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Abstract

Quantification of discard per unit effort rates (DPUE) has been proposed by the European Commission as a measure to manage the discarding of commercially fished organisms. In the Spanish fresh trawling fleet operating in North West Africa, both target species of black hakes, *Merluccius polli* and *Merluccius senegalensis* are data limited stocks (DLS). Hence, discards of these fleets are even more unknown but not unimportant part of the total catch (retained and discarded). Onboard observer data from commercial surveys from 2016 to 2018 provide a detailed source of scientific information about catches, discards, effort and technical factors in this fleet. This is the first quantitative analysis to model DPUE through generalised linear mixed models (GLMM), based on the explicit distinction between abundance and technical factors coming from information of observer surveys. We describe the relationship between discards and environment, catches of target and other species, effort of the fleet, spatial and temporal variation in discard accessibility, vessel characteristics, strategy of the skippers and market decisions. Unlike hake catches, discards were higher and more dispersed in shallower than in deeper waters. We identified two separate métiers for the Spanish fresh trawling fleet determined by depth and treated total discards as a stock unit susceptible of being monitored, managed and assessed. The strategy of the skipper appears to have a more important effect on discards than vessel characteristics. This study shows the importance of observer data for this fishery and identifies recommendations for the improvement in the scientific usefulness of logbook information.

Keywords	discard per unit of effort (DPUE); data limited stocks (DLS); generalised linear mixed models (GLMM); logbooks; observers on board; trawl
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Dear editor,

EU fisheries in NW Africa are often characterised by lack of precise data on catches and discards, affecting to economic, social and management topics. Multi-specific catch and complex dynamics of demersal trawl fleet in the area can only be deeply analysed from quality on-board scientific data, particularly those concerning discards. Our study has been supported by a strong cooperation team, integrated by researchers, scientific observer, skippers and ship-owner that allowed us to identify existing gaps of this fishery, improve the knowledge about the effect affecting discards and propose recommendations for management of discards in this fishery. Thus, the following results are completely new and promising:

1. Analysis of the most detailed scientific information of the most important demersal fleet in the North-West Africa based on board observer data from 2016 to 2018.
2. Quantitative analysis of the main effects affecting discards taking into account abundance and technical variables.
3. Description of the relationship between discards and the environment, catches of target and other species, effort of the fleet, spatial and temporal variation in discard accessibility, vessel characteristics, strategy of the skippers and market decisions.
4. Index of discards rate (DPUE).
5. Replicable methodology based on GLMM fitted to DPUE data for the Spanish fresh trawling fleet in North-West African waters.
6. The treatment of total discards as a stock unit susceptible of being monitored, managed and assessed.
7. Emphasize the importance of observer data for this fishery and provide recommendations to improve the logbook information so that data can be scientifically exploited better.

Our article, in accordance to MP, is related with management and conservation issues related with the marine environment. Above all, our results are a necessary starting point to further studies heading for improving management in living resources in NW Africa: discard biodiversity, impact on marine habitat, estimating and reducing discards, etc.

Quantification of discard per unit effort rates (DPUE) has been proposed by the European Commission as a measure to manage the discarding of commercially fished organisms. In the Spanish fresh trawling fleet operating in North West Africa, both target species of black hakes, *Merluccius polli* and *Merluccius senegalensis* are data limited stocks (DLS). Hence, discards of these fleets are even more unknown but not unimportant part of the total catch (retained and discarded). Onboard observer data from commercial surveys from 2016 to 2018 provide a detailed source of scientific information about catches, discards, effort and technical factors in this fleet. This is the first quantitative analysis to model DPUE through generalised linear mixed models (GLMM), based on the explicit distinction between abundance and technical factors coming from information of observer surveys. We describe the relationship between discards and environment, catches of target and other species, effort of the fleet, spatial and temporal variation in discard accessibility, vessel characteristics, strategy of the skippers and market decisions. Unlike hake catches, discards were higher and more dispersed in shallower than in deeper waters. We identified two separate métiers for the Spanish fresh trawling fleet determined by depth and treated total discards as a stock unit susceptible of being monitored, managed and assessed. The strategy of the skipper appears to have a more important effect on discards than vessel characteristics. This study shows the importance of observer data for this fishery and identifies recommendations for the improvement in the scientific usefulness of logbook information.

Towards discard quantification of Data Limited Stocks based in on-board observers data: the case of Spanish fresh trawlers targeting black hake in NW Africa.

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Quantification of discard per unit effort rates (DPUE) has been proposed by the European Commission as a measure to manage the discarding of commercially fished organisms. In the Spanish fresh trawling fleet operating in North West Africa, both target species of black hakes, *Merluccius polli* and *Merluccius senegalensis* are data limited stocks (DLS). Hence, discards of these fleets are even more unknown but not unimportant part of the total catch (retained and discarded). Onboard observer data from commercial surveys from 2016 to 2018 provide a detailed source of scientific information about catches, discards, effort and technical factors in this fleet. This is the first quantitative analysis to model DPUE through generalised linear mixed models (GLMM), based on the explicit distinction between abundance and technical factors coming from information of observer surveys. We describe the relationship between discards and environment, catches of target and other species, effort of the fleet, spatial and temporal variation in discard accessibility, vessel characteristics, strategy of the skippers and market decisions. Unlike hake catches, discards were higher and more dispersed in shallower than in deeper waters. We identified two separate métiers for the Spanish fresh trawling fleet determined by depth and treated total discards as a stock unit susceptible of being monitored, managed and assessed. The strategy of the skipper appears to have a more important effect on discards than vessel characteristics. This study shows the importance of observer data for this fishery and identifies recommendations for the improvement in the scientific usefulness of logbook information.

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1. Introduction

Studies of discards in EU waters have aimed to develop robust methods of estimating discards [1](Sigridur Sigurdardottir et al, 2015), allowance for discards in fishery management plans [2](Nekane Alzorriz et al, 2018), development of by-catch management plans, and promotion of best practices for by-catch reduction and mitigation of incidental catches. Additional studies at national and regional levels are required to improve the precision of the global estimation of discards [3](Kelleher, 2005). Also, recent updates [4](Pérez Roda et al., 2019) establish that an understanding of the relative importance of factors affecting indirect fishing mortality is necessary to estimate the total fishing-induced mortality [5](Bellido et al., 2007), to design and implement mitigation measures, and to provide a baseline for a time series of global marine fisheries discards in the last 20 years. Also, scientists discuss whether the current sampling regime should be either expanded or complemented by other data sources, such as comparable discard per unit of effort (DPUE) between fisheries and regions. In Europe, the Data Collection Framework (DCF) [6](EEC, 2017) legislates on how to estimate the amount of discards. As part of nationally adopted onboard observer programmes, trained personnel collect discard data from their most important commercial fisheries [7](Uhlmann et al., 2013). The latest reform of the EU Common Fisheries Policy [8](EU No 1380, Article15, 2013) considers the landing obligation as a basic regulation of the policy, which required a gradual phasing-in of the obligation to land all catches. This landing obligation contains numerous exemptions between fleets affected by total allowable catches (TACs) or quotas [4](Pérez Roda et al., 2019). In addition, the 2009 revision of the Data Collection Regulation [9](EU, 2008a) has changed the focus of fisheries sampling programmes, recognising that the new context required new methodological developments to analyse the collected data since the implementation of the DCF [10](Jardim et al., 2015).

Stocks in European waters have been categorized as either data-rich or data-limited/data-poor and methodologies to provide the best scientific advice have been addressed in the last years [11]y [12](ICES. 2012a, b). Nevertheless, more categories were needed [12](ICES, 2012b) and even more so in the case of stocks out of the EU waters, where the sampling, monitoring and controlling of the fisheries are subject to political, economical and social trade-offs between EU and other countries. This is the general situation for target and by-catch stocks exploited by EU fleets under sustainable partnership fisheries agreements (SPFAs) in North West Africa and managed by the Fishery Committee for the Eastern Central Atlantic (CECAF). Discards also automatically inherit the condition of data poor from the

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62 fishery and are even more difficult to analyze since they are sampled only by observers onboard
63 [19]([STECF, 2016](#)).

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65 Generally, this study aimed to reveal the condition of data poor in stocks outside EU waters which are in
66 need of: *i*) gathering further information through logbooks (haul geo-referenced and retained and
67 discarded catches); *ii*) methodologies adapted to their particular data poor conditions; and, in particular
68 for discards, *iii*) this study responds to issues mentioned in the beginning of this introduction: to develop
69 a robust method to estimate discards, improve the precision of the estimates, understand the relative
70 importance of factors affecting discards and develop a comparable DPUE between fisheries and regions.

71
72 In particular, our case of study is the Spanish fresh trawling fleet targeting black hake, where discard
73 information has been collected by scientific observers throughout an on board programme [6]([ECC,](#)
74 [2017](#)). This is a reference fleet in CECAF, as it has been monitored by Spanish Administration and Spanish
75 Oceanographic Institute since 1989 and is the principal fleet targeting black hakes in North-West Africa.
76 The annual catches of this fishery ranged widely from 2007 to 2016, and comprised between 4800 t in
77 2013 and 17,000 t in 2016 [13]([FAO, 2018](#)). Global discards of this fleet are very fluctuating (both in
78 species composition as in volume of catches), depending on diverse issues (depth, fishing strategy,
79 season, among others).

80
81 Neither black hake stock nor the discards of the EU and African local fleets are fully in line with ICES DLS
82 categories. First, there is a lack of stock unit definition; under *Merluccius* spp. (black hake), two different
83 species are mixed up, *Merluccius polli* and *M. senegalensis* [14], [15], [16], [17], [18]([Fernández-Peralta](#)
84 [et al., 2011; 2017; Rey et al., 2012, 2015, 2016](#)). Second, the stock distribution of black hakes is not
85 considered in the assessments, as they are been assessed either independently by country (Morocco,
86 Mauritania and Senegal) or for the total region. Third, catches and effort data are collected by local
87 countries from their own fleets and, even from EU fleets, but data are insufficient or not
88 available/published for scientist. Fourth, there are no fishing surveys information to provide direct
89 knowledge of the stocks and biological parameters are unknown. Regarding discards, , the information is
90 poorer, only known by the observer programs (EU fleets), although it is known that discards are an
91 important part of the catches in most fleets.

92
93 One strict limitation in the Spanish fleet under study, concerning the researchers, is that logbooks are
94 not much valuable from a scientific perspective. Trip information is aggregated by fishing day, where
95 each daily set is not differentiated or geo-referenced separately. Also, as in many other fisheries there is
96 no information of discards in logbooks. Fortunately, an on-board sampling programme is currently
97 ongoing in this fresh fleet to obtain direct geo-referenced information on commercial and discarded
98 [20]([García-Isarch et al., 2019](#)). Also, other relevant information regarding fishing techniques,
99 environmental data, skipper skills and vessel characteristics is recorded as well. Setting up this observer
100 programme faced several technical and economic problems that have affected its continuity. However,
101 a very consistent set of data on discards has been analysed from 2016. Until now, information on
102 discards of Spanish fisheries in the area has been presented within the DCF framework but has yet to be
103 fully exploited. However, it is of paramount importance for the assessment of marketed species, for the
104 analysis of the impact on biodiversity of different fisheries, and in order to comply with EU requirements
105 for quantification of discards to improve management measures for regulating catches by TACs [8]([EU,](#)
106 [2013](#)).

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108 Several studies have analysed the components affecting discarding practices in trawling fisheries [21],
109 [22], [23], [24], [25], [26]([Mahévas et al., 2004; Carbonell et al., 2017; Vilela, 2015; Rochet et al., 2005;](#)
110 [Da-Rocha et al., 2018; Feekings et al., 2012](#)), and some studies also distinguish and quantify catchability
111 components in these types of fisheries, which affect both targeted and bycatch species [27], [28], [29],
112 [30], [31], [32]([Mahévas et al., 2011; Mamouridis et al., 2013; Pennino et al., 2014; Gascuel et al.,](#)
113 [1993; Gavaris,1980; Marchal et al., 2003](#)).

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122 This study describes the main effects on discards variability in relation to abundance covariates and
123 technical characteristics of the fleet, identifies two different métiers within the fresh trawl Spanish fleet
124 [33](Castro, 2012), and examines the strategy of the skipper. We also define a global DPUE and propose
125 a method of DPUE standardisation for the Spanish fresh trawling fleet in north-west African waters,
126 based on the explicit distinction between abundance and technical factors that affect variations in
127 discarding rates. The proposed method is applied to the detailed data gathered by observers on-board
128 commercial trawlers targeting black hakes, in surveys from 2016 to 2018. Obtaining standardized global
129 DPUE trends in this study is a first step previous to elaborate detailed and comparable DPUE by species,
130 once we have established the main factors affecting global discards. Finally, through this study we
131 would like to emphasize that, at least from the scientific point of view, is essential and feasible to
132 register the daily information by set in logbooks instead of information aggregated by day. So total
133 discards could be estimated with extrapolation techniques from observers surveys.
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136 2. Materials and methods

137 2.1. Observer data

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139 Data from 606 hauls along 31 surveys of one-week each were analyzed in Mauritanian and in Moroccan
140 waters (Western Sahara) (Figure 1), at latitudes between 17°-24° N from January 2016 to December
141 2018 (Table 1). Hauls were performed at depths comprised between 93 and 815 m. The surveys
142 encompassed the spatial-temporal distribution of the Spanish fresh trawling fleet in north-west African
143 waters. Observers surveys covered the 16% of the activity of the whole fleet during the period 2016-
144 2018.
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146 Vessel characteristics were similar and no detailed information on their characteristics (i.e. mesh, shape
147 and material of the net) was used. Data was explored graphically using ggplot2 package in R (Wickham,
148 2016), and we tried to analyse the personal skipper preferences and skills, considered as part of the
149 causes affecting discards (Kelleher et al., 2005). Since our knowledge about fishing techniques is limited
150 and not quantified at this stage, the skipper factor masks a variety of variables relating to the fishing
151 power of the vessel and, in particular, selectivity of the gear. We analysed the relationships between
152 space and time factors and the skipper through interaction plots, in order to visualise the possible
153 fishing strategies of each skipper and their effects on discards.

154 Table 2 describes the general information recorded for each haul and classified in groups of information
155 related to abundance, environment, efficiency and strategy of the skipper. We considered the route as a
156 covariate related to the dominant currents in the area (Ramos et al., 2017). Hauls carried out between
157 6:00 and 15:59 were considered daylight hauls; between 16:00 and 18:59 evening hauls; and between
158 19:00 and 4:59 night-time hauls.

159 2.2. DPUE model

160 In the absence of oceanographic fishing surveys, commercial catch and effort data are the alternatives
161 for estimating abundance indices. In multispecies fleets, different approaches have been applied to deal
162 with the impossibility of splitting the effort by species (Soto et al., 2009). Discarding rates are an
163 example of this issue and have been discussed in several studies (Pérez Roda et al., 2019; Uhlmann et
164 al., 2013; Carbonell et al., 2017). At this initial stage, we are interested in the estimation of the trend of
165 the total discard. Hence, this study deals with populations of total discards as a stock unit
166 (Restrepo, 2000), interpreted as a term susceptible of being monitored, managed and assessed, and it is
167 considered as a group of populations with similar spatial geographical limits affected by the Spanish
168 fresh trawl fishery. Discarded catches are defined by aggregating the total catches of all species
169 discarded by haul. A suitable model for DPUE is:

$$170 D/E = qN \quad (1)$$

171 where q denotes the catchability coefficient, and N the total abundance of the discarded population
172 (Mahévas et al., 2004). Therefore, two vessels fishing at the same time in the same area and with the
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180 same nominal fishing effort can have different discarding rates due to variations in catchability and/or
181 abundance (Beverton and Holt, 1957) of the global discard. One of the definitions of catchability is the
182 probability of a fish being caught by one unit of effort (Marchal et al., 2003). (2)
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184 In the Spanish trawl fishery, fishing effort is traditionally measured in number of fishing days.
185 Nevertheless, set duration has been calculated from observer data to define a nominal measure of
186 effort for the trawl fishery. This measure is a more accurate representation of fishing mortality, as the
187 set durations positively related to the area swept by the net. When a trawl vessel increases its total time
188 available for fishing, it will take more fish per set (Hilborn and Walter, 1992). Nominal observed total
189 DPUE (kg/h) was calculated for each haul as follows:

$$DPUE_i = \frac{totaldiscard_i}{setduration_i}$$

192 where i refers to haul i .

193 Space-temporal heterogeneity in distribution of discards, and changes in efficiency of the fleet result in
194 DPUE data that are not proportional to the total discarded population, and hence equation (1) is not
195 met. To approximate the equality, we fitted a model to the DPUE data that will be used as the discard
196 abundance index, taking into account the changes in factors related to abundance and efficiency.
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198 2.3. Model selection

200 To identify the main effects on DPUE trends, we fitted separate models to each of the four groups of
201 covariates in Table 2 to make a preliminary selection of the significant factors and interactions. Then, by
202 gradually nesting the covariate groups, we observed the sequential contribution of each of them to the
203 explained deviance of DPUE variability. Month factor captures the seasonality in the discards. Total
204 allowable catch (TAC) on hake fisheries usually reach the limit at the end of the year. The TAC is
205 therefore related to the fishing strategy, which involves both the choices made by the skipper between
206 Moroccan and Mauritanian waters (ground), and the target species(hake or other species). A numerical
207 variable was defined to consider this fact. The TAC for the Spanish trawl fishery was established at 6000
208 tons of annual total catch in Mauritanian waters during the study period. For sequential hauls,
209 cumulative catch throughout the year was therefore introduced as a continuous variable to explain
210 decisions taken by the skipper and mainly related to target species. Vessel power was considered in two
211 ways: first, through the technical characteristics of vessels, such as HP and GRT; and second, through the
212 vessel factor instead of vessel characteristics. In this case, vessel encompasses all the encrypted
213 technical information not available at this stage. Haul order is important during the survey, since the
214 decisions about targeting hake or mixed species could affect the discarding rate and therefore, it was
215 included as a numerical covariate in the DPUE model. Regarding the environment, previous studies
216 (Meiners, 2010) had shown cross-correlation between catch per unit of effort (CPUE) of black hake and
217 North Atlantic Oscillation (NAO) index three years before, over a 40-year study period. In our case, the
218 relationship between discards and the NAO index was explored to determine the possible influences on
219 the time interval between 2016 and 2018. Finally, we assessed the contribution of DPUE variability due
220 to the skipper as a random effect, in order to take into account the changes in inter-skipper efficiency
221 that varied randomly. We intended to ascertain the extent of the skipper effect on discards, though we
222 were not interested in knowing the exact nature of this effect. We assume that the variation around the
223 intercept, for each skipper, is normally distributed with a certain variance (Feekings et al., 2012).

224 Hence, variation in discarding rates was investigated through the most common models, Generalised
225 Linear Models (GLMs) (McCullagh and Nelder, 1989) in R (R Core Team, 2014), Generalised Linear Mixed
226 Models (GLMMs) (Pineiro and Bates, 2000) and through Generalised Additive Models (GAM) (Hastie and
227 Tibshirani, 1990).

228 Visualisation of histogram, previous QQ-plots and goodness-of-fit tests, Kolmogorov-Smirnov and Chi-
229 squared (Draper and Smith, 1981) were performed to select the best fit for DPUE distribution.

230 First, the GLM method (McCullagh and Nelder, 1989) was applied with log-normal error distribution of
231 DPUE, to determine the set of systematic factors and interactions that significantly explained the
232 observed variability in each model, based on the Akaike Information Criterion (AIC) (Akaike, 1974). A
233 Chi-squared test was used to evaluate the statistical significance of an additional factor between the set
234 of systematic factors and the interactions (McCullagh and Nelder 1989). Furthermore, the
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 239 corresponding percentage of deviance explained by each factor relative to the maximum model was
 240 estimated to obtain a profile of the most important explanatory factors in the model (Mamouridis *et al.*,
 241 2013). A statistically significant variable may, in some instances, be omitted from the model if the
 242 amount of variation explained by the variable is small in relation to the complexity that it adds
 243 (Stefánsson, 1996). After fixed factors were assessed, skipper was tested as a random effect in the form:

$$244 \log DPUE_i = \alpha + \sum_{j=1}^n \beta_j \cdot x_{ij} + \text{random factor}(1|\text{skipper}_i) + \varepsilon_{ij}, \varepsilon_{ij} \sim N(0, \sigma_{ij}^2)$$

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 247 where β_j are the linear effects of the fixed factors for all hauls, $i = 1, \dots, 608$.

248 249 2.4. DPUE abundance index

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 251 Year effect is interpreted as a relative index of abundance of DPUE. The relative abundance DPUE index
 252 was obtained as the least squares means (LSM) of the year effect in the GLM (Littell *et al.*, 2006), using
 253 the lsmeans package in R (Lenth, 2016). In unbalanced, multi-way designs, the LSM estimation is often
 254 assumed to be closer to reality, as if the data were balanced, because LSM somehow correct the
 255 design's imbalance. LSM are computed based on the linear model, and the variance of the index is
 256 estimated from the LSM standard deviations. Model evaluation and diagnosis were carried out through
 257 residual analysis (McCullagh and Nelder, 1989).

258 259 3. Results

260 261 3.1. Observer data

262 263 *Global discards*

264 Global averaged discard in the sampled hauls during 2016 to 2018 accounted for the 39% of total
 265 catches in surveys. By year, this percentage varied from 43% en 2016, 39% in 2017 and 32% in 2018.

266 267 *Métier definition*

268 Exploratory analysis confirms the presence of two separated métiers of the Spanish fresh trawling fleet.
 269 Along the period 2016-2018, 87% of the hauls targeted hake, while the remainder targeted a mix of
 270 several other species. Average haul duration was 4.8 hours and varied between 1 and 9 hours. The haul
 271 bathymetric distribution showed two main fishing groups (Figure 2). The first group corresponded to the
 272 shallowest hauls (<300 m), that were done in Mauritanian waters only during daytime. The second
 273 group corresponded to the deepest hauls (>300 m) done in Morocco and Mauritania throughout the
 274 whole day. The coral reef between 400 and 500 m appeared as a boundary. In shallower waters, the
 275 predominant catches consisted of mixed species.

276 Distributions of total catches of discards, hake and other species by depth are shown in (Figure 3).
 277 Discards were higher and more dispersed in shallow than in deeper waters, exactly the opposite to
 278 hake. A small fraction of hake catches was captured up to 300 m, where a mix of species other than
 279 hake predominated. These results showed that depth determines the type of fishing strategy and,
 280 indeed, generates two different métiers for the Spanish trawling fleet. Hence, we established two
 281 alternative strategies (métiers) for the trawling fleet, based on target species and depth:

$$282 \text{target} = \begin{cases} \text{hake} & \text{hauls} < 300 \text{ m depth} \\ \text{other species} & \text{hauls} \geq 300 \text{ m depth} \end{cases}$$

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 285 The distribution of the log-transformed discard rates showed different patterns by target species
 286 (Figure 4, left panel). Hauls targeting hake, and hauls targeting other species appeared to come from
 287 two different probability distributions according to the histogram patterns. Hence, the evidence of two
 288 métiers indicates that discard analysis must be made separately accordingly to this classification. Clear
 289 differences were also observed in all the covariates by target, validating the evidence of two métiers.
 290 Once this classification of total observer data was assumed, we focused our analysis to model a robust
 291 relative abundance discard index of the Spanish trawl fishery in the hake métier.

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298 Total discards for the hake métier were always greater than zero, so DPUE had a positive distribution.
299 The histograms, QQ-plots and Kolmogorov test showed that the best distribution followed by the DPUE
300 is lognormal (Figure 4, right).

301 Abundance and environmental covariates revealed systematic inter- and intra-métier differences in
302 DPUE (Figure 5). Significant differences in the year, month, ground and depth affected changes in DPUE,
303 as did latitude. Despite the apparent nonlinear relationship between the average monthly NAO index
304 and the DPUE, this covariate was not considered in the standardisation model, due to the short study
305 period of discard rates, which limited the interpretation of results. Further analysis are needed for a
306 extended period to investigate effects of the environment on discards.

307 The fishing strategy factors also reveal differences by métier. Mixed species are only caught during
308 daytime and, in general, show higher associated discard rates than hake.

309 To explore the vessel effect we consider by one side, the vessel technical characteristics of the fleet, HP
310 and GRT and, alternatively, only the vessel instead its own characteristics. DPUE is always lower when
311 the target species is hake, independently of the vessel, the HP levels or the GRT values. Also, differences
312 by HP, GRT or vessel are not significant in the hake métier. On the contrary, these differences are more
313 evident when target species is not hake (mixed). One particular vessel showed less and similar discards
314 rates in both métiers than the other vessels. In fact, this particular vessel had the highest values of HP
315 and GRT. This could mean that the effect of the vessel or the technical characteristics are not as
316 important as the skipper effect, i.e. the skipper who worked in the best vessel had the lower discard
317 rates. Another conclusion could be that the best vessel also has the best skipper combining the best
318 technology with the best fishing practices.

319 *Skipper effect*

320 Figure 6 confirm that the most efficient skipper (s3) minimised discard rate in both métiers. As skipper
321 interactions with abundance, efficiency and fishing strategy were unbalanced due to lack of data in
322 several of all the possible combination of covariates, they were investigated through interaction plots
323 (Figure 7) or contingency tables (not presented). The most efficient skipper fished in deeper waters than
324 the others, clearly reaching the minimum discard rates in Mauritanian waters, followed the south-east
325 route and operated between evening and night hours.

326 *3.2. Model selection for DPUE*

327 Robust GLM models were fitted to selected data excluding outliers. A abundance covariates effects on
328 the logarithm of DPUE of the GLM abundance model are shown in Figure 8. Discards were less abundant
329 in the deepest waters, and DPUE increased with latitude, which was consistent with the observed
330 increase in nominal DPUE in Moroccan waters. Annual discarding rates fell sharply from 2016 to 2018.
331 Month effect captures both the seasonality and a global decreasing trend throughout the year,
332 confirming the decline of discards at the end of the year, when the hake spawning season begins and
333 the fishery moves in search of adults in deeper waters. After a exploratory model of abundance effects
334 was fitted, the remaining separated groups of effects were examined. After selection of covariates in
335 each group, nested effects and interactions were fitted (Table 3). Values of the percentage of explained
336 deviance, AIC and BIC (Bayesian Information Criterion) (Schwarz, 1978) were calculated by group of
337 effects and the absolute difference between AIC and BIC. Not all models were nested because of the
338 interactions. Therefore, predictive power and significance of covariates were tested according to more
339 than one criterion, apart from the principle of parsimony. There is always a chance that the AIC will
340 choose a model that is too big, regardless of the number of observations. The BIC has very little chance
341 of choosing too big a model if the sample is sufficient, but it has a greater chance than AIC of choosing
342 too small a model.

343 There were no differences between the two alternative models of efficiency, since vessel factors alone
344 explained as much as vessel characteristics. Variability in efficiency was better explained by GRT than by
345 HP.

346 The NAO index modelled by the GAM with a Gaussian response is too short to assess fluctuations or
347 seasonality in DPUE, or to explore time series correlation across different lags. However, it does explain
348 7.36% of the deviance in DPUE. The significance of the smooth term evidenced that a smooth curve was
349 needed compared with a linear, but at this stage, we do not consider environment for standardisation
350 through this variable due to the short period considered in this study.

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357 The selected covariates of each effect were summarised in Table 6. Once the selection of covariates in
358 the separated groups of abundance, efficiency and strategy were investigated, stepwise models were
359 fitted starting with the abundance covariates, followed by terms of efficiency, strategy and interactions.
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361 Table 4 shows the full model with all possible selected covariates and interactions. The statistically
362 significant covariates at 95% confidence level included in the log-normal model fitted to hauls of the
363 hake métier were year, month, ground, latitude at the end of the haul, depth, route and the interaction
364 month:ground and ground:log(cumulative catches). This model explained 58.60% of total deviance in
365 DPUE data. Besides the statistical significance of covariates, we considered a covariate candidate to be
366 included in the final standardisation model if it captured more than 5% of the deviance explained by the
367 model. This is the case only for abundance covariates **year**, **month**, **ground** and **depth**. These four
368 covariates account for 82% of the total deviance explained by the model. Hence, for standardisation
369 purposes and future improvements in the collection of data and statistics, we considered that with
370 information about these four covariates is possible to provide admissible DPUE standardised index.
371 After selection of fixed effects, we nested the skipper covariate as a random effect. Inter-skipper
372 variability explained 41% of the total variance. This variability represents the existing differences in the
373 efficiency between a combination of gear characteristics and skipper skills, which is hard to quantify.
374 Both AIC and BIC chose the simplest model of fixed effects to predict DPUE rates as the best candidate.
375 Diagnostic plots (Figure 9) show that the hypothesis of error distributions in the final selected model is
376 acceptable, as there were no trends in residuals and no significant deviations from normality.

377 3.4. The index

378
379 The final DPUE standardised index (Figure 13) presented a clearly decreasing trend in DPUE abundance
380 in the short period considered. Nominal DPUE was underestimated during this period compared with
381 the standardised, but was very similar in 2018. Nominal DPUE laid under the lower confidence interval
382 of the standardised DPUE.
383

384 4. Discussion

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386 This study remarks the condition of DLS of a particular fishery in North West Africa, as it has been
387 addressed in a recent study (García-Isarch, E. 2016). While current landing discards obligation
388 (Regulation EU 1380/2013), does not affect directly to EU fleets under non-EU jurisdiction the European
389 Commission is making efforts on their monitoring at the most global extension in EU and non-EU waters
390 (STECF, 2016). Furthermore, in North West African waters discards are not managed in this area by local
391 countries.
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393 The assessment of global marine fisheries discards in nowadays progressing, taking advantage of
394 available data and promoting studies focused in quantifying discards and standardizing sampling
395 protocols (Pérez Da Roda, 2019). In this frame, it has been undertaken the task of quantify discards in
396 the Spanish fresh trawler fleet operating in NW Africa, a fishery dealing with serious limitations of data
397 but with a challenging potential to lead improvements in the assessment framework for African stocks.
398 In particular, this fleet target mainly black hakes, *Merluccius polli* and *M. senegalensis*, composing the
399 vast majority of the retained catch and a portion of discards.
400

401 From a total of 260 stocks assessed by ICES, 60% are considered DLS (ICES, 2012a). It would be desirable
402 to extend this classification to stocks exploited by EU fleets under SPFAs, considering the particularities
403 of these fisheries. Under DLS, target stocks from SPFAs have a particular identity, mainly because EU
404 fleets operates under the sovereignty and jurisdiction of third countries, which assess the stocks
405 attending to political boundaries (Fernández Peralta et al. 2019; CECAF, 2018) instead of real stock
406 identification.
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408 Stock assessments of DLS in African countries are limited by SPFAs discontinuities and methodological
409 issues. Nevertheless, significant improvements in the sampling programs of the Spanish fleets can be
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416 achieved. Although Spanish fisheries administration collect logbook catch information from national
417 fleets, discard data are only recorded from observer programs on-board, managed by scientific national
418 organisms (Spanish Oceanographic Institute and AZTI). Logbooks catch data are aggregated and
419 recorded daily, without specifying the correspondent to each fishing operation. Furthermore, no
420 information on discard volume is recorded in logbooks. From a scientific point of view it is desirable to
421 expand the data gathered in logbooks to each fishing operation as well as include an estimation of total
422 discard volume per haul. This study shows the usefulness of on-board observer data in identifying
423 factors and their relative effect on discards, as well as its composition. Such data provide quantitative
424 discard information, essential to establish priorities in the global management of discard practices, and
425 are a first step in the ecosystem approach (Bellido *et al.*, 2011), aimed at maintaining sustainable levels
426 of catches on target species, while keeping discards under control. Unfortunately, the discontinuity and
427 vulnerability of observer programs threaten the continuity of annual updates of discards statistics,
428 including estimation of DPUE.
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431 This study describes the main effects on discards variability in relation to abundance covariates and
432 technical characteristics of the fleet, identifies two different métiers within the fresh trawl Spanish fleet,
433 and examines the strategy of the skipper. Also a global DPUE is defined and standardized for the Spanish
434 fresh trawling fleet in north-west African waters, based on the explicit distinction between abundance
435 and technical factors that affect variations in discarding rates. Standardized global DPUE trend is a first
436 step previous to elaborate detailed and comparable DPUE by species, once the main factors affecting
437 global discards have been established.
438

439 *Métier*

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441 Two métiers were identified and defined in the Spanish fresh trawl fishery: one targeting 'hake' (>300
442 m) and other targeting 'mixed species' (<300 m). As in other studies for metier identification (Castro,
443 2012), multivariate techniques (PCA) have been used to validate this result in species composition of
444 discards. Haul data (geo-referenced and by bathymetric levels) were essential to conclude that depth is
445 the most important spatial variable determining the amount of discards and clearly defining two métiers
446 with different fishing strategies in this fishery. This point should be considered in the compilation of
447 fisheries statistics, assessment and management, not only for black hake fishery landings, but also for
448 total catches (Vilela and Bellido, 2013; European Parliament Plenary, 2013).
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451 *Effects on discards*

452 This study shows that time-area related factors affecting catchability are the most important to explain
453 variations in discard rates. A marked seasonality characterizes this fishery, associated with the spawning
454 season of the black hakes (Fernández-Peralta *et al.*, 2011). In this season, the fleet mainly targets the
455 biggest female spawners at deeper waters, where discards usually are lower. Accordingly, a decreasing
456 DPUE trend throughout the beginning of the spawning season has been observed (October-December).
457 Other factors affecting discards are not considered here, such as the distribution of the consumption of
458 the legal quota per vessel; other market decisions or regulations; and whether or not the fish is
459 damaged (Feekings *et al.*, 2012).
460

461 Using the vessel itself as a covariate in model instead of vessel characteristics is preferred here for
462 assessing the efficiency components of catchability because its simplicity. Despite scant data on detailed
463 vessel and gear characteristics of the Spanish trawl fleet, this result is consistent with similar studies on
464 hake trawler fleets (Mahévas *et al.*, 2004, 2011). Hence, differences in the catch rates could be more
465 related to gear selectivity and to the specific fishing skills of the skipper, responding to his strategy in the
466 long term or to the tactic in the short term (Marchal *et al.*, 2006).
467

468 The most efficient skipper developed throughout 2018 an efficient strategy which visibly reduced
469 discards in both metiers in Mauritania. Future research should be carried out on 'how' skippers operate
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475 in order to thoroughly identify the features of a sustainable fishing strategy (gear setting, selection of
476 materials, fishing operations, boat dynamics and vulnerability of the species to the gear).

477 Several studies have attempted to model DPUE using a multitude of covariates related to abundance
478 and efficiency (Carbonell *et al.*, 2017; Mahévas, *et al.* 2004; 2011). In this study, the causes of discarding
479 have been investigated, but also tried to simplify the task of predicting discards in a pragmatic way,
480 taking into account the data poor conditions of the fishery. Some simple but extended information from
481 commercial trips would allow estimating a reliable discard statistics but also relative abundance index of
482 discards for the whole fleet. Adding data on *depth* and *total discard estimation* at fishing operation level
483 in logbooks can offer significant estimations of relative discard rates. Hence, from the management
484 point of view, regulations of fishing seasons, grounds and depths could reduce discards in the short term
485 in a very effective way.
486

487 488 *The index*

489 Results showed an apparent decrease of discards between 2016 and 2018. However, a reduction in the
490 discard rates of this fleet in such a short period of years can be interpreted in different ways: *i)* the
491 abundance of the whole stock is declining, and therefore overfishing consequences are emerging; *ii)* on
492 the contrary, good fishing practices are being implemented, resulting in an effective reduction of
493 discards and *iii)* a combination of these two facts. This conclusion is based on results obtained from on-
494 board observer data and not on the complete data set of Spanish fresh trawling fleet hauls. In one hand,
495 observed decreasing trend in discards in this period matches with black hake nominal CPUE trend in
496 these years (Fernández-Peralta, 2019), supporting the first option. Nevertheless, discard reduction in
497 2018 due to a particular skipper skill and ability, evidenced the second choice. Thus, a combination of
498 both does occur, although further analysis would be necessary to draw a definite picture of which are
499 the main effects driving discard trends.
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501 502 *The surveys*

503 The synergy between ship-owners, skippers, observers and scientists has indeed facilitated this study.
504 Unfortunately, designing an optimal survey for scientists is not always possible. Currently, several issues
505 (mainly ship-owner cooperation and vessels habitability) determine the availability of commercial
506 vessels for sampling. In 2018, surveys were performed only on-board one single vessel (and same
507 skipper) and therefore sampling was necessary biased. That said, the information gathered from this
508 year, skipper and vessel has been extremely valuable for improving our knowledge about the nature and
509 causes of discards, the distribution of effort in the fleet, and how the trawling fleet could succeed in
510 reducing them in the future.
511

512 513 *Future research*

514 This initial analysis of discard should be complemented by further studies, extending to other related
515 perspectives. Once a global index of discards has been calculated as proposed here, multi-specific
516 indices of abundance can be derived from multinomial or delta GLM approaches, based on information
517 of specific proportions of catches for each species (Soto, 2009). Another important forward step is to
518 extrapolate rates of discards to the whole fresh trawling fleet. Also, other fleets discards would be
519 characterized through simulation techniques, as suggested (García-Isarch, 2016) in the estimation of
520 surplus for SPFAs fleets in North West Africa or be included as percentage of misreporting catches in
521 assessment production models as SPICT (Pedersen and Casper, 2017).
522

523 Summarizing, a more precise and conclusive discard estimation would be achieved when different
524 sources of information concurred. Two basic lines for improvement have been highlighted here: first,
525 observer information should ideally cover a broader range of trips, sampling all the vessels in the fleet
526 with the commitment of all skippers and owners; and second, a more efficient template should be
527 designed to collect geo-referenced logbook information by set, recording the time of day, duration and
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534 the corresponding depth of the each fishing operation. At this stage, skippers should be disturbed as
535 little as possible when registering information haul by haul. These improvements have minimum
536 economic costs, but would lead to a better quantification of discards, a global index of discards for the
537 whole fleet, and a better assessment for the management of the target and discarded populations.
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Table 1. Survey information

SURVEY	GROUND	YEAR	MONTH	% discard in survey	Depth (m)		number of hauls by depth		hauls sampled
					min	max	MIXED (<300 m)	HAKE (>300m)	
1	MOROCCO	2016	11	52%	595	725	8	3	

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2		2016	11	40%	506	756		18	8	
3		2016	11	45%	502	763		21	10	
4		2017	12	44%	446	740		17	7	
5		2017	12	47%	558	789		13	6	
6		2018	6	43%	476	779		18	10	
7		2016	1	32%	104	725	2	17	8	
8		2016	2	35%	334	815		15	6	
9		2016	3	51%	102	712	12	14	10	
10		2016	4	50%	112	688	11	13	10	
11		2016	6	38%	112	766	6	14	8	
12		2016	6	45%	121	697	7	16	9	
13		2016	10	40%	112	735	3	17	7	
14		2017	2	48%	98	697	8	18	12	
15		2017	2	49%	106	727	9	18	14	
16		2017	4	37%	502	787		20	9	
17		2017	4	35%	493	725		23	12	
18		2017	6	43%	106	707	3	19	10	
19	MAURITANIA	2017	6	38%	478	740		23	11	
20		2017	9	33%	115	744	2	20	8	
21		2017	9	31%	132	725	2	14	6	
22		2017	11	25%	521	725		18	8	
23		2018	1	41%	225	768		18	6	
24		2018	2	34%	93	697	5	16	9	
25		2018	3	29%	100	787	8	12	9	
26		2018	4	38%	231	763		23	11	
27		2018	5	45%	651	800		15	7	
28		2018	7	25%	362	732		21	4	
29		2018	8	38%	304	690		16	3	
30		2018	9	19%	407	770		18	7	
31		2018	10	13%	409	781		15	6	
TOTAL HAULS								78	528	254

Table 2. Description of covariates for the Spanish trawl hake fishery in African waters

Variable	Unit	Number of levels/continuous	Related to
Year	From 2016 to 2018	3	Abundance
Month	From 1 to 12	12	Abundance

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Ground	Morocco/Mauritania	2	Abundance
Latitude	17.04–23.74 m	1	Abundance
Depth*	110.5–824.5 m	1	Abundance
NAO	0.01, 0.10, 0.54	Continuous	Environment
Vessel	Vessel code	3	Efficiency
GRT	209, 405 tons	2	Efficiency
HP	480, 486, 490 CV	3	Efficiency
Skipper	Skipper code	5	Efficiency
Haul number	(integer from 1 to 27)	Continuous	Strategy
Target	Mixed or hake	2	Strategy
Day_Night	Day/evening/night	3	Strategy
Route	NE, NW, SE, SW	4	Strategy
Cumulative_catching survey route	660–479861 kg	Continuous	Strategy
	NE, NW, SE, SW	4	Strategy

* average value between depths at the initial and final latitudes of the trawl.

Table 3. AIC, BIC and percentage of deviance explained for log-normal models fitted to selected variables in each group. Best option is emboldened for each criteria.

Effect	%Deviance explained	AIC	BIC	AIC-BIC
Abundance	50.19%	593.39	669.25	75.86
Abundance+interactions	53.73%	582.51	713.16	130.65
Abundance+efficiency	50.59%	591.31	671.39	80.08
Abundance+efficiency+interactions	52.59%	568.63	652.93	84.29
Abundance+efficiency+Strategy	52.68%	577.84	678.99	101.15
Abundance+efficiency+Strategy +interactions	56.28%	553.95	684.61	130.65

Table 4. Deviance table for the log-normal model. Explanatory final covariates selected for standardisation are in bold.

DPUE

Model formulation	DF	Change in deviance	Residual deviance	p-value	Percentage of total deviance
1			179.19		
Covariates					
+year	2	42.71	136.48	<0.001	40.67%
+month	11	15.06	121.43	<0.001	14.34%
+ground	1	18.39	103.03	<0.001	17.51%
+lat.stop	1	4.23	98.81	<0.001	4.03%
+depth.mean	1	9.55	89.26	<0.001	9.09%
+vessel	1	0.72	88.53	0.034	0.69%
+haul	1	0.23	88.30	0.234	0.22%
+logcumcatch	1	0.62	87.68	0.050	0.59%
+route	3	3.21	84.47	<0.001	3.05%
Interactions					
+month:ground	1	3.95	80.53	<0.001	3.76%
+ground:lat.stop	1	0.71	79.82	0.036	0.68%
+month:depth.mean	11	1.44	78.37	0.631	1.37%
+ground:haul	1	0.35	78.03	0.143	0.33%
+ground:logcumcatch	1	2.63	75.40	<0.001	2.50%
+depth.mean:haul	1	0.08	75.32	0.476	0.08%
+depth.mean:logcumcatch	1	1.12	74.20	0.009	1.06%
+vessel:logcumcatch	2	0.02	74.18	0.929	0.02%

Figure 1. Hauls distribution map.

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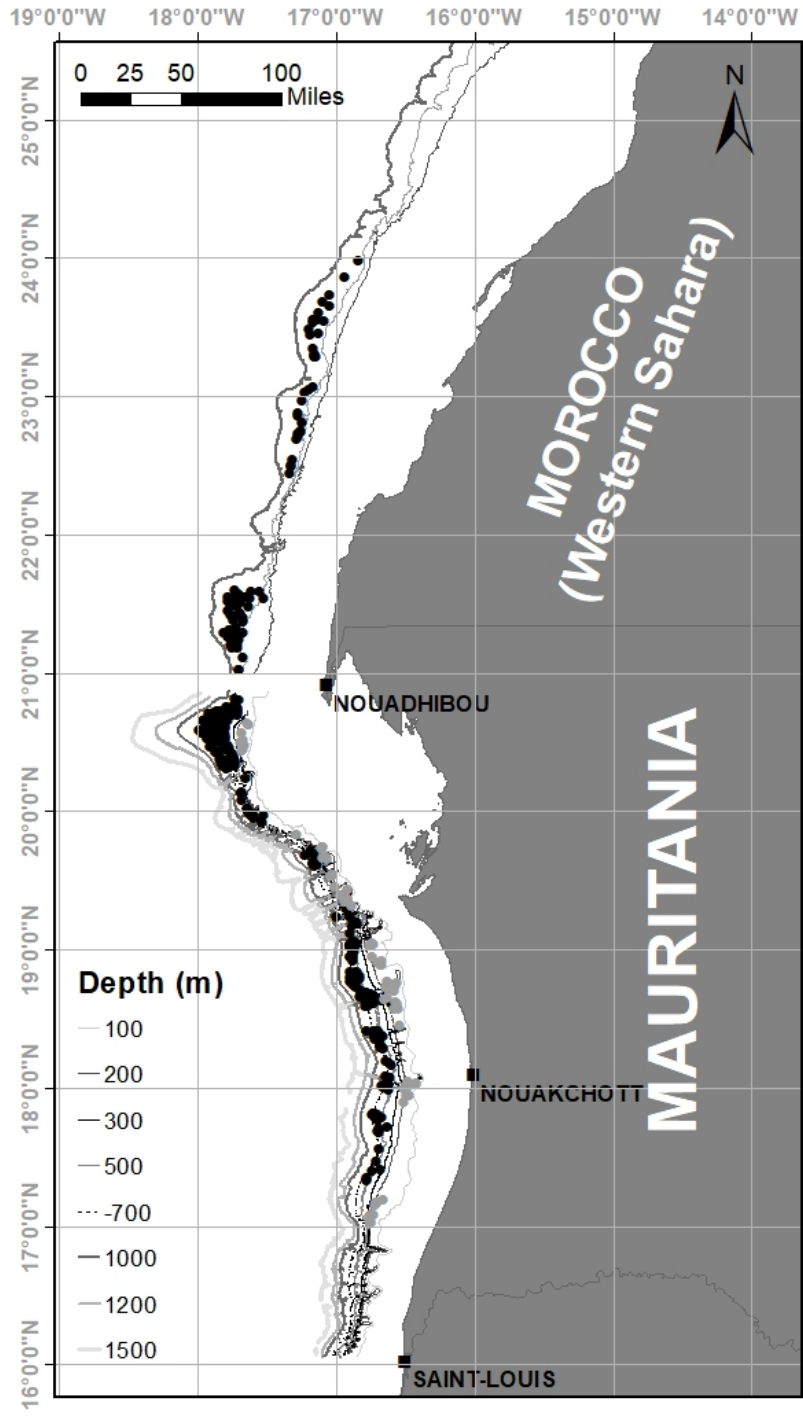


Figure 2. Proportion of hauls by depth .

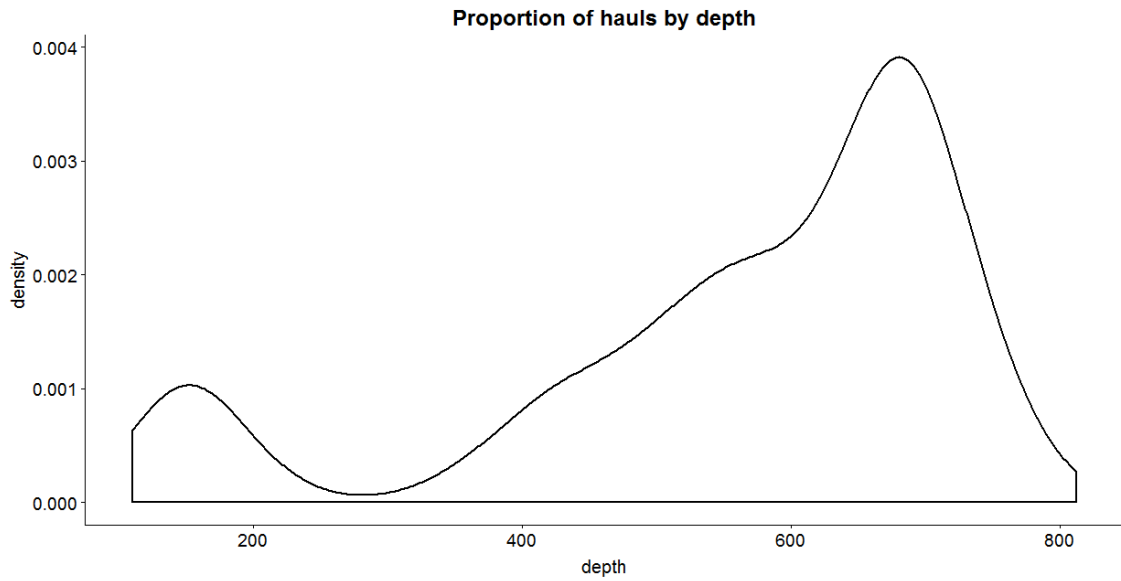


Figure 3. Discards, catches of hake and other species vs. depth.

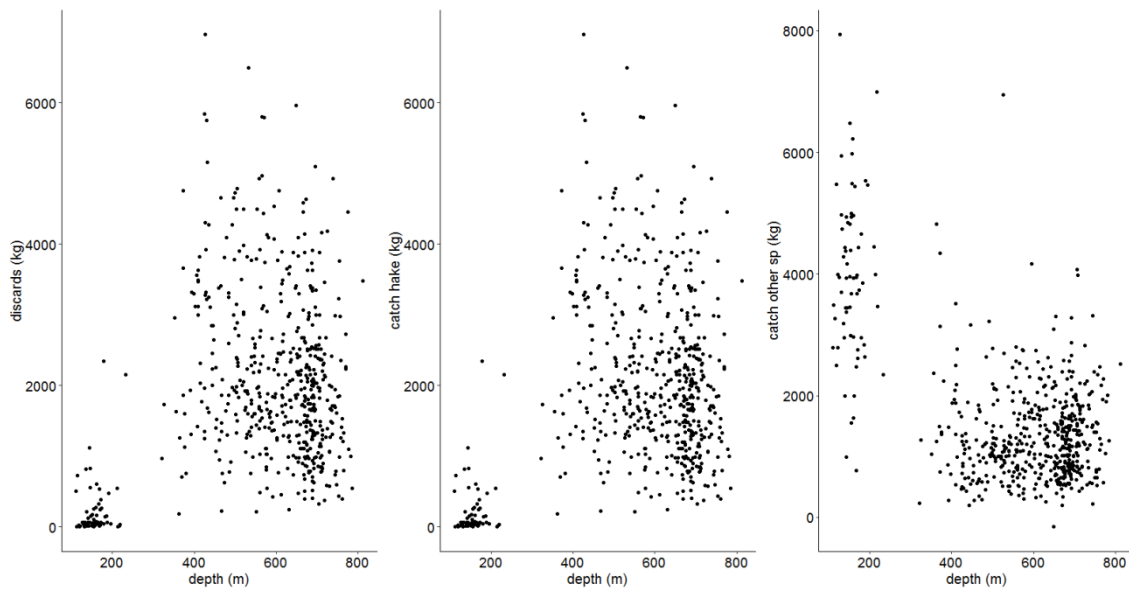


Figure 4. Histogram of log(DPUE) for hauls targeting hake compared with normal density (right panel) and histogram of log(DPUE) by target species (left panel).

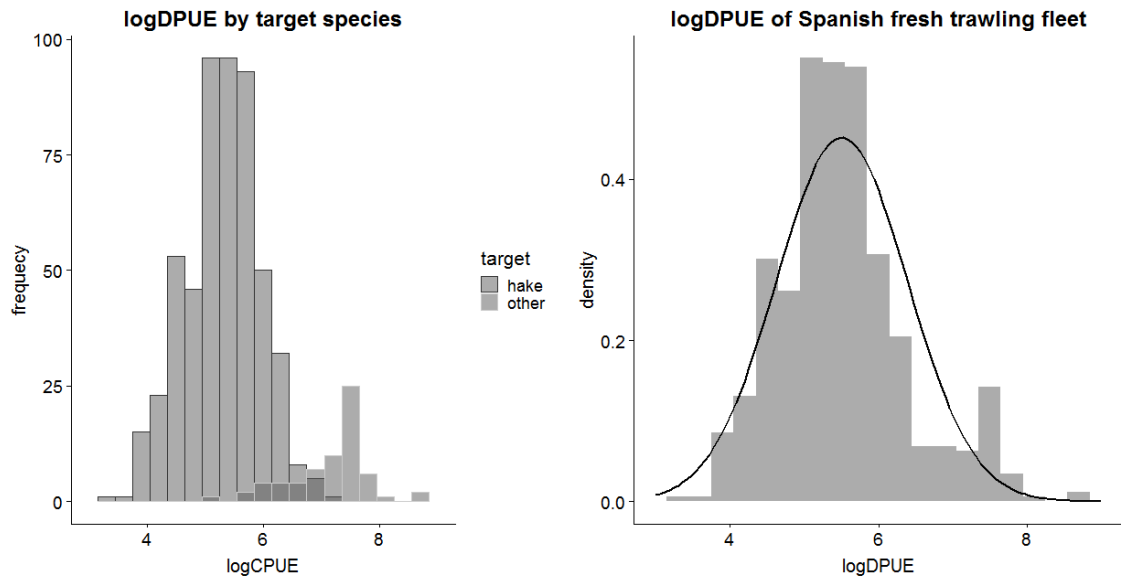


Figure 5. log(DPUE) vs. abundance covariates by target species.

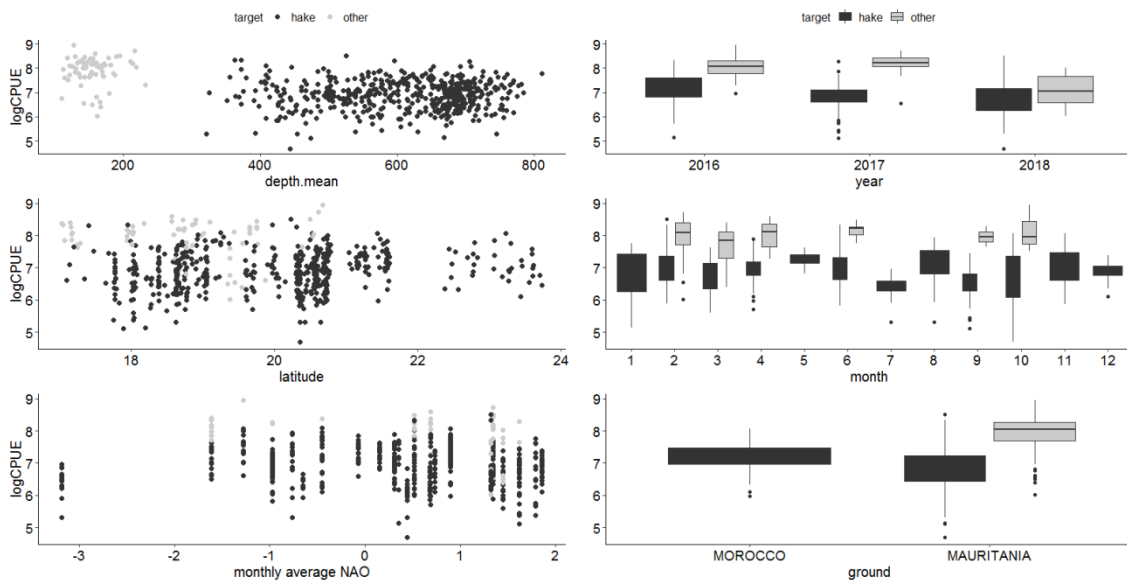


Figure 6. Discard rate by skipper and métier.

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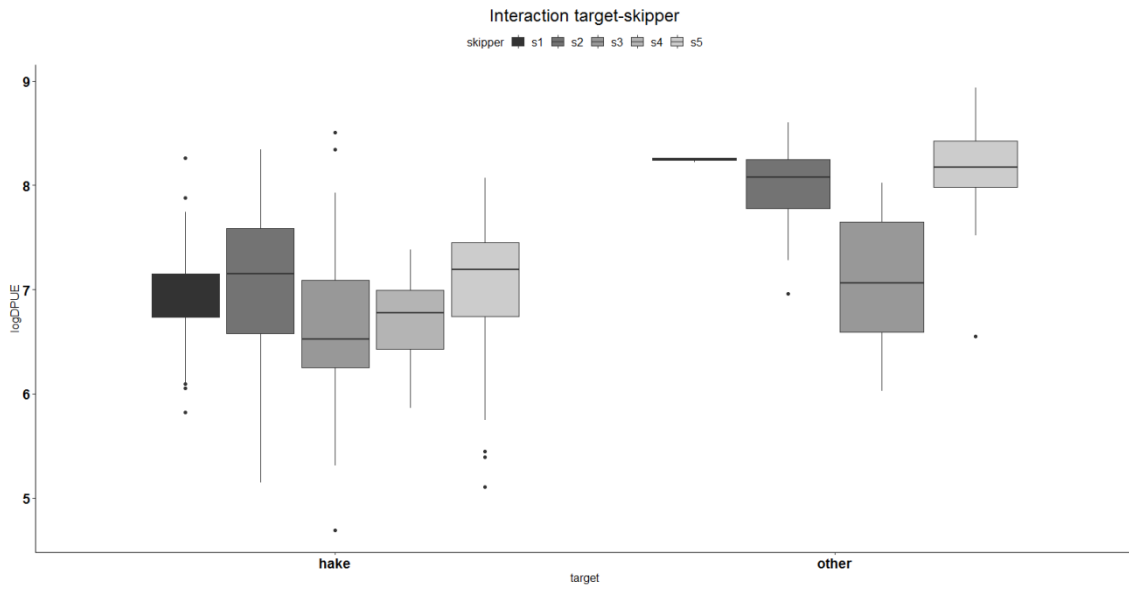


Figure 7. log(DPUE) for skipper interactions with ground, day-night, route and target species.

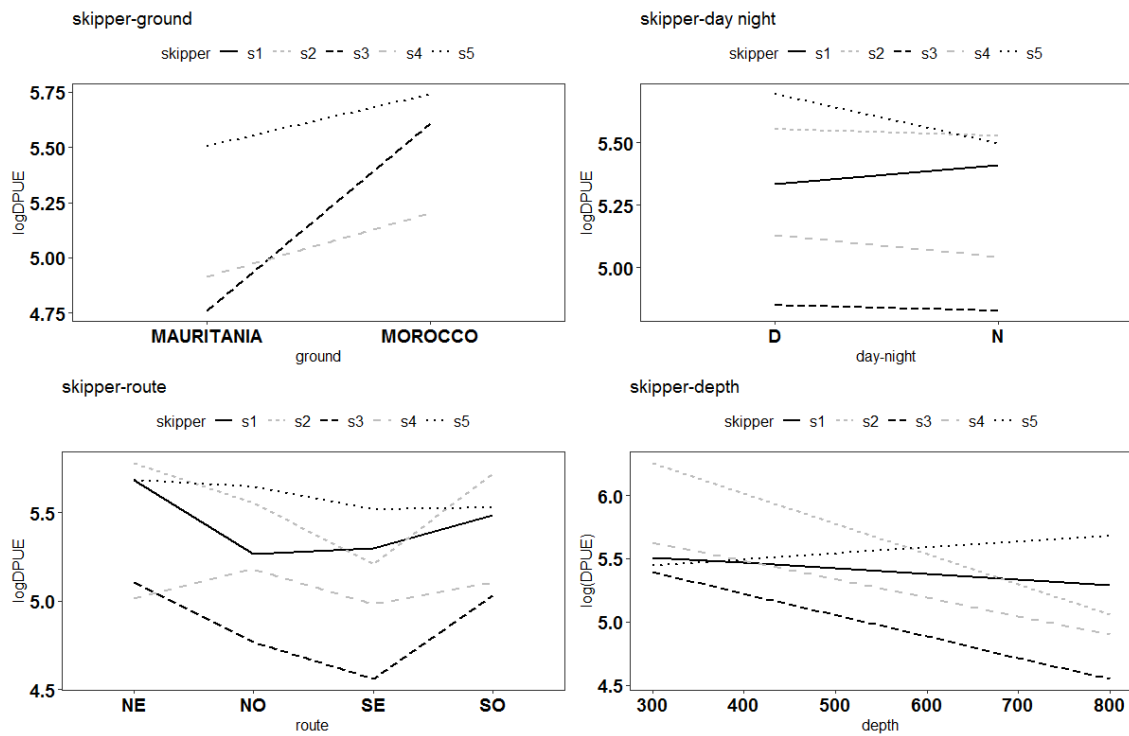


Figure 8. Effects of the GLM model fitted to abundance covariates.

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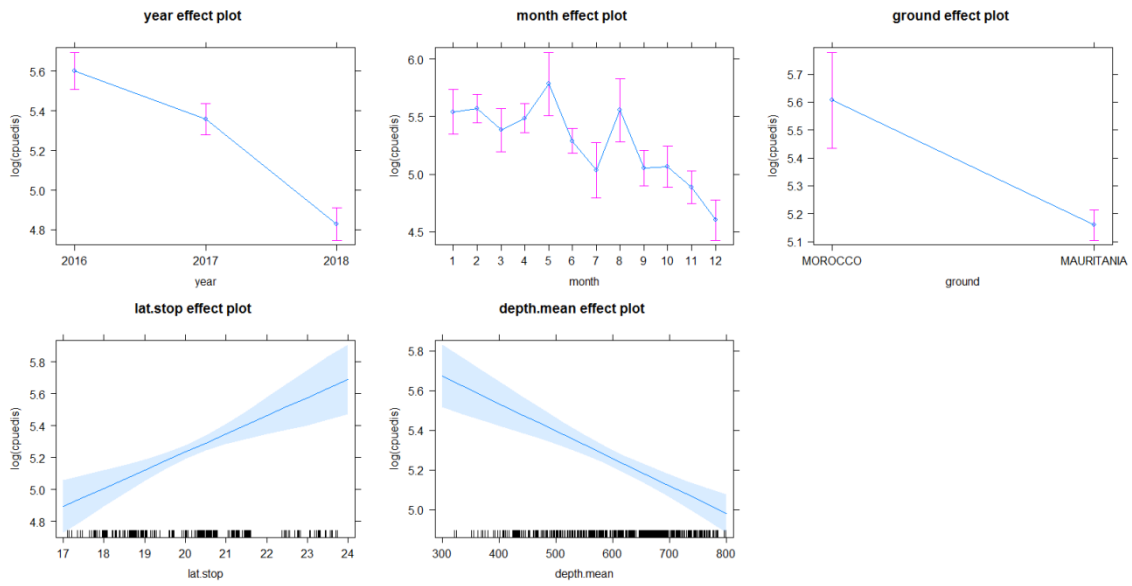


Figure 9. Diagnostic plots for the residuals of the selected log-normal GLM.

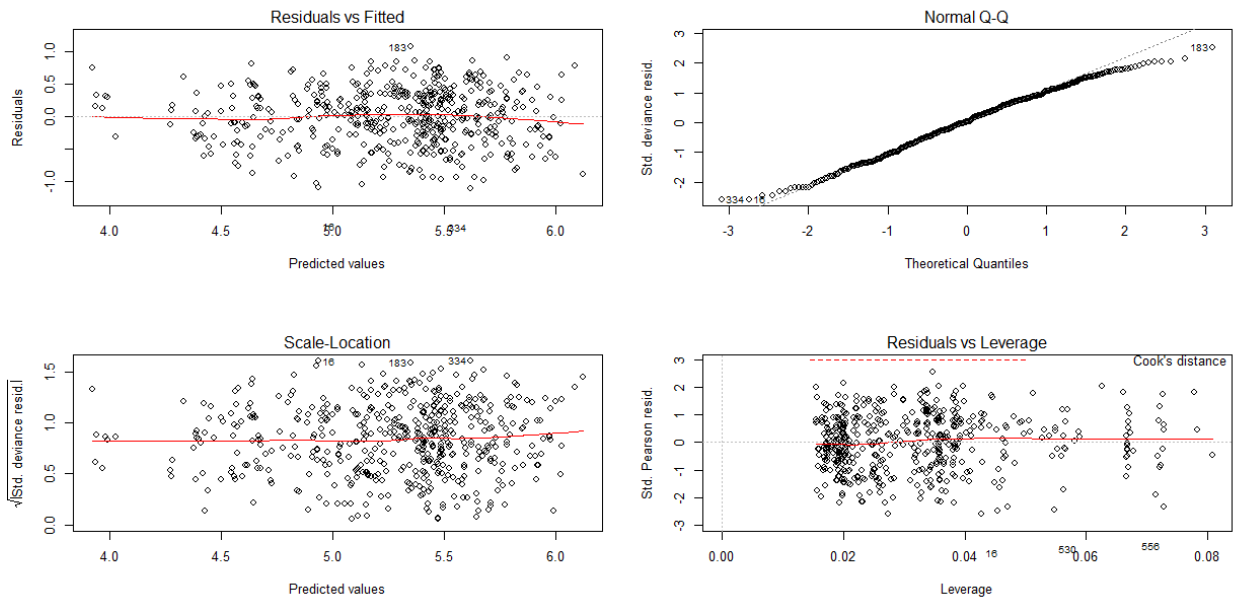


Figure 10. Scaled nominal and standardised DPUE for the Spanish trawl fishery.

