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15 **TITLE: DEPREDATION IN PELAGIC SURFACE LONGLINES IN THE**
16 **ATLANTIC AND INDIAN OCEANS**

17 **ABSTRACT**

18 Depredation has aroused great interest over the last few decades, mainly due to the
19 expansion of distant fishing, in particular longlines. For this study, captures and
20 depredation records were taken by scientific observers on board Portuguese commercial
21 longline vessels in the Atlantic and Indian Oceans, between 2011-2016. A total of 1336
22 fishing sets were monitored, with a total of 86,183 fish captures, including 1681
23 depredation events. The percentage of depredation tended to increase along the time
24 series, except in the last year where a decrease was noted. Significant differences
25 between sizes of swordfish (*Xiphias gladius*) damaged by predators were observed in
26 the Indian Ocean but not in the Atlantic. The highest proportions of depredation were
27 observed on tuna and small pelagic fishes in both oceans. For swordfish, the effects of
28 spatial variables were significant on the rate of depredation events. The results
29 presented in this study provide a first overview of the depredation patterns in the
30 Portuguese pelagic longline fishery in the Atlantic and Indian Oceans, which can inform
31 and improve fisheries management and contribute to the development of effective
32 mitigation measures to reduce the impacts of depredation on fisheries.

33

34 *KEYWORDS: Depredation, pelagic longline fisheries, swordfish, sharks, Indian Ocean,*
35 *Atlantic Ocean.*

36

37 **Introduction**

38 Depredation has aroused increasing interest over the last few decades due to the
39 expansion of distant fisheries, in particular pelagic longlines. Depredation is usually
40 defined as '*the partial or complete removal of hooked fish or bait from fishing gear... by*
41 *predators likes cetaceans, sharks, bony fish, birds, squids, crustaceans and others*'
42 distinguishing it from predation, i.e., '*the taking of free swimming fish (or other*
43 *organisms) ...*' (Donoghue, Reeves, & Stone, 2003; Gilman *et al.*, 2007; Romanov *et al.*,
44 2013). The partial or even complete removal of the catch and bite off of the gear can
45 lead to significant financial losses to the fisheries (Nishida & Shiba, 2005; Rabearisoa *et*
46 *al.*, 2012; Kumar *et al.*, 2016).

47 Depredation events have been documented to some extent in the Atlantic and
48 Indian Oceans. However, detailed information collected systematically is still rare for
49 both areas. Therefore, there is a need for the development of specific indicators to assess
50 the degree of depredation, which remains a poorly understood phenomenon, especially
51 in poorly studied areas of the Indian (Mutombene, 2015; Rabearisoa, Sabarros,
52 Romanov, & Bach, 2015; Varghese, Somvanshi, & Varghese, 2008) and Atlantic
53 Oceans (Hernandez-Milian *et al.*, 2008; MacNeil, Carlson, & Beerkircher, 2009;
54 Mandelman, Cooper, Werner, & Lagueux, 2008).

55 In this paper, the Portuguese pelagic longline fishery, a surface drifting longline
56 fishery targeting mainly swordfish (*Xiphias gladius* [SWO]) that operates over wide
57 regions of the Atlantic and Indian Oceans, was analyzed. Specific objectives of the
58 paper were to 1) analyze depredation events in relation to total captures, 2) evaluate
59 species-specific depredation events, 3) provide information on the main variables that
60 are related to the depredation events, and 4) discuss this case study within the context of
61 oceanic pelagic fisheries.

62 **Material and methods**

63 **Data collection**

64 Depredation records were taken by scientific observers on board Portuguese commercial
65 pelagic longline vessels that operate over wide areas of the Atlantic and Indian Oceans.
66 In the Atlantic, data were collected mainly in the Temperate, Tropical, Equatorial and
67 Subtropical waters of southern hemisphere (between 30°S to 43°N and 44°W to 7°E). In
68 the Indian Ocean data were collected mainly in the Subtropical waters (between 23°S to
69 34°S and 36°E to 96°E) (Fig. 1 - A). Data were compiled for the period from 2011 to
70 2016. A total of 1336 fishing operations, 787 in the Atlantic Ocean and 549 in the
71 Indian Ocean, were covered. In the Atlantic Ocean fleet, the fishing effort per set
72 averaged 1236 hooks and ranged from 668 to 2013 hooks. The fishing effort per set
73 averaged 1438 hooks and ranged from 505 to 2601 hooks for the Indian Ocean.

74 Data on specimen size (lower-jaw fork length [LJFL] for billfishes and fork
75 length [FL] for other bony fishes and sharks), location, depredation episodes and date
76 were recorded. Within the context of data reporting to the Regional Fisheries
77 Management Organizations, specifically ICCAT in the Atlantic and IOTC in the Indian
78 Ocean, the depredation events and rates are recommended to be reported in the Indian
79 Ocean but not in the Atlantic. As such, depredation episodes in the Atlantic Ocean were
80 recorded exclusively for individuals with high damages, i.e., those individuals with
81 large bites, tears or amputation of some parts of the body, such as the tail or belly area.
82 Predators were recorded only for depredated individuals in the Indian Ocean whenever
83 possible. To identify predator, observers analyzed the bite of the depredated individuals
84 and observed if predators were swimming near the vessel when the longline was being
85 hauled. In some cases, mainly with sharks and pelagic fish, a captured individual
86 contained the remains of other individuals previously depredated in the mouth of this

87 first, sometimes even with the hook inside. In the case of seabird depredation, they were
88 observer biting prey when the longline was being hauled.

89 Differences between sizes of swordfish damaged by predators were analyzed in
90 the Atlantic and Indian Oceans. Only individuals partially depredated, where size could
91 still be known, were taken into account in this analysis.

92 **Catch and depredation indicators**

93 The nominal CPUE (Catch Per Unit Effort), defined as the total number (N) of fish
94 caught (including both damaged or intact) per 1000 hooks was calculated for each
95 fishing set, and summarized by quarter and year for each region.

$$96 \quad CPUE = \frac{\text{Number of fish caught}}{\text{Number of hooks}} * 1000$$

97 Depredation Per Unit Effort (DPUE), defined as the number of fish depredated
98 per 1000 hooks, was calculated per set and assessed by quarter by using quarterly
99 pooled catch and fishing effort data, including non-depredated sets (e.g., Rabearisoa *et*
100 *al.*, 2015b; Ramos-Cartelle & Mejuto, 2008; Romanov *et al.*, 2013).

$$101 \quad DPUE = \frac{\text{Number of fish depredated}}{\text{Number of hooks set}} * 1000$$

102 The Interaction Rate (IR) was defined as the proportion of longline sets
103 depredated. IR was calculated using the entire dataset (operational set level data) of
104 longline operations. A fishing operation was considered depredated if at least one fish
105 (either a commercial or non-commercial species) was depredated on the longline (e.g.,
106 Nishida & Tanio, 2001; Rabearisoa *et al.*, 2015b; Romanov *et al.*, 2013).

$$107 \quad IR = \frac{\text{Number of depredated sets}}{\text{Total number of fishing operations}} * 100$$

108 The Gross Depredation Rate (GDR) was defined as the total number of fish
109 depredated divided by the total number of fish caught. Quarterly and yearly values of
110 GDR were calculated on the quarterly or yearly pooled catch, including non-depredated
111 sets (e.g., Donoghue *et al.*, 2003; Rabearisoa *et al.*, 2015b; Romanov *et al.*, 2013).

$$112 \qquad \qquad \qquad GDR = \frac{\textit{Number of fish depredated}}{\textit{Number of fish caught}}$$

113 **Data analysis**

114 Data from the Atlantic and Indian Oceans were compiled, analyzed and compared.
115 Catch data for each ocean was tested for normality with Shapiro-Wilk normality tests
116 (Shapiro & Wilk, 1965) and for homogeneity of variances with Levene tests (Levene,
117 1960). Due to violation of those parametric assumptions, univariate non-parametric
118 statistical tests (chi-squared) were used to compare total and depredated captures
119 between oceans.

120 The annual trends of total and depredated captures were plotted and analyzed, as
121 well as the proportions of depredated captures by species. The size distributions were
122 compared between depredated and non-depredated capture. This analysis was carried
123 out for swordfish, the main target species of the fleet.

124 A binomial Generalized Additive Model (GAM) with logit link function was
125 created to determine the effects of spatial variables (latitude and longitude) on the
126 depredation rates of swordfish in both oceans. The response variable was the swordfish
127 depredated/non-depredated captures, with each specimen coded as: 1=depredation event
128 occurred and 0=depredation event did not occur. The model also accounted for the year
129 effect, as a fixed categorical factor. Other variables, such as SST were also tested in the
130 model, but were not used due to collinearity with the spatial effects, particularly with

131 latitude. The predicted swordfish depredation occurrences (binomial proportions) from
132 this final GAM model were plotted along the study areas in each ocean.

133 The analysis for this paper was carried out using the R language for statistical
134 computing version 3.3.2. (R Core Team, 2016). Additional packages that were used
135 included “car” (Fox & Weisberg, 2011), “cowplot” (Wilke, 2015), “descr” (Aquino,
136 Enzmann, Schwartz, Jain, & Kraft, 2016), “ggmap” (Kahle & Wickham, 2013),
137 “ggplot2” (Wickham, 2009), “gridExtra” (Auguie, 2016), “gtable”(Wickham, 2016a),
138 “lattice” (Sarkar, 2008), “maps” (Brownrigg, Minka, Becker, & Wilks, 2010),
139 “mapdata” (Becker, Wilks, & Brownrigg, 2016), “mgcv” (Wood, 2011), “nortest”
140 (Gross & Ligges, 2015), “perm” (Fay & Shaw, 2010), “raster” (Hijmans, 2016),
141 “RColorBrewer” (Neuwirth, 2014), ”reshape2” (Wickham, 2007), “RgoogleMaps”
142 (Loecher & Ropkins, 2015),“Rmisc” (Hope, 2013), “scales” (Wickham, 2016b).

143 **Results**

144 **Spatial distribution of catches and depredation events**

145 In the Atlantic Ocean, a total of 55,482 captures were recorded and considered within
146 the scope of this study. The sample covered a wide geographical area of the Atlantic
147 Ocean, with most sets taking place in the tropical and equatorial regions, but also in the
148 temperate north and south (Fig. 1 - B). The total number of individuals depredated were
149 778, representing about 1.4% of the total catch. These depredations events occurred in
150 54% of the total sets, concretely in 421 of the 784 sets during the study period and area
151 (Fig. 1 - C).

152 In the case of the Indian Ocean, a total of 30,701 captures were recorded during
153 the study. The sample covered a large geographical area of the south Indian Ocean, with
154 most sets taking place in the SW region (Fig. 1 - B). The individuals depredated

155 represented about 2.9% of the total capture, with a total number of 903 individuals
156 depredated. These depredations events occurred in 395 of the 548 sets during the study
157 period and area, representing depredation occurrences in 72% of the total sets (Fig. 1 -
158 C).

159 Total and depredated captures per set data were not normally distributed
160 (Shapiro-Wilk test: (Atlantic Ocean) $W = 0.834$, $P < 0.001$ and $W = 0.714$, $P < 0.001$.
161 (Indian Ocean) $W = 0.985$, $P < 0.001$ and $W = 0.841$, $P < 0.001$). Variances of total
162 captures were heterogeneous between oceans (Levene test: $F = 34.814$, $df = 1$, $P <$
163 0.001) but homogeneous for the depredated captures ($F = 0.374$, $df = 1$, $P = 0.541$).
164 Using univariate non-parametric statistical tests revealed that total and depredated
165 captures per set were significantly different between oceans (Permutation test: chi-
166 squared = 82.051, $df = 1$, $P < 0.001$ and chi-squared = 5.531, $df = 1$, $P = 0.019$)
167 respectively.

168 **Depredation indicators**

169 Quarterly CPUE was variable, dependent on capture distributions and presence/absence
170 of predator attacks. The overall CPUE varied from 47 to 78.6 and 28.1 to 50.4
171 specimens/1000 hooks respectively for the Atlantic and Indian Oceans (Table 1). High
172 values of CPUE in some year-quarters combinations for the Atlantic Ocean (2014-
173 4=106.6; 2016-1=157.5; 2016-3=243.2) were mainly due to fishing taking place in areas
174 where blue shark individuals were very abundant. Annual DPUE values varied from 0.5
175 to 1.2 specimens/1000 hooks in the Atlantic Ocean, and slightly lower for the Indian
176 Ocean, specifically varying from 0.7 to 1.8 specimens/1000 hooks (Table 1).

177 A total of 801 fishing operations were depredated, specifically 406 in the
178 Atlantic Ocean (IR=52.7%) and 395 in the Indian Ocean (IR=71.8%) (Table 1). The

179 main depredation hotspots were located in the tropical and equatorial Atlantic Ocean,
180 and in the southwest and central-south Indian Ocean (Fig. 1 - C). The yearly values of
181 the gross depredation rate varied between 1.1% and 1.9% in the Atlantic Ocean and
182 between 1.5% and 4.9% in the Indian Ocean (Table 1).

183 **Annual trends of depredated catches**

184 Total and depredated captures variances were homogeneous between years in the
185 Atlantic (Levene test: $F = 0.659$, $df = 5$, $P = 0.655$ and $F = 0.92$, $df = 5$, $P = 0.467$) and
186 heterogeneous in the Indian Ocean (Levene test: $F = 3.586$, $df = 5$, $P = 0.003$ and $F =$
187 3.705 , $df = 5$, $P = 0.003$). Using univariate non-parametric statistical tests revealed that
188 total and depredated captures were significantly different between years for both oceans
189 (Permutation test: (Atlantic Ocean) chi-squared = 28.275, $df = 5$, $P < 0.001$ and chi-
190 squared = 29.249, $df = 5$, $P < 0.001$. (Indian Ocean) chi-squared = 150.73, $df = 5$, $P <$
191 0.001 and chi-squared = 60.839, $df = 5$, $P < 0.001$).

192 In the Atlantic Ocean, the fraction of depredated captures had an increasing
193 annual trend of 0.26 % per year on average, ranging from 1.1 % in 2011 and reaching
194 2.1 % in 2015. However, depredation captures decreased by 1 % in the last year,
195 specifically to 1.1 % in 2016 (Table 2). The fraction of depredated captures also
196 increased similarly in the Indian Ocean, with an increasing annual trend of 0.85 % on
197 average, being 1.5 % in 2011 and reaching 4.9 % in 2015. Depredated captures also
198 decreased in the last year (2016), in this case by 2 % and reaching 2.9 % (Table 2).

199 **Depredated species**

200 *Atlantic Ocean*

201 In the Atlantic Ocean, the Portuguese longline fishery catch composition is mostly
202 composed of 6 species, blue shark (*Prionace glauca* [BSH]), swordfish, bigeye tuna

203 (*Thunnus obesus* [BET]), common dolphinfish (*Coryphaena hippurus* [DOL]),
204 crocodile shark (*Pseudocarcharias kamoharai* [PSK]) and shortfin mako (*Isurus*
205 *oxyrinchus* [SMA]) (Table 3). These species represent 86.7% of the fish catch in
206 numbers, with particular highlights to blue shark and swordfish with 54.4 and 22.2% of
207 the total catches, respectively (Table 3).

208 A total of 24 species were depredated, with swordfish representing 49% of the
209 depredated captures, followed by bigeye tuna, blue shark, common dolphinfish, wahoo
210 (*Acanthocybium solandri* [WAH]), escolar (*Lepidocybium flavobrunneum* [LEC]),
211 atlantic sailfish (*Istiophorus albicans* [SAI]), atlantic white marlin (*Tetrapturus albidus*
212 [WHM]) and yellowfin tuna (*Thunnus albacares* [YFT]) (Table 3).

213 The percentage of depredated individuals against total catch by species is
214 represented in Table 4. Tuna and small pelagic fishes had the highest relative
215 percentages of depredation in relation to their total catches. Atlantic pomfret (*Brama*
216 *brama* [POA]), driftfish (*Cubiceps* spp. [CUP]), striped marlin (*Tetrapturus audax*
217 [MLS]) and large tunas (*Thunnus* spp. [TUS]) stand out as the most depredated
218 species/taxa, with a range of 14-34% of the individuals captured having been
219 depredated (Table 4).

220 *Indian Ocean*

221 In the Indian Ocean Portuguese longline fishery the catch composition is mostly
222 composed of 6 species, swordfish, blue shark, common dolphinfish, escolar, shortfin
223 mako and bigeye tuna (Table 3). These species represent 89% of the fish catch in
224 numbers, highlighting again to swordfish and blue shark captures, with 36.4 and 27.9%
225 of total catches, respectively (Table 3).

226 A total of 24 species were depredated, with swordfish representing 55.6% of the
227 depredated captures, followed by escolar, common dolphinfish, blue shark, bigeye tuna,
228 wahoo, albacore (*Thunnus alalunga* [ALB]), shortfin mako and long snouted lancetfish
229 (*Alepisaurus ferox* [ALX]) (Table 3).

230 The percentage of depredated individuals against total catch by species is
231 represented in Table 4. Similarly to the Atlantic, tuna and small pelagic fishes also had
232 the highest percentages of depredation in relation to their total catches. In this case,
233 Wahoo and snake mackerel (*Gempylus serpens* [GES]) stand out as the most depredated
234 species, ranging between 9-12% of individuals depredated.

235 Predators were recorded only for depredated individuals in the Indian Ocean. It
236 was not possible to identify the predator in 61% of depredated individuals. For the ones
237 that the predator could be identified, 21% of the depredation was from sharks species,
238 including blue shark, shortfin mako, porbeagle (*Lamna nasus* [POR]) and the small
239 cookie cutter shark (*Isistius brasiliensis* [ISB]). Small pelagic fish preyed on about 13%
240 of the depredated individuals. Marine mammals and seabirds were responsible for 1.9
241 and 0.3% respectively of the depredation events. Only 2.4% of the depredated captures
242 were targeted by more than one predator.

243 **Size distribution of depredated and total catch of swordfish**

244 As the main target species in the fishery and the one with more depredation events, a
245 specific size composition analysis was carried out for swordfish. There were no
246 differences in the sizes of swordfish between depredated and total catches for the
247 Atlantic Ocean (Proportion test: chi-squared = 7.0798, df = 17, $P = 0.9825$), but there
248 were differences for the Indian Ocean (Proportion test: chi-squared = 43.169, df = 17, P

249 = 0.0005) (Fig. 2). It was not possible to compare the sizes of other species due to the
250 limited number of damaged individuals.

251 **Modelling depredation rates on swordfish**

252 The effects of continuous spatial variables (latitude and longitude) were significant on
253 the rate of depredation events in swordfish specimens in the Atlantic and Indian Oceans
254 (Fig. 3). It is possible to see that in general there were major depredation rates towards
255 western longitudes in the Atlantic Ocean. In terms of latitude, the higher depredation
256 rates are in the tropical zone of the operational areas of the Portuguese fleet. The map of
257 depredation rates spatial predictions showed that spatial depredation rates were closely
258 related to latitude, with two distinct areas of high-depredation rates, one close to the
259 west coast of Africa around 10°N, and the other located in the southeastern Atlantic
260 Ocean around 15°S (Fig. 4). In the Indian Ocean, there were higher depredation rates
261 mainly towards eastern longitudes, even though there was also a peak in the middle of
262 the western areas, closer to the African continent (Fig. 3). Regarding latitude, the higher
263 rates are in the extremes (higher and lower latitudes) of the areas of operation of the
264 Portuguese fleet. The plot with the predictions of the depredation rates along the study
265 area of the Indian Ocean showed an area of moderate depredation rates probability in
266 the eastern of Indian Ocean around 30°S 90°E (Fig. 5).

267 **Discussion**

268 This work provides the first study of depredation in the Portuguese pelagic longline
269 fleet that targets mainly swordfish in the Atlantic and Indian Oceans, compiled by
270 fisheries observers on board commercial longline vessels. The Portuguese pelagic
271 longline fleet is affected by occurrences of depredation events on the catches, with
272 impacts to the fishery, similar to many other fleets around the world (Gilman *et al.*,

273 2007). This study also reports the extent and spatial distribution of the depredation
274 occurrences and the main species that are impacted in both oceans.

275 Several depredation mitigation measures have been or are being tested
276 worldwide to mitigate this issue, including physical protection of the catch or acoustic
277 devices, but this remains a challenging work (Tixier *et al.*, 2010; Løkkeborg, 2011;
278 Hamer *et al.*, 2012; O'Connell *et al.*, 2015; Rabearisoa *et al.*, 2015; Straley *et al.*, 2015;
279 Tixier *et al.*, 2014; Werner *et al.*, 2015). For this reason, it is important to know the
280 mechanisms by which depredation episodes occur in pelagic longline fleets.

281 Very few previous studies have discussed the effects of depredation in pelagic
282 longline fleets. The total values of depredation captures of Portuguese pelagic longline
283 fleet described in this work for the Atlantic Ocean are similar to those obtained by
284 Mandelman *et al.* (2008), that indicated that the damage inflicted in the catch by
285 depredation between 1990 and 1997 in the U.S. Atlantic pelagic longline fishery was
286 4% of total observed catch. Of those, 68% occurred on captures of swordfish, yellowfin
287 and bigeye tuna collectively, similar to the results obtained in our study. However, this
288 work also report events on other species such as escolar, dolphinfish or blue shark. Our
289 results are also similar with those of Hernandez-Milian *et al.* (2008), whose reports of
290 occurrences of depredation in the Atlantic Ocean were of less than 1% of the total catch.
291 As in our study, Rabearisoa *et al.* (2015b) also did not find annual values of DPUE
292 exceeding 2 specimens/1000 hooks. By the contrary, our results of IR are higher
293 compared with other studies, such as Hernandez-Milian *et al.* (2008) in the Atlantic and
294 Rabearisoa *et al.* (2015b) in the Indian Ocean.

295 On the other hand, a higher proportion of depredated captures was observed for
296 some species in both oceans, like tunas and small pelagic fishes, possibly showing that
297 there is a depredation preference for some species by the predators. These same results

298 were obtained in the study of shark depredation in pelagic longline fishery in the
299 Northwest Atlantic (MacNeil *et al.*, 2009) and in the pelagic longline fleet of Reunion
300 Island in the Indian Ocean (Rabearisoa, Sabarros, *et al.*, 2015) where tunas showed the
301 highest ratios of depredation. Depth, as well turbidity can determine the catchability of
302 these species, as depredation cases seem to have less success in areas with poor
303 visibility (Ward & Myers, 2005).

304 Significant differences in the size of fish damaged by predators were observed
305 for swordfish in our study for the Indian Ocean, but not for the Atlantic. Some studies
306 of fishes' prey preference show selection for certain sizes (e.g., Hart, 1986; Løkkeborg
307 & Bjordal, 1995). The results of Barnes *et al.* (2010) suggest that very general rules
308 determine dominant trends in predator–prey mass ratios in diverse marine ecosystems,
309 leading to the ubiquity of size-based trophic structuring and the consistency of observed
310 relationships between the relative abundance of individuals and their body size.
311 However, in our work caution should be taken when interpreting these results as the
312 depredated captures sample size is relatively small.

313 Very few previous studies have discussed the effects of spatial variables on the
314 rate of depredation events. For swordfish, the effects of the spatial variables (latitude
315 and longitude) were significant on the rate of depredation events both in the Atlantic
316 and Indian Oceans. It is possible that variables such as temperature or depth, related to
317 spatial variables, are related to the distribution of the catches of oceanic migratory
318 species (Hernandez-Milian *et al.*, 2008), which in turn can lead to more depredation
319 situations. GAM models as applied in our study predict the probability of having or not
320 having predation events, but the most likely responsible species in each region was not
321 explored because of the limits in the analysis and specifically in the taxa-specific data
322 availability. Future research recommendations should therefore include exploring with

323 different models for the various predator taxa (sharks, mammals, etc.). The reason
324 being that different depredation rates observed in specific areas might not necessarily
325 be directly related with the economic losses if the predator specific-depredation levels
326 and damage are different (e.g., cetacean or large sharks depredation that creates severe
327 damage *versus* cookiecutter or birds depredation that typically produces much less
328 damage to the catches). As such, the analysis as presented is valid for the purpose of
329 comparing wide oceanic areas where depredation events are more likely to occur, even
330 though it is limited in terms of the most likely predators and consequently the most
331 likely levels of damage and losses.

332 With respect to the spatial distribution of the data, while the observations
333 reported in part reflect the spatial dynamics of catches, there is also a large influence of
334 the seasonal and spatial patterns of the fishing effort of the fleet. In a study on the
335 salmon troll fishery, Abrahams & Healey (1990) reported that vessels from the fleet
336 differ substantially in their competitive capacity, and also, these differences add to a
337 considerable temporal and spatial variation in catch rates. That is why these aspects
338 have to be taken into account when discussing the analysis of total catches and the ones
339 that are preyed upon.

340 Depredation events have potential implications for fisheries management and
341 should be taken into account in the stocks assessment of highly migratory species. This
342 new information about depredation events can help future specific studies by taxa or in
343 more specific zones, delimiting different areas according to the predation rates, and
344 providing the fishing industry with relevant information about the fishing areas. In this
345 sense, full bio-socio-economic assessments of the costs and benefits of changing fishing
346 practices are needed. For example, the “move-on” technique may involve increased
347 non-fishing time and motor-fuel consumption that can render this fishing strategy less

348 advantageous to fishers or sustainable to the fishery itself (Janc *et al.*, 2018). Besides,
349 new technical means and mitigation measures that can reduce depredation on pelagic
350 longline fishery should be developed, thus reducing the number of discarded dead
351 catches caused by these events.

352 Depredation is an inevitable part of conducting longline operations in the open
353 ocean (MacNeil *et al.*, 2009), and can cause significant economic losses to the fishing
354 industry and ecological for the marine environment, especially when the captures are
355 discarded (Gilman *et al.*, 2008). For this reason it is crucial to monitor this phenomenon
356 more closely and periodically, even though it has been poorly studied so far.

357 Depredation monitoring should involve both scientists and the fishing sector, and
358 include the collection of standardized data (Romanov, Bach, & Rabearisoa, 2009). In
359 the future such studies should be continued as more data is being continuously collected
360 on the onboard observer programs. Improving knowledge of depredation will provide
361 valuable information for the development of effective mitigation measures, reducing the
362 impacts of depredation on fisheries.

363

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519 **Figures**

520 Figure 1. (A): Fishing operations of the Portuguese pelagic longline fleet in the Atlantic
521 and Indian Oceans between 2011-2016 where depredation event data were recorded and
522 analyzed. (B): Distribution in 5*5 degrees of total catches recorded for this study in the
523 Atlantic and Indian Oceans (2011-2016). (C): Distribution in 5*5 degrees of depredated
524 catches recorded for this study in the Atlantic and Indian Oceans (2011-2016).

525 Figure 2. Size-frequency distributions of total (n=11,967 and n=10,929) and depredated
526 (n=88 and n=224) catches of swordfish in the Atlantic (A-B) and Indian (C-D) Oceans
527 for the Portuguese pelagic longline fishery. Sizes are grouped in 10-cm lower-jaw fork
528 length (LJFL) classes.

529 Figure 3. Generalized Additive Model (GAM) plots with the non-linear effects of
530 latitude and longitude in the depredation events on swordfish specimens, in the pelagic
531 longline fishery operating in the Atlantic (A) and Indian (B) Oceans.

532 Figure 4. Prediction of the depredation rates on swordfish (binomial response) from a
533 Generalized Additive Model (GAM), for the Atlantic Ocean study region.

534 Figure 5. Prediction of the depredation rates on swordfish (binomial response) from a
535 Generalized Additive Model (GAM), for the Indian Ocean study region.

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541 **Tables**

542 Table 1. Catch and depredation indicators per year quarter combination in the Atlantic and Indian Oceans. Nsets is the number of sets; CPUE is
 543 the catch per unit of effort (number of fish caught per 1000 hooks); DPUE is the depredation per unit of effort (number of fish depredated per
 544 1000 hooks); IR is the interaction rate (proportion of depredated sets); GDR is the gross depredation rate (percentage of fish depredated within
 545 the entire catch).

Quarters	Atlantic Ocean					Indian Ocean				
	Nsets	CPUE	DPUE	IR	GDR	Nsets	CPUE	DPUE	IR	GDR
2011-1	59	40.0	0.4	33.9	1.0	-	-	-	-	-
2011-2	-	-	-	-	-	33	56.7	0.7	57.6	1.3
2011-3	23	64.7	0.8	60.9	1.2	70	40.7	0.7	62.9	1.6
2011-4	124	54.1	0.6	37.1	1.1	-	-	-	-	-
Annual-2011	206	51.5	0.6	38.8	1.1	103	45.5	0.7	61.2	1.5
2012-1	37	56.4	0.6	45.9	1.0	-	-	-	-	-
2012-2	54	85.8	1.4	66.7	1.6	-	-	-	-	-
2012-3	70	54.7	0.6	50.0	1.2	43	48.1	1.2	69.8	2.5
2012-4	49	60.6	0.8	55.1	1.4	13	57.8	0.9	69.2	1.6
Annual-2012	210	63.5	0.8	54.8	1.3	56	50.4	1.1	69.6	2.3
2013-1	-	-	-	-	-	20	59.8	2.3	90.0	3.8
2013-2	5	40.0	0.2	20.0	0.4	67	46.1	1.1	74.6	2.5
2013-3	67	47.2	0.6	43.3	1.2	-	-	-	-	-
2013-4	1	68.0	1.7	100.0	2.4	43	33.1	0.9	69.8	2.7
Annual-2013	73	47.0	0.5	42.5	1.2	130	43.7	1.2	75.4	2.8

2014-1	13	44.5	1.0	61.5	2.3	49	28.3	1.1	75.5	4.0
2014-2	16	44.6	0.8	56.3	1.7	-	-	-	-	-
2014-3	69	52.3	1.1	63.8	2.1	-	-	-	-	-
2014-4	7	106.6	0.9	71.4	0.9	-	-	-	-	-
Annual-2014	105	53.4	1.0	62.9	1.9	49	28.3	1.1	75.5	4.0
2015-1	51	52.8	1.2	66.7	2.2	-	-	-	-	-
2015-2	41	88.9	1.2	58.5	1.3	67	31.4	1.8	88.1	5.8
2015-3	10	71.0	1.4	80.0	2.0	41	47.2	1.7	92.7	3.7
2015-4	-	-	-	-	-	-	-	-	-	-
Annual-2015	102	69.4	1.2	64.7	1.7	108	37.0	1.8	89.8	4.9
2016-1	13	157.5	0.7	30.8	0.4	-	-	-	-	-
2016-2	40	55.9	0.8	55.0	1.5	38	28.6	1.5	84.2	5.2
2016-3	14	243.2	3.0	57.1	1.2	65	27.9	0.4	44.6	1.5
2016-4	24	57.0	0.6	58.3	1.1	-	-	-	-	-
Annual-2016	91	78.6	0.9	52.7	1.1	103	28.1	0.8	59.2	2.9

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551 Table 2. Annual catches, in %, for depredated and non-depredated individuals, during
 552 the period 2011-2016, in the Atlantic and Indian Oceans. *n* is the number of individuals.

Year	Atlantic Ocean		Indian Ocean	
	Not Depredated % (<i>n</i>)	Depredated % (<i>n</i>)	Not Depredated % (<i>n</i>)	Depredated % (<i>n</i>)
2011	98,9% (16622)	1,1% (184)	98,5% (6283)	1,5% (95)
2012	98,6% (13686)	1,4% (189)	97,7% (3629)	2,3% (84)
2013	98,8% (4025)	1,2% (47)	97,1% (7860)	2,9% (238)
2014	98,1% (6796)	1,9% (135)	96,3% (2110)	3,7% (80)
2015	97,9% (6857)	2,1% (149)	95,1% (5506)	4,9% (281)
2016	98,9% (6718)	1,1% (74)	97,1% (4210)	2,9% (125)

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567 Table 3. Percentage of total catches of the main target species recorded for this study in
568 the Atlantic and Indian Oceans (2011-2016) (n=55,482 and n=30,701, respectively) and
569 contributions of species to depredated catches of longline fishery in percentage (n=778
570 and n=903 respectively). Bigeye thresher (*Alopias superciliosus* [BTH]), pelagic
571 stingray (*Dasyatis violacea* [PLS]), blue marlin (*Makaira nigricans* [BUM]), oceanic
572 whitetip shark (*Carcharhinus longimanus* [OCS]), longfin mako (*Isurus paucus*
573 [LMA]), indo-Pacific sailfish (*Istiophorus platypterus* [SFA]), olive ridley turtle
574 (*Lepidochelys olivacea* [LKV]), toli shad (*Tenualosa toli* [TOL]), smooth hammerhead
575 (*Sphyrna zygaena* [SPZ]), leatherback turtle (*Dermochelys coriacea* [DKK]), silky
576 shark (*Carcharhinus falciformis* [FAL]), devil fish (*Mobula mobular* [RMM]), greater
577 amberjack (*Seriola dumerili* [AMB]), shortbill spearfish (*Tetrapturus angustirostris*
578 [SSP]), longbill spearfish (*Tetrapturus pfluegeri* [SPF]), southern bluefin tuna (*Thunnus*
579 *maccoyii* [SBF]), opah (*Lampris guttatus* [LAG]), barracuda (*Sphyrna* spp. [BAR]),
580 oilfish (*Ruvettus pretiosus* [OIL]).

Atlantic Ocean						Indian Ocean					
Species	Observed	in %	Species	Depredations	in %	Species	Observed	in %	Species	Depredations	in %
BSH	30167	54,37	SWO	382	49,10	SWO	11168	36,38	SWO	502	55,59
SWO	12337	22,24	BET	81	10,41	BSH	8568	27,91	LEC	110	12,18
BET	1934	3,49	BSH	46	5,91	DOL	2787	9,08	DOL	58	6,42
DOL	1481	2,67	DOL	43	5,53	LEC	2185	7,12	BSH	48	5,32
PSK	1250	2,25	WAH	35	4,50	SMA	1518	4,94	BET	41	4,54
SMA	943	1,70	LEC	35	4,50	BET	1082	3,52	WAH	31	3,43
YFT	868	1,56	SAI	34	4,37	ALB	368	1,20	ALB	21	2,33
BTH	626	1,13	WHM	27	3,47	ALX	340	1,11	SMA	21	2,33
PLS	581	1,05	YFT	24	3,08	PLS	326	1,06	ALX	20	2,21
SAI	574	1,03	BTH	15	1,93	WAH	265	0,86	GES	17	1,88
WHM	520	0,94	BUM	13	1,67	FAL	222	0,72	MLS	6	0,66
LEC	448	0,81	ALB	9	1,16	OIL	212	0,69	POA	4	0,44
BUM	424	0,76	PSK	8	1,03	GES	178	0,58	SSP	4	0,44
OCS	414	0,75	SMA	7	0,90	YFT	175	0,57	YFT	4	0,44
LMA	360	0,65	ALX	6	0,77	POR	161	0,52	FAL	4	0,44
WAH	353	0,64	AMB	3	0,39	SFA	142	0,46	OIL	3	0,33
LKV	271	0,49	TUS	2	0,26	MLS	121	0,39	SFA	2	0,22
ALX	230	0,41	SPF	2	0,26	SSP	107	0,35	BAR	1	0,11

TOL	205	0,37	POA	1	0,13	SBF	86	0,28	LAG	1	0,11
SPZ	193	0,35	CUP	1	0,13	POA	80	0,26	SPZ	1	0,11
DKK	143	0,26	MLS	1	0,13	BTH	78	0,25	BUM	1	0,11
FAL	135	0,24	OIL	1	0,13	RMM	63	0,21	BTH	1	0,11
RMM	108	0,19	GES	1	0,13	BUM	60	0,20	SBF	1	0,11
ALB	103	0,19	LMA	1	0,13	PSK	41	0,13	POR	1	0,11

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597 Table 4. Percentage of depredated catches compared to the species-specific total catches
 598 for the Portuguese pelagic longline fishery in the Atlantic and Indian Oceans.

Atlantic Ocean				Indian Ocean			
Species	Observed	Depredations	Proportion	Species	Observed	Depredations	Proportion
<i>Brama brama</i>	3	1	33,33	<i>Acanthocybium solandri</i>	265	31	11,70
<i>Cubiceps</i> spp.	4	1	25,00	<i>Gempylus serpens</i>	178	17	9,55
<i>Tetrapturus audax</i>	14	2	14,29	<i>Sphyrna</i> spp.	15	1	6,67
<i>Thunnus</i> spp.	7	1	14,29	<i>Alepisaurus ferox</i>	340	20	5,88
<i>Acanthocybium solandri</i>	353	35	9,92	<i>Thunnus alalunga</i>	368	21	5,71
<i>Thunnus alalunga</i>	103	9	8,74	<i>Lepidocybium flavobrunneum</i>	2185	110	5,03
<i>Lepidocybium flavobrunneum</i>	448	35	7,81	<i>Brama brama</i>	80	4	5,00
<i>Seriola dumerili</i>	43	3	6,98	<i>Tetrapturus audax</i>	121	6	4,96
<i>Istiophorus albicans</i>	574	34	5,92	<i>Lampris guttatus</i>	21	1	4,76
<i>Tetrapturus albidus</i>	520	27	5,19	<i>Xiphias gladius</i>	11168	502	4,49
<i>Thunnus obesus</i>	1934	81	4,19	<i>Thunnus obesus</i>	1082	41	3,79
<i>Tetrapturus pfluegeri</i>	56	2	3,57	<i>Tetrapturus angustirostris</i>	107	4	3,74
<i>Xiphias gladius</i>	12337	382	3,10	<i>Sphyrna zygaena</i>	40	1	2,50
<i>Makaira nigricans</i>	424	13	3,07	<i>Thunnus albacares</i>	175	4	2,29
<i>Coryphaena hippurus</i>	1481	43	2,90	<i>Coryphaena hippurus</i>	2787	58	2,08
<i>Ruvettus pretiosus</i>	35	1	2,86	<i>Carcharhinus falciformis</i>	222	4	1,80
<i>Thunnus albacares</i>	868	24	2,76	<i>Makaira nigricans</i>	60	1	1,67
<i>Alepisaurus ferox</i>	230	6	2,61	<i>Ruvettus pretiosus</i>	212	3	1,42
<i>Alopias superciliosus</i>	626	15	2,40	<i>Istiophorus platypterus</i>	142	2	1,41
<i>Gempylus serpens</i>	60	1	1,67	<i>Isurus oxyrinchus</i>	1518	21	1,38
<i>Isurus oxyrinchus</i>	943	7	0,74	<i>Alopias superciliosus</i>	78	1	1,28
<i>Pseudocarcharias kamoharai</i>	1250	8	0,64	<i>Thunnus maccoyii</i>	86	1	1,16
<i>Isurus paucus</i>	360	1	0,28	<i>Lamna nasus</i>	161	1	0,62
<i>Prionace glauca</i>	30167	46	0,15	<i>Prionace glauca</i>	8568	48	0,56