum temperatures, and indeed we found sightings of *Mola* to be absent from 309 waters below 7°C. However, our data suggested that Mola have a similarly-defined upper 310 thermal threshold, of approximately 23°C, beyond which habitat suitability declined rapidly. 311 In the Atlantic, M. mola were found to spend ~99% of their time in water temperatures 312 between 10 - 19°C over a three month period (Sims et al., 2009), with a similar thermal 313 preference of 16 - 17°C suggested from Pacific studies (Nakamura et al., 2015). The 314 suggested thermal preference of approximately 16°C is further supported by our results, with 315 habitat suitability peaking at this value. Interestingly, the warmest ambient water conditions 316 317 recorded by external data loggers on free swimming M. mola, was 22°C (Nakamura et al., 2015) with internal body temperatures ranging from 12 - 21°C; considerably narrower than 318 external ambient water temperatures experienced by the fish (3 - 22°C). More recently, a 319 study on spatial occupancy of tagged M. mola in the North East Atlantic suggested 320 movements were strongly related to water temperature on regional scales with an "escape" 321 from regional maxima of approx. 25°C (Sousa et al., 2016b). By combining such evidence 322 alongside the modelled thermal response curves, we suggest that the genus *Mola* may have an 323 upper thermal tolerance limit of approximately 23°C, although occasional forays above this 324 temperature may occur as demonstrated by the recording of an individual M. ramsayi at a 325 maximum of 27.5°C (Thys et al., 2016). Further support for a thermal optimum of 16°C can 326 be derived from a recent study comparing optimum temperatures for performance in the wild 327 to maximum temperature experiences in fish species' ranges (Payne et al., 2016a). If a 328 thermal optimum of 16°C is aligned with the expected response curve, then an upper thermal 329 limit of 23°C would be expected from this genus (Payne et al., 2016b). The thermal limits 330 identified in our study may, therefore, reflect a loss of performance beyond such limits, at a 331 genus level, although further research will be required to confirm species specific responses. 332

From post-hoc analysis of the range extent of Mola, it appears that presence is also 333 associated with dissolved oxygen levels between 5 and 7 ml/L. However, Thys et al. (2015) 334 recently suggested that M. mola may be able to tolerate very low oxygen levels after 335 observing individuals within ocean hypoxic zones at 60 m. Following periods exposed to 336 such conditions, it is likely that individuals may need to recover in well-oxygenated waters 337 (Cartamil & Lowe, 2004). To date, Mola mola and Mola ramsayi have been observed at 338 maximum depths of 844 m (Potter & Howell, 2011) and 483 m respectively (Phillips et al., 339 2015), suggesting that mesopelagic ranging of sunfishes is perhaps more common than 340 341 previously thought (Phillips et al., 2015). However, although the Mola are capable of deep water ranging, large schools of small Mola spp. are often noted in coastal areas, possibly a 342 reflection of their mixed diet at this life stage (e.g. Syväranta et al., 2012; Harrod et al., 2013; 343 344 Nakamura & Sato, 2014). The increased availability of benthic prey and discards in coastal waters may function as a driver of seasonal abundance in shallow water in the genus Mola 345 (Harrod et al., 2013). 346

347

#### 348 Seasonal movements

We identified large areas of suitable habitat available year-round for Mola, however, our 349 results also suggested that the total suitable sea surface area and latitudinal position of varied 350 significantly between seasons (see Fig. 5a). The predictive models (see Fig. 4) suggested that 351 352 Mola thermal tolerance enables movement to higher latitudes in the Northern Hemisphere during the boreal spring to late summer, before retreating further south over the boreal 353 autumn and winter months. Within the confines of this study, we were only able to model 354 Mola presence in surface waters, however, these latitudinal movements may correspond to 355 shifts in deep prey fields (Angel & Pugh, 2000; Houghton et al., 2008). Our predicted 356 seasonal movement of Mola supports evidence from tagging studies in the northwest and 357

northeast Atlantic (e.g. Sims et al., 2009; Potter & Howell, 2011; Sousa et al., 2016b), and 358 northeast and northwest Pacific (e.g. Dewar et al., 2010; Thys et al., 2015), which identified 359 seasonal movements of individuals driven by temperature and patchily distributed prey. 360 However, despite a range of tagging studies providing data across the Northern Hemisphere, 361 there are relatively few data available from the Southern Hemisphere on Mola movements. 362 From the SDMs, we suggest that a similar pattern occurs in the Southern Hemisphere, where 363 Mola are able to move to maximum southern latitudes during the austral spring to late 364 summer and then retreat towards the equatorial regions during the austral winter (Fig. 5a and 365 366 b). These broad scale movements reflect the migration patterns of many species, in accordance with the seasonal migratory paradigm, where warmer temperatures during 367 summer months enable range extensions poleward, and which then contract as the seasons 368 369 change; example species include bluefin tuna (Lutcavage et al., 1999), swordfish (Sedberry et al., 2001) and loggerhead turtles (Mansfield et al., 2009). 370

Our data suggest that although the average latitudinal position of *Mola* in surface waters 371 varied over the seasons, much of the world's oceans remain suitable for Mola year-round, 372 with a wide latitudinal range. It is apparent, therefore, that Mola cannot be classified as 373 obligate migrants, owing to discrepancies in distribution between populations. Although the 374 species within this study were all considered to be Mola mola, the more common of the two 375 *Mola* species, inferred differences in movement strategy between populations may be due to 376 377 misidentification and behavioural differences between M. mola and the lesser studied M. ramsayi (Pope et al., 2010). Mola ramsayi is morphologically very similar to M. mola (Bass 378 et al., 2005), identified by 16 fin rays with 12 closely spaced ossicles, compared to the 12 fin 379 rays and 8 broadly spaced ossicles and reduced band of denticles prior to the clavus of M. 380 mola (Fraser-Brunner, 1951; Thys et al., 2013). Mola ramsayi was initially suggested to be 381 the Southern Hemisphere species (Fraser-Brunner, 1951), however, individuals have since 382

been identified in the Northern Hemisphere, including the Sea of Oman (Al Ghais, 1994), the Indian waters of Chennai (Mohan et al., 2006) and even co-occurring with *M. mola* (Bass et al., 2005). Further molecular genetic analyses are required to confirm species identification and to assess the movement ecology of these species (Pope et al., 2010).

Alongside the predicted distribution patterns modelled here, the average position of Mola 387 raw sightings was consistent with the concept of seasonal migration. However, outliers to this 388 pattern do exist, supported by evidence of prolonged residency (e.g. Hays et al., 2009; Harrod 389 et al., 2013). Since this study only assessed *Mola* surface distribution, it does not provide 390 391 information on depth distribution, however several studies suggest that Mola spends a significant proportion of time (up to 30%) in surface waters less than 10 m deep (Potter & 392 Howell, 2010). Although sightings data alone will be insufficient to fully determine the 393 394 seasonal distribution patterns of marine species (Southall et al., 2005), the frequent sightings of Mola in surface waters is related to their universal basking behaviour at the sea surface 395 (Norman & Fraser, 1938). We suggest that the surface prediction of *Mola* distribution will 396 provide a useful measure of their global distribution. 397

Although the results of this study do not provide direct evidence of a reciprocal migration, they do support the suggestion that some populations move latitudinally as suitable conditions shift over the course of the year. Such long distance movements may be restricted to populations near the latitudinal limits of their distribution; however, further study is required to test this assertion. Taken together, these results suggest that the genus *Mola* contains populations subject to differing drivers of distribution and, therefore, we propose they may be classed as facultative seasonal migrants.

405

#### 406 Conclusions

This study provides a first assessment of the spatio-temporal global biogeography of the 407 genus Mola. Taken together, our results suggest that the genus is globally distributed with 408 significant clustering in specific locations, influenced by sea surface temperatures ranging 409 from ~7 to 23°C. Based on SDMs, we suggest that populations act as facultative seasonal 410 migrants with differing regional drivers of distribution. Although this study was able to 411 consider the potential influence of productivity (using the proxy variable of chlorophyll 412 concentration), future work may be able to assess smaller regions which have better data 413 availability. Further studies on the ontogenetic shifts in the diet of ocean sunfishes are also 414 415 required to integrate SDMs with international databases of putative prey items to explore the life history significance of shallow water and offshore habitats in more detail. 416

417

### 418 Acknowledgements

The authors would like to acknowledge the generous contribution of all those who collected and provided data for this study (for full list please see Appendix S1), particularly the thousands of citizen scientists globally, without whom this work would not be possible. Many thanks also to Lawrence Eagling for proof-reading the manuscript on several occasions. N.D.P was funded by The Fisheries Society of the British Isles.

#### 424 **References**

- Al-Ghais, S. (1994) A first record of *Mola ramsayi* (Osteichthyes: Molidae) for the United
  Arab Emirates. *Tribulus*, 4, 22.
- 427 Allouche, O., Tsoar, A., Kadmon, R. (2006) Assessing the accuracy of species distribution
- 428 models: prevalence, kappa and the true skill statistic (TSS). *Journal of Appliped Ecology*,

**429 43**, 1223-1232.

- Altman, D.G. (1990) Practical statistics for medical research. Chapman and Hall, CRC PressLondon.
- 432 Anderson, R.P. (2003). Real vs. artefactual absences in species distributions: tests for
- 433 Oryzomys albigularis (Rodentia: Muridae) in Venezuela. Journal of Biogeography, 30,

434 591–605.

Angel M.V. & Pugh P.R. (2000) Quantification of diel vertical migration by macroplankton
and micronektonic taxa in the northeast Atlantic. *Hydrobiologia*, 440, 161–179.

437 Audubon (2008) Available at: http://www.audubon.org/Bird/cbc/.

- Baddeley A., Rubak E., & Turner R. (2015) Spatial Point Patterns: Methodology and
  Applications with R.
- 440 Bass A.L., Dewar H., Thys T., Streelman J.T., & Karl S. a. (2005) Evolutionary divergence

among lineages of the ocean sunfish family, Molidae (Tetraodontiformes). *Marine Biology*, 148, 405–414.

- Bibby C.J. (2003) Fifty years of Bird Study : Capsule Field ornithology is alive and well, and
- 444 in the future can contribute much more in Britain and elsewhere. *Bird Study*, **50**, 194–
  445 210.
- Cartamil D.P. & Lowe C.G. (2004) Diel movement patterns of ocean sunfish Mola mola off
  southern California. *Marine Ecology Progress Series*, 266, 245–253.
- 448 Danielsen F., Pirhofer-Walz K., Adrian T.P., Kapijimpanga D.R.K., Burgess N.D., Jensen

- P.M., Bonney R., Funder M., Landa A., Levermann N., & Madsen J. (2014) Linking
  public participation in scientific research to the indicators and needs of international
  environmental agreements. *Conservation Letters*, 7, 12–24.
- 452 Dobbyn P. (2015) Available at: http://www.ktuu.com/news/news/a-series-of-bizarre-fish453 sightings-reported-in-alaska/35733240.
- 454 Elith J., Graham C., Anderson R., Dudik M., Ferrier S., Guisan A., Hijmans R., Huettmann
- 455 F., Leathwick J., Lehmann A., Li J., Lohmann L., Loiselle B., Manion G., Moritz C.,
- 456 Nakamura M., Nakazawa Y., Overton J., Peterson A., Phillips S., Richardson K.,
- 457 Scachetti-Pereira R., Schapire R., Soberon J., Williams S., Wisz M., & Zimmermann N.
- 458 (2006) Novel methods improve prediction of species' distributions from occurrence
- 459 data. *Ecography*, **29**, 129–151.
- 460 Elith J. & Leathwick J.R. (2009) Species Distribution Models: Ecological Explanation and
- 461 Prediction Across Space and Time. *Annual Review of Ecology, Evolution, and*
- 462 *Systematics*, **40**, 677–697.
- 463 Ferrier S., Biology S., & Apr N. (2002) Mapping Spatial Pattern in Biodiversity for Regional
- 464 Conservation Planning : Where to from Here ? Mapping Spatial Pattern in Biodiversity
- 465 for Regional Conservation Planning : Where to from Here ? **51**, 331–363.
- 466 Franklin J. (2009) Mapping species distributions: spatial inference and prediction. *Board*467 *Member of Landscape Ecology Journal of Vegetation Science*, 336.
- 468 Fraser-Brunner A. (1951) The ocean sunfishes (Family Molidae). *Bull Br Mus (Nat Hist)*469 Zoo, 87–121.
- 470 GBIF (2015) Available at: http://doi.org/10.15468/dl.q3wkgk.
- 471 Harrod C., Syväranta J., Kubicek L., Cappanera V., & Houghton J.D.R. (2013) Reply to
- 472 Logan & Dodge: "stable isotopes challenge the perception of ocean sunfish Mola mola
- as obligate jellyfish predators". *Journal of fish biology*, **82**, 10–6.

- 474 Hays G.C., Farquhar M.R., Luschi P., Teo S.L.H., & Thys T.M. (2009) Vertical niche
- 475 overlap by two ocean giants with similar diets: Ocean sunfish and leatherback turtles.
- 476 *Journal of Experimental Marine Biology and Ecology*, **370**, 134–143.
- 477 Houghton J.D.R., Cedras A., Myers A.E., Liebsch N., Metcalfe J.D., Mortimer J. a., & Hays
- 478 G.C. (2008) Measuring the state of consciousness in a free-living diving sea turtle.
- 479 *Journal of Experimental Marine Biology and Ecology*, **356**, 115–120.
- 480 Houghton J.D.R., Doyle T.K., Davenport J., & Hays G.C. (2006) The ocean sunfish Mola
- 481 mola: insights into distribution, abundance and behaviour in the Irish and Celtic Seas.
- 482 *Journal of the Marine Biological Association of the UK*, **86**, 1237.
- 483 IUCN (2016) Available at: http://www.iucnredlist.org/about/overview.
- Jing, L., Zapfe, G., Shao, K.-T., Leis, J.L., Matsuura, K., Hardy, G., Liu, M., Robertson, R. &
- 485 Tyler J. (2010) Available at: http://www.iucnredlist.org/.
- 486 Nakamura I., Goto Y., & Sato K. (2015) Ocean sunfish rewarm at the surface after deep
- 487 excursions to forage for siphonophores. *Journal of Animal Ecology*, **84**, 590–603.
- 488 Nakamura I. & Sato K. (2014) Ontogenetic shift in foraging habit of ocean sunfish Mola
- 489 mola from dietary and behavioral studies. *Marine Biology*, **161**, 1263–1273.
- 490 NASA (2003) Available at:
- 491 http://neo.sci.gsfc.nasa.gov/view.php?datasetId=GEBCO\_BATHY.
- 492 NASA (2012) Available at:
- 493 http://neo.sci.gsfc.nasa.gov/view.php?datasetId=MY1DMM\_CHLORA.
- 494 NOAA (2015) Available at: http://www.nodc.noaa.gov/OC5/woa13/woa13data.html.
- 495 Payne N.L., Smith J.A., van der Meulen D.E., Taylor M.D., Watanabe Y.Y., Takahashi A.,
- 496 Marzullo T.A., Gray C.A., Cadiou G., & Suthers I.M. (2015) Temperature-dependence
- 497 of fish performance in the wild: links with species biogeography and physiological
- 498 thermal tolerance. *Functional Ecology*, .

499	Pearson R.G., Raxworthy C.J., Nakamura M., & Townsend Peterson A. (2007) Predicting
500	species distributions from small numbers of occurrence records: A test case using
501	cryptic geckos in Madagascar. Journal of Biogeography, 34, 102–117.
502	Petersen S. & McDonell Z. (2007) A bycatch assessment of the Cape horse mackerel
503	Trachurus trachurus capensis mid- water trawl fishery off South Africa. Birdlife/WWF
504	Responsible Fisheries Programme Report 2002–2005, .

- Phillips N.D., Harrod C., Gates A.R., Thys T.M., & Houghton J.D.R. (2015) Seeking the sun
  in deep, dark places: Mesopelagic sightings of ocean sunfishes (Molidae). *Journal of*
- 507 *Fish Biology*, **87**, 1118–1126.
- 508 Pope E.C., Hays G.C., Thys T.M., Doyle T.K., Sims D.W., Queiroz N., Hobson V.J.,
- 509 Kubicek L., & Houghton J.D.R. (2010) The biology and ecology of the ocean sunfish
- 510 Mola mola: A review of current knowledge and future research perspectives. *Reviews in*

511 *Fish Biology and Fisheries*, **20**, 471–487.

- 512 Potter I.F. & Howell W.H. (2011) Vertical movement and behavior of the ocean sunfish,
- 513 Mola mola, in the northwest Atlantic. *Journal of Experimental Marine Biology and*
- 514 *Ecology*, **396**, 138–146.
- 515 R Development Core Team (2008) R Development Core Team. .
- 516 Ricklefs R.E. (2004) A comprehensive framework for global patterns in biodiversity. *Ecology*517 *Letters*, 7, 1–15.
- 518 Rissler L.J. & Apodaca J.J. (2007) Adding more ecology into species delimitation: ecological
- 519 niche models and phylogeography help define cryptic species in the black salamander
- 520 (Aneides flavipunctatus). *Systematic Biology*, **56**, 924–942.
- Rushton S.P., Merod S.J.O.R., & Kerby G. (2004) New paradigms for modelling species
  distributions? *Journal of Applied Ecology*, 41, 193–200.
- 523 Silvani L., Gazo M., & Aguilar a. (1999) Spanish driftnet fishing and incidental catches in

- 524 the western Mediterranean. *Biological Conservation*, **90**, 79–85.
- 525 Sims D.W., Queiroz N., Doyle T.K., Houghton J.D.R., & Hays G.C. (2009) Satellite tracking
- of the World's largest bony fish, the ocean sunfish (Mola mola L.) in the North East
  Atlantic. *Journal of Experimental Marine Biology and Ecology*, **370**, 127–133.
- Sims D.W. & Southall E.J. (2002) Occurrence of ocean sun ¢ sh, Mola mola near fronts in
  the western English Channel. *J. Mar. Biol. Ass. U.K*, 82, 927–928.
- 530 Sousa L.L., Queiroz N., Mucientes G., Humphries N.E., & Sims D.W. (2016) Environmental

influence on the seasonal movements of satellite-tracked ocean sunfish Mola mola in the
north-east Atlantic. *Animal Biotelemetry*, 4, 7.

- 533 Syväranta J., Harrod C., Kubicek L., Cappanera V., & Houghton J.D.R. (2012) Stable
- isotopes challenge the perception of ocean sunfish Mola mola as obligate jellyfish
  predators. *Journal of fish biology*, **80**, 225–31.
- Thuiller A.W., Georges D., Engler R., Georges M.D., & Thuiller C.W. (2015) Package
  "biomod2.".
- 538 Thys T.M., Ryan J.P., Dewar H., Perle C.R., Lyons K., O'Sullivan J., Farwell C., Howard
- 539 M.J., Weng K.C., Lavaniegos B.E., Gaxiola-Castro G., Miranda Bojorquez L.E., Hazen
- 540 E.L., & Bograd S.J. (2015) Ecology of the Ocean Sunfish, Mola mola, in the southern
- 541 California Current System. Journal of Experimental Marine Biology and Ecology, 471,
- 542 64–76.
- 543 **Supporting information**
- 544 Appendix S1 Data sources table of global sunfish sightings
- 545
- 546 Biosketch
- 547 Natasha Phillips is a PhD researcher at the University of Belfast interested in the movement
  548 ecology, diet and energetics of ocean sunfishes (family Molidae).
- 549

- 550 Editor and Handling Editor
- 551 Michelle Gaither and Şerban Procheş
- 552

# 553 Author contributions:

- 554 NDP, JDRH, TT, NR and CH conceived the ideas. TT and CM collected data. NDP and JH
- 555 led the writing. HJW, SP, NR, NP advised on analysis. NDP analysed the data.

		IUCN Red Listing		
Species	Common name	Global Scale	European Scale	
Mola mola (L. 1758)	Ocean sunfish	Vulnerable	Data Deficient	
Mola ramsayi (Giglioli 1883)	Southern ocean sunfish	Not Assessed	Not Assessed	
Masturus lanceolatus (Liénard 1840)	Sharptail sunfish	Least Concern	Not Assessed	
Ranzania laevis (Pennant 1776)	Slender sunfish	Least Concern	Data Deficient	

**Table 1.** IUCN Red List designation for ocean sunfishes on both global and European scales.

SDM type Kappa Value **TSS Value ROC Value** Surface Range Envelope  $0.14\pm0.01$  $0.19\pm0.02$  $0.60\pm0.01$ Classification Tree Analysis  $0.42\pm0.03$  $0.62\pm0.08$  $0.83\pm0.05$ Random Forest  $0.63\pm0.04$  $0.72\pm0.04$  $0.93\pm0.02$ Multiple Adaptive Regression Splines  $0.36\pm0.04$  $0.48\pm0.07$  $0.81\pm0.04$ Flexible Discriminant Analysis  $0.31\pm0.03$  $0.41\pm0.04$  $0.76\pm0.05$ Generalised Linear Model  $0.25\pm0.01$  $0.35\pm0.05$  $0.71\pm0.03$ Generalised Additive Model  $0.35\pm0.05\,$  $0.45\pm0.07$  $0.79\pm0.04$ 

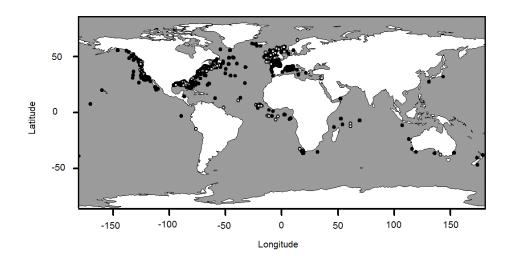
**Table 2.** Evaluation metrics Kappa, true skill statistic (TSS) and receiver operating characteristic (ROC) values for all species distribution models (mean value of five model runs  $\pm$  standard deviation). All models were performed in R, using package "Biomod2".

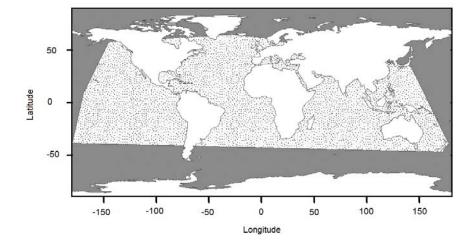
	Season			
	Jan-Mar	Apr-Jun	Jul-Sep	Oct-Dec
Mean predicted probability of <i>Mola</i> presence across all cells (± standard deviation)	$0.49\pm0.50$	$0.50\pm0.50$	$0.44\pm0.50$	$0.51 \pm 0.50$
Total number of cells modelled	41,009	41,369	41,579	41,369
No. cells predicted as suitable ( $p>0.7$ )	19,914	21,094	18,257	21,270
% of ocean surface predicted as suitable	48.56	50.99	43.91	51.42

**Table 3.** Analysis of random forest species distribution models to consider the extent of suitable habitat for *Mola* using the seasonal predicted probability of genus presence in each grid cell.

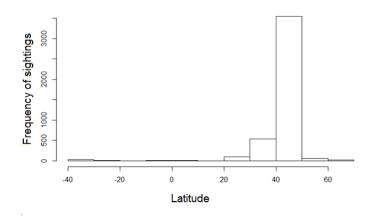
# Figures

**Figure 1a.** Global distribution of presence sightings of *Mola* (black) and pseudo-absences provided by sightings of leatherback turtles (grey) used in the species distribution model. **b.** Minimum convex hull range extent of *Mola* sightings data from 2000-2015. The origin of the base map is a spheroid WGS 1984 projection.

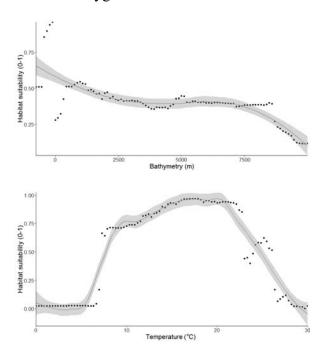


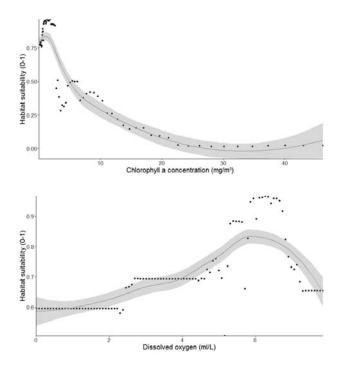


**Figure 2.** Frequency distribution of *Mola* sightings by latitude. The data showed a significant bias toward the Northern Hemisphere, particularly between 40-50° N. The origin of the base map is a spheroid WGS 1984 projection.

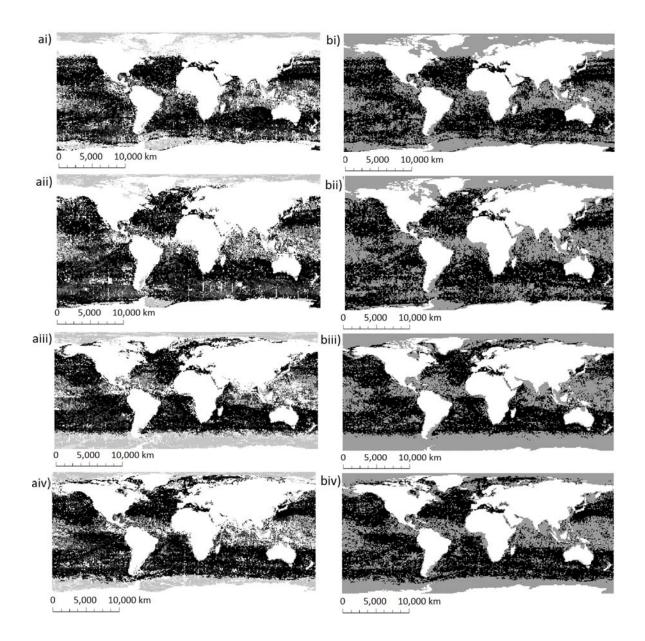


**Figure 3.** Species response curves for each climatic variable included in the final species distribution models (SDM) for *Mola* showing range suitability at the sea surface (0-1) against: a) bathymetry, b) sea surface temperature, c) chlorophyll concentration, and d) dissolved oxygen concentration.





**Figure 4a.** Seasonal range suitability at the sea surface for *Mola* from species distribution model (SDM) outputs for i) January to March, ii) April to June, iii) July to September and iv) October to December. Probability of presence shaded from 0 < 0.1 (white) to 0.9 < 1 (black) at 0.1 intervals. **b.** predicted presence of *Mola* (probability >0.7). Predicted presence (black) and predicted absence (grey). The origin of the base maps is a spheroid WGS 1984 projection.



**Figure 5a.** Seasonal predicted latitudinal range of *Mola* in both hemispheres (outliers removed). **b.** Intra-annual variation in latitude of sightings data plotted by day of year with locally weighted scatterplot smoothing curve (LOESS) curve and 95% confidence interval for both hemispheres.

