

# Drifting Fish Aggregating Devices (DFADs) as ecological traps: no evidence displayed by a long-term analysis of yellowfin tuna condition

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 ecological traps: no evidence displayed by a

3 long-term analysis of yellowfin tuna condition

## 4 Authors

- 5 Amaël Dupaix<sup>1\*</sup>, Laurent Dagorn<sup>1</sup>, Antoine Duparc<sup>1</sup>, Aurélie Guillou<sup>1</sup>, Jean6 Louis Deneubourg<sup>2</sup>, Manuela Capello<sup>1</sup>
- 7<sup>1</sup>: MARBEC, Univ. Montpellier, CNRS, Ifremer, IRD, Sète, France
- 8 <sup>2</sup>: Unit of Social Ecology, Université Libre de Bruxelles (ULB), Bruxelles,
  9 Belgium
- 10 \* corresponding author: e-mail: amael.dupaix@ens-lyon.fr

#### 11 Abstract

12 Habitat modifications major human-induced rapid are one of the environmental changes. Drifting Fish Aggregating Devices (DFADs), used by 13 14industrial purse seine tropical tuna fisheries, significantly increased the 15number of floating objects with which tropical tuna associate, raising 16 concerns over the risk of modifying the behavior and altering the biology

17of these fish (the so-called ecological trap hypothesis). Relying on a time-series of more than 25,000 length-weight samples collected in the 18 western Indian Ocean (1987-2019), we reject the hypothesis that the body 19 condition (Le Cren's relative condition factor  $K_n$ ) of yellowfin tuna 20 21 (Thunnus albacares) decreased concurrently with the increased number of 22 DFADs. This result suggest the absence of negative long-term impacts of 23 DFADs on the biology of tuna. However, as other factors may have 24 counteracted possible negative effects of DFADs, we recommend a long-term 25 monitoring of habitat, biological and behavioral parameters of tunas to 26 detect any critical change.

## 27 Keywords

Indicator log, relative condition factor, *Thunnus albacares*, Indian Ocean,
industrial tuna fisheries, floating objects

## 30 Introduction

Floating objects (FOBs), such as logs or branches, are natural components of the oceanic habitat of tropical tunas, which naturally associate with them. If the reasons for this associative behavior are poorly understood, fishers have traditionally used this behavior to find and capture associated fish (Freon and Dagorn 2000). In the early 1980s, industrial tropical tuna purse-seine fleets began to commonly deploy their own FOBs, 37 called drifting Fish Aggregating Devices (DFADs, Dagorn et al. 2013b). In 38 the Indian Ocean (IO), deployment and use of DFADs began in the 1990s and 39 has steadily increased since then, such that from 2012 to 2018, DFADs were 40 demonstrated to represent more than 85% of the total floating objects in 41 the western IO (Dupaix et al. 2021).

42 Soon after their wide-scale use began, it was hypothesized that DFADs may 43 act as "ecological traps" for tropical tunas (see Figure 1) (Marsac et al. 2000; Hallier and Gaertner 2008). An ecological trap occurs when 44 45 individuals select poor-quality habitats, being misled by cues that no 46 longer correlate to habitat quality due to anthropogenic changes (Gilroy and Sutherland 2007). This decorrelation between habitat quality and 47 48 habitat selection cues ultimately leads to a reduction in the fitness of 49 individuals (Gilroy and Sutherland 2007; Swearer et al. 2021). The hypothesis of DFADs acting as ecological traps, as it was first formulated, 50 51relies on the assumption that tropical tunas and other associated species 52 originally used floating objects to select good-quality habitats (the 53 *indicator-log* hypothesis, see Freon and Dagorn 2000). The massive deployment of DFADs would modify the density and spatial distribution of 54 floating objects, with potentially large numbers of artificial objects 55 56 occurring in areas that are not optimal for tunas, creating the risk of an 57 ecological trap. Hence, there is an urgent need to assess the likelihood of DFADs acting as ecological traps. 58

59 A proxy which can be used to assess tuna fitness is physiological condition. Tunas caught at DFADs are considered to be in poorer condition 60 61 than those caught in free-swimming schools which infers a negative 62 biological consequence from the association with DFAD (Marsac et al. 2000; Hallier and Gaertner 2008). Robert et al. (2014) also found a difference 63 between the condition of associated and non-associated tunas, but in an 64 area (the Mozambique Channel, Indian Ocean) that was rich in natural 65 66 floating objects and thus only marginally modified by the addition of 67 DFADs. Hence, it is possible that the association with a floating object 68 results in a poorer condition, but that the evolutionary advantage of the associative behavior would not be related to short-term trophic benefits. 69 Tunas could recover faster after associating because they are in a more 70 71productive area or in larger schools. This led us to consider the 72 ecological trap hypothesis over a long period of time, to examine the 73 condition of tuna before and after the use of DFADs.

The objective of this study was to test the hypothesis that the body condition of yellowfin tuna has decreased since the wide-spread use of DFADs began in the Indian Ocean. We used length and weight measurements to calculate Le Cren's relative condition factor  $(K_n)$ , and investigated the temporal evolution of the body condition of yellowfin tuna (*Thunnus albacares*) from 1987 to 2019 in the Indian Ocean. We expected, that the 80 condition factor of yellowfin tuna would decrease during the study period 81 (Figure 1).

#### 82 Material and Methods

## 83 Biological data

A total of 25,914 yellowfin tuna (Thunnus albacares) were sampled from 1987 84 to 2019 in the Indian Ocean Tuna canning factory (IOT) in Victoria, 85 Seychelles (Guillou et al. 2021). All the sampled fish were caught by purse 86 seine vessels in the western IO (details of the sample sizes are provided 87 88 in Table S1). The total weight (W) of the individuals and their fork length 89 (FL) were measured. For each sampled tuna, the fishing vessel and the 90 fishing trip were recorded, but not the specific fishing set from which it 91 was caught, which meant that the exact fishing date and location were 92 unknown (see statistical analyses section for details on how the 93 uncertainty on location and date was managed). The year (Y) and quarter (Q) 94 of the catch of each tuna were estimated from the middle of the interval 95 covered by the fishing trip dates. The quarters were defined to be 96 synchronous with the general movement of the fleet, fishing seasons and areas (Dupaix et al. 2021): Q1, December to February; Q2, March to May; Q3, 97 98 June to August; and Q4, September to November. The total range of FL was

99 divided in three intervals, defining size classes (SC): small (<75cm), 100 medium (75-120cm) and large (>120cm).

#### 101 Relative condition factor

102 To calculate the theoretical weight of individuals ( $W_{th}$ ), FL and W measures 103 were used to estimate the parameters of the length-weight allometric 104 relationship, using the theoretical power-law equation:  $W_{th} = a FL^{b}$ . Details 105 on the fit of this power-law are presented in Supplementary Material 1. 106 Secondly, for each individual fish, the relative condition factor (Le Cren 107 1951) was calculated as follows:

$$K_n(i) = \frac{W(i)}{W_{th}(i)}$$

108 where  $W_{th}(i)$  is the theoretical weight of individual *i* calculated from 109 length-weight allometric relationship coefficients according to FL(i), and 110 W(i) is the measured total weight. By definition,  $K_n(i)$  measures the 111 deviation of an individual from the weight of a mean individual of the same 112 length. For the remainder of the text, the relative condition factor of one 113 individual is denoted as  $K_n(i)$ . The mean relative condition factor, denoted 114  $K_n$ , calculated for a group of individuals is denoted as  $K_n$ .

#### 115 Statistical analysis

116 In order to determine if  $K_n$  decreased with the concurrent increase in DFAD 117 numbers during the study period, a Generalized Additive Model (GAM) was 118 performed considering  $K_n(i)$  as the dependent variable, with a Gaussian link function to account for explanatory variables. Explanatory variables were 119 120 chosen to assess the effect of the fishing year (Y), season (fishing 121 quarter, Q), size of the individuals (size class, SC), and fishing location 122 (longitude, Lon; latitude, Lat). Longitude and latitude were included in 123 the model as a smoothed term, and other variables were considered as 124factors. No precise time-series of DFAD number exist in the IO over 1987-1252018, but the deployment of DFADs increased during that period, hence we 126 considered the fishing year as a proxy.

127 Because  $K_n$  is the ratio of two correlated random variables (Pearson's correlation coefficient between W and  $W_{th}$ , Pearson' s  $\rho = 0.99$ ), it did not 128 129 follow a normal distribution and displayed overdispersion. For this reason, 130 and because it did not change the interpretation of the GAM results, we transformed the  $K_n(i)$ using a Geary-Hinkley transformation before 131132performing the GAM (Geary 1930, see Supplementary Material 2). The 133Generalized Additive Model was performed on the transformed  $K_n(i)$ , noted 134 $T(K_n(i))$ . A complementary analysis showed that size class and its 135interaction with other explanatory variables did not impact the main results of the study (see Figure S3). 136

137 As the exact geographic coordinates were not available for most of the 138 sampled fish, a bootstrap process was applied: a dataset was generated by 139 sampling one set of coordinates from all the fishing sets of the trip for each individual and a GAM was then performed. This operation was repeated 141 1,000 times and for every model built, we selected the most parsimonious 142 explanatory variables based on the Akaike information criterion (AIC), 143 using a stepwise selection procedure and a threshold of 2. The iterated GAM 144 coefficients of the explanatory variables considered as factors (Y, Q and 145 SC) were averaged over the bootstrap replica and their standard deviation 146 calculated.

147 Results

## 148 Mean relative condition factors (K<sub>n</sub>)

The mean relative condition factor value  $(K_n)$  was  $1.01 \pm 0.088$  and mean annual  $K_n$  values varied between  $0.93 \pm 0.064$  (in 1987) and  $1.07 \pm 0.079$  (in 2012). The relative condition factor displayed annual variations, with low  $K_n$  values in 1987–1990 and around 2006, and the highest  $K_n$  values observed around 2012 (Figure 2A). The mean annual  $K_n$  displayed similar variations per size class as when all the sampled fish were considered together (Figure 2A). No clear trend in  $K_n$  variations were observed.

## 156 Yearly variations of K<sub>n</sub>

157 The most parsimonious model, selected using the AIC, included year (Y), 158 quarter (Q), size class (SC) and the smoothed term for longitude and 159 latitude. The selected model explained 29.2% of the deviance. The residuals 160 displayed no spatial autocorrelation and their distribution was not 161 significantly different from a Gaussian distribution (Figure S4). The GAM 162 performed on the transformed relative condition factor,  $T(K_n(i))$ , showed 163 that strongest  $T(K_n(i))$  variations were significantly correlated with fishing year (Figure 2B; Figure S5 & S6). The annual GAM coefficients 164 165 displayed a non-monotonous trend which was non-decreasing in time, with 1987 being the year with the lowest coefficient (-0.475  $\pm$  0.007) while 2012 166 167 was the year with the highest coefficient (0.673  $\pm$  0.006; Figure 2B). The 168 observed patterns were similar to those displayed when considering only the 169 mean annual  $K_n$  (Figures 2A&B).

## 170 Discussion

171The main hypothesis tested in this study was that the introduction of DFADs 172would have negatively impacted the condition of tunas, following roughly 173three decades of DFAD deployment in the ocean (Figure 1). Under the 174hypothesis that DFAD number increased during the study period, we expected 175a monotonous decrease of yellowfin tuna condition throughout the years. The 176relative condition factor  $(K_n(i))$  values obtained here did not display any 177clear temporal trend over the study period (Figure 2), which goes against 178 the tested hypothesis. Hence, the present study concludes that under the 179conditions encountered by yellowfin tuna in the Indian Ocean during the 180 last three decades, the addition of DFADs to the pelagic environment has 181 not led to the creation of an ecological trap for this species.

182 However, several precautions must be considered when interpreting this 183 result. Sampling was not uniformly distributed across size classes and 184 years, which could have influenced the results (Table S1 & Figure S7). 185 However, Wang et al. (2002) found no relationship between the size and the 186 condition factor of yellowfin and bigeye tunas. Also, when performing a GAM 187 on data of each size class independently, no decreasing trend of 188 physiological condition was observed (Figure S3). Therefore, we are 189 confident that the methodology used here did not suffer from any bias 190 induced through the uneven sampling.

191 Also, morphometric indices, such as K<sub>n</sub> used here, should be used with 192 caution as they do not always reflect the energy compounds in the tissues 193 (Lloret et al. 2014). Hence, further efforts are needed to validate them 194 against proper benchmarks (Lloret et al. 2014), which could be achieved studies. 195 through the development of experimental However. these 196 morphometric indices can be informative (Hallier and Gaertner 2008), and 197 are often the only available metrics as it was the case in this study (no long-term time-series of biochemical indices exist for tuna). 198

199 Furthermore, the habitat change induced by DFADs could also impact other 200 biological processes, such as natural mortality, growth rate (Hallier and 201 Gaertner 2008), or reproduction (Zudaire et al. 2014), that could not be 202 assessed through our biological index. For example, if the growth rate of 203 tunas is negatively altered by DFADs (see Hallier and Gaertner 2008), a 204 tuna of a given size in the 1980s could be younger than a tuna of the same 205 size in the 2010-2020s, with the two individuals having a similar 206 physiological condition.

207 Our study is based on the assumption that, since the late 1980s, the number 208 of DFADs increased in the western IO. While assessing a time-series of the exact density of DFADs is cumbersome, studies suggest that the number of 209 210 DFADs in the Indian Ocean continued to increase throughout the study period 211 (Dupaix et al. 2021). The conditions leading to an ecological trap might 212 also have only been reached in recent years, without being detectable in 213 our study. This highlights the need to keep developing long-term monitoring 214programs, collecting data on DFAD densities and standardized physiological 215indices, to study the impact of DFADs on tuna physiology.

216 As stated above, the study is valid only for the conditions experienced by 217 yellowfin tuna in the western IO in the last three decades. Aside from the 218 increase of DFAD density, other long-term phenomena could also impact the 219 physiology of tropical tuna. Since the 1980s, climate change has already 220 impacted tuna habitat, by inducing changes in sea surface temperature or 221 oxygen concentration, for example (Erauskin-Extramiana et al. 2019). Erauskin-Extramiana et al. (2019) projected that yellowfin tuna will 222 223 become more abundant under a climate change scenario (the authors used the

RCP8.5 and expect the same projected changes of abundance under other scenarios until 2050). Hence, a potential negative impact of increasing DFAD density may have been concealed by other phenomena like climate change.

228 Ecological traps are likely to become more common as human-induced 229 environmental changes increase, and they have important implications for 230 the management of animal populations (Swearer et al. 2021). The hypothesis that DFADs could act as ecological traps was developed over 20 years ago 231232(Marsac et al. 2000). It suggests that, by modifying the density and 233 spatial distribution of FOBs, the massive deployment of DFADs could retain 234 or transport individuals to areas that are ecologically unsuitable for them 235(Marsac et al. 2000). By demonstrating the absence of any decreasing trend 236 in yellowfin tuna condition concurrent to the observed increasing DFAD 237 density in the Indian Ocean during the past three decades, this study 238 rejects the ecological trap hypothesis as it was originally formulated. 239 However, DFADs can impact the biology of tuna in a variety of complex ways, 240 and this impact can be intertwined with the effects of other factors such 241 as climate change. Therefore, it is necessary to establish long-term 242monitoring programs to track (i) habitat changes, (ii) temporal variations 243 of behavioral features in tuna, such as large-scale movements and schooling behavior, and (iii), temporal variations of physiological parameters in 244 245tuna.

## 246 Declaration of Competing Interest

247 The authors declare that they have no known competing financial interests 248 or personal relationships that could have appeared to influence the work 249 reported in this paper.

#### 250 Data availability statement

- 251 Data used in this study is available in Guillou et al. (2021).
- 252 All analyses were performed using R software v. 4. 0. 3, and the scripts used
- 253 for the study are available on GitHub:
- 254 https://github.com/adupaix/Historical\_YFT\_condition
- 255 https://doi.org/10.5281/zenodo.6123417
- 256 A preprint version of the article is available at:
- 257 https://hal.archives-ouvertes.fr/hal-03690665

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## 265 Figures

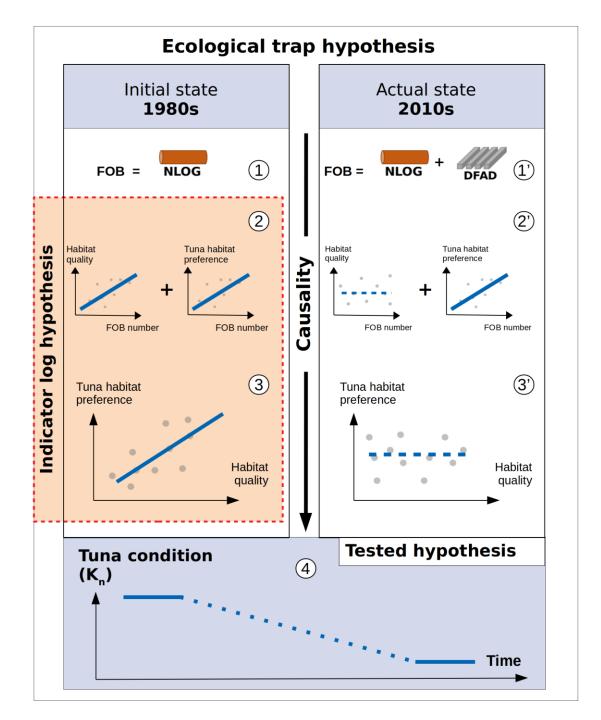


Figure 1: Schematic representation of the ecological trap hypothesis applied to Fish Aggregating Devices and tropical tuna. FOB: Floating object of any kind; DFAD: Fish Aggregating Device; NLOG: Natural floating object. Under this hypothesis, before DFAD introduction, when only NLOGs 270were present (1), floating objects were distributed in productive areas 271(2), hence tunas, which associate with floating objects, preferred high 272 quality habitats (3). Since DFAD introduction (1'), the distribution of 273 floating objects has been modified and is no longer correlated with habitat 274quality (2'). Hence, tunas, which still associate with floating objects, do not select high quality habitat anymore (3'). As a consequence of this 275 276 habitat modification, the physiological condition of tunas would have decreased since the 1980s (4). Preference is defined here as the likelihood 277278 of a resource being chosen if offered as an option with other available 279 options.

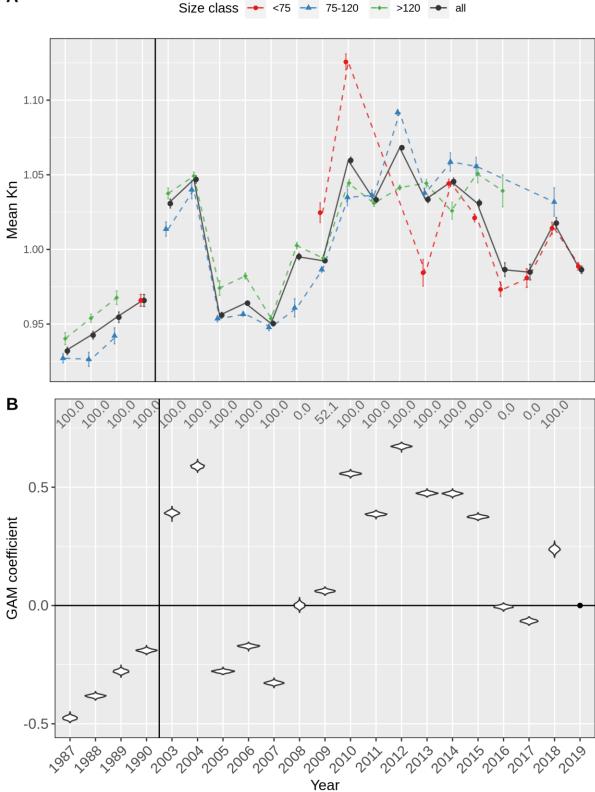


Figure 2: No observed trend in yellowfin tuna condition: (A) Mean relative condition factor per year. The K<sub>n</sub> is represented for all individuals (all, black circles), for small individuals (<75, red circles), medium-size</p>

283 individuals (75-120, blue triangles) and large individuals (>120, green 284 diamonds). Values are represented only when more than 50 individuals of the 285 given class were measured. Error bars represent the standard error of the 286 (B) Coefficients of the fishing year in the Generalized Additive mean. Model. Each coefficient represents the mean deviation of  $T(K_n)$  from the 287 values for a year of reference (2019, represented by a black dot). The 288 289 shape of the points represents the distribution of the values obtained with 290 the bootstrap process. Numbers in grey in the upper part of the panels 291 represent the percentage of the models generated in the bootstrap for which 292 a given category was significantly different from the category of 293 reference.