Fish aggregating devices drift like oceanographic drifters in the near-surface currents of the Atlantic and Indian Oceans

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Abstract

Knowledge of ocean surface dynamics is crucial for oceanographic and climate research. The satellite-tracked movements of hundreds of drifters deployed by research and voluntary observing vessels provide high-frequency and high-resolution information on near-surface currents around the globe. Consequently, they constitute a major component of the Global Ocean Observing System (GOOS). However, maintaining this array is costly and in some oceanic regions such as the tropics, spatio-temporal coverage is limited. Here, we demonstrate that the GPS-buoy equipped fish aggregating devices (FADs) used in tropical tuna fisheries to increase fish catchability are also capable of providing comparable near-surface current information. We analyzed millions of position data collected between 2008 and 2014 from more than 15,000 FADs and 2,000 drifters, and combined this information with remotely-sensed near-surface current data to demonstrate that the surface velocity components of FADs and drifters are highly correlated in the Atlantic and Indian Oceans. While it was noted that the subsurface structures of FADs (typically made of recycled fishing nets) did slow them down

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relative to the drifters, particularly in the Atlantic Ocean, this bias was measurable and could be accounted for in future studies. Our findings show that the physical meteorological and oceanographic data collected by fishermen could provide an invaluable source of information to the GOOS. Furthermore, by forging closer collaborations with the fishing industry and ensuring their contributions to global ocean databases are properly acknowledged, there is significant scope to capture this data more effectively.

Keywords: drifter, fisheries, Lagrangian transport, oceanography, surface currents

1 Introduction

Oceans cover 70% of the Earth's surface and are much harder to observe 2 than terrestrial systems (Richardson and Poloczanska, 2008). For centuries, 3 mariners have been observing the states of oceans and the atmosphere by recording oceanographic and physical meteorological data near the ocean's 5 surface (Woodruff et al., 1987). As early as the nineteenth century, inter-6 national collaborative efforts were initiated to coordinate the collection and curation of ocean-atmosphere data from voluntary observing ships (VOS) 8 and build large-scale marine data sets. Such data sets are now considered 9 essential for oceanographic and climate studies (Woodruff et al., 1987; Kent 10 et al., 2010; Freeman et al., 2017). From the 1970s, ocean data collection was 11 revolutionized with the advent of satellite technology and the development 12 of sensors that were capable of measuring a large range of oceanographic and 13 atmospheric features (Martin, 2004). 14

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Combining *in-situ* and remotely-sensed satellite observations has proven 16 to be an essential step to improving our understanding of how ocean circu-17 lation affects climate at regional and global scales through the transport of 18 water and heat received from the sun (Maximenko et al., 2009; Lee et al., 19 2010). Remotely-sensed measurements of sea surface temperature, altimetry 20 and vector winds provide a synoptic view of ocean surface current patterns 21 at consistent and regular spatial and temporal scales (Lagerloef et al., 1999; 22 Sudre and Morrow, 2008; Dohan and Maximenko, 2010). At a finer scale, in-23 situ velocity measurements of near-surface currents are routinely collected by 24 satellite-tracked drifters maintained by the Global Drifter Program (GDP), 25 an operational component of the Global Ocean Observing System (GOOS) 26

and the Global Climate Observing System (GCOS). This data provides a 27 direct measurement of water properties and complements the satellite data 28 by supplying information on high-frequency, small-scale oceanic processes 29 (Niiler and Paduan, 1995; Reverdin et al., 2003; Lumpkin and Elipot, 2010). 30 These drifters are floating devices that comprise a surface buoy equipped 31 with a satellite transmitter and a subsurface sea anchor (Fig. 1). Since 32 the 2010s, the GDP has maintained a global array of $\sim 1,200-1,500$ drifters 33 that have been deployed from VOS, research vessels and planes to cover the 34 world's oceans (Joseph, 2013; Lumpkin and Johnson, 2013; Elipot et al., 35 2016). In addition to supporting oceanographic and climate research, the 36 ocean circulation information acquired by these systems has been instru-37 mental in supporting both military and civil applications, including search 38 and rescue operations that use the data to improve their field of search pre-39 dictions (Davidson et al., 2009). More recently, their role in tracking floating 40 debris (Law et al., 2010; Cózar et al., 2014) has garnered attention as con-41 cerns about marine plastics pollution increase. 42

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A knowledge of ocean dynamics is also key for fishermen who use it to 44 both navigate and find fish resources. Monitoring surface water characteris-45 tics is essential in pelagic fisheries where the use of satellite remote-sensing 46 has long been recognized as a fish harvesting aid (Simpson, 1992; Chassot 47 et al., 2011). Modern fishing vessels are now equipped with a large range of 48 sensors and electronic tools that constantly monitor the marine environment, 49 enabling fishermen to identify the suitable habitats of target fish species (e.g. 50 Torres-Irineo et al., 2014). In tuna fisheries, the purse seine vessels that 51 target fish schools have extensively deployed satellite-tracked fish aggregat-52 ing devices (FADs) over the last decade. Typically made of a bamboo raft 53 equipped with floats (to ensure buoyancy) and a sea anchor built of old fish-54 ing nets (Fig. 1), these FADs attract tuna and increase fishery productivity 55 (Fonteneau et al., 2013; Maufroy et al., 2017). In recent years, the num-56 ber of GPS-buoy equipped FADs used globally in this fishery has increased 57 markedly. Currently, it is estimated that more than 100,000 FADs are now 58 drifting around the globe at any given time (Baske et al., 2012; Scott and 50 Lopez, 2014). While the average lifespan of a FAD at sea is shorter than a 60 typical drifter, there are many more in circulation, particularly in the trop-61 ical areas where the purse seine fleets operate. Consequently, it is likely 62 that FADs could provide the GDP with complementary data, particularly 63 in equatorial regions. Given that these areas are currently under sampled 64

due to factors such as infrequent deployment of drifters and equatorial divergence (Lumpkin and Pazos, 2007), this increased FAD data coverage is especially important. As an illustrative case, a few FAD positions were used to complement the drifter data and ocean model outputs analyzed to locate the wreckage of the Air France flight that crashed in 2009 en route from Rio de Janeiro to Paris (Drévillon et al., 2013).

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The overarching objective of this study is to test the hypothesis that FADs 72 deployed by fishermen follow near-surface currents in the Atlantic and Indian 73 Oceans. If true, this would mean that FAD data could provide valuable in-74 formation on oceanic circulation. To test this hypothesis, we combined and 75 analyzed large data sets from GDP drifters, a satellite-derived surface cur-76 rent product available from the Ocean Surface Currents Analyses Real-time 77 (OSCAR) processing system and approximately 5 million FAD positions col-78 lected by French tuna fishing companies between 2008 and 2014. To begin 79 with, we directly compared the velocities of FAD and drifter pairs observed 80 in close proximity over similar time periods. We then used the OSCAR cur-81 rents as an indirect comparison point for both the FAD and drifter data. 82 For the large biogeographical provinces of the Atlantic and Indian Oceans 83 (Longhurst, 2007), we estimated the correlations between the OSCAR cur-84 rents and the observed FAD and drifter velocities. We then compared FAD 85 and drifter movements with short-term OSCAR current projections. 86 87

88 Material ans methods

⁸⁹ Fish Aggregating Devices

The GPS locations of the buoys attached to the FADs used by the French 90 fishing fleet operating in the Atlantic and Indian Oceans have been avail-91 able since 2008 through a collaborative agreement between the Institut de 92 Recherche pour le Développement (IRD) and the French frozen tuna pro-93 ducers' organization ORTHONGEL. The full methodology used to filter and 94 process the raw GPS data to derive FAD trajectories at sea can be found 95 in Maufroy et al. (2015). The current FAD data set consists of 4,777,524 96 positions, belonging to a total of 21,047 distinct buoys that were deployed at 97 sea between 2008 and 2014. The periodicity of FAD position varies from 15 98 minutes (minimum) to 2 days (maximum). This function can be remotely 99 modified to facilitate detection when a vessel is on its final approach to a 100



Figure 1: Description of the structure and design (in the water column) of a typical drifter (left), fish aggregating devices (FADs) used in purse seine fisheries with a sea anchor made of 'curtain' nets (middle), and 'sausage' nets (right).

FAD. Approximately 20% of the FAD data set consists of contiguous lo-101 cations emitted within a time period of less than 6 hours and most FADs 102 emitted two contiguous signals within a 24-hour period. FADs generally 103 consist of bamboo rafts covered in old pieces of purse seine nets with several 104 floats that ensure their buoyancy (Fig. 1). The subsurface structures found 105 below FADs are typically made out of old fishing nets and extend to depths 106 of 30-80 m (Franco et al., 2009). Initially, these nets hung in 'curtains', but 107 newer designs feature 'sausages' of nets or ropes. This design was introduced 108 to prevent the entanglement of fish and turtles. French GPS buoys have 109 also been deployed on floating objects of natural (e.g. palm trees, logs) or 110 anthropogenic (e.g. ropes) origins that represented about 20% of all floating 111 objects encountered at sea by observers on French purse seiners during 2008-112 2014, with the Mozambique Channel being characterized by a relatively high 113 percentage of these natural objects (Maufroy et al., 2017). 114 115

116 Surface drifters

The drifters are made up of a surface buoy (~ 30 cm diameter) that is 117 attached by a long, thin tether to a holey sock drogue (sea anchor) that is 118 centered at ~ 15 m below the surface (Fig. 1). The buoy measures sea sur-119 face temperature and other properties such as air pressure and wind direction 120 and sends this information to passing satellites using an ARGOS transmitter 121 (Lumpkin and Pazos, 2007). While the size of the buoy and drogue can vary, 122 their drag area ratio is standardized, which acts to constrain their downwind 123 slip (Niiler and Paduan, 1995). The GDP archives most of the data collected 124 by the drifters. We downloaded our data set (1,092,362) positions belong-125 ing to 2,285 distinct, drogued drifters having occurred in the Indian and 126 Atlantic Oceans during 2008-2014) from ftp://ftp.aoml.noaa.gov/phod/ 127 pub/buoydata/. Hansen and Poulain (1996) detail the corrections that are 128 applied to the raw data. 129

131 Data filtering

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¹³² A very small number of velocity values collated from the IRD and GDP ¹³³ databases were found to be inconsistent with the maximum speed expected ¹³⁴ for ocean currents. We therefore removed data points that had velocity val-¹³⁵ ues higher than the 99.99% quantile values of 471.6 cm s⁻¹ (i.e. 9.17 knots) ¹³⁶ and 234 cm s⁻¹ (i.e. 4.55 knots) for fish aggregating devices (FADs) and ¹³⁷ drifters, respectively.

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139 Data distribution

There is twice as much data for the Indian Ocean as the Atlantic Ocean 140 but overall, the number of FAD locations has increased markedly in both 141 oceans over the study period while the amount of drifter data remained rela-142 tively constant (Table 1). This reflects the significant expansion in the FAD 143 fishery that has taken place in both regions (Maufroy et al., 2017). In this 144 study, we focused on eight large biogeographical provinces, four of which oc-145 curred in the Atlantic Ocean (i.e. Guinea Current Coastal (GUIN), Eastern 146 Tropical (ETRA), North Atlantic Tropical (NATR), and Western Tropical 147 Atlantic (WTRA)) and four of which occurred in the Indian Ocean (East 148 Africa Coastal (EAFR), North West Arabian Upwelling (ARAB), Indian 140 Monsoon Gyres (MONS), and Indian Southern Subtropical Gyre (ISSG)) 150 (Longhurst, 2007). The total number of FAD data points collated for these 151 provinces was >50,000 (Appendix Table A1). 152

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Table 1: Annual number of fish aggregating device (FAD) and drifter observations analyzed in the Atlantic and Indian Ocean.

Device	ocean	2008	2009	2010	2011	2012	2013	2014
FADs	Atlantic	$17,\!849$	45,469	102,216	$153,\!990$	$286,\!156$	$322,\!490$	464,930
FADs	Indian	$105,\!356$	149,211	200,983	382,315	$580,\!547$	$784,\!130$	$1,\!181,\!882$
Drifters	Atlantic	$93,\!540$	108,828	84,912	65,851	$75,\!974$	$118,\!125$	87,067
Drifters	Indian	$51,\!479$	38,311	48,303	$46,\!603$	$52,\!377$	80,777	$140,\!215$

154 Satellite currents

The satellite-derived surface current information produced by the OS-155 CAR processing system is provided in near-real time from a combination of 156 quasi-steady geostrophic and locally wind-driven dynamics (Lagerloef et al., 157 1999) (http://www.oscar.noaa.gov). The OSCAR product combines: (i) 158 a geostrophic term computed from the gradient of ocean surface topogra-150 phy fields using several sources of spatial observation through time, (ii) a 160 wind-driven velocity term computed from an Ekman-Stommel formulation 161 with variable eddy viscosity using QuikSCAT and National Centers for En-162 vironmental Prediction winds, and (iii) a thermal wind adjustment using 163 Reynolds sea surface temperature (Reynolds and Rayner, 2002). Dohan and 164

Maximenko (2010) provide a full description of the OSCAR product. In this
study, we used the 1/3 degree grid and 5-day interval resolution of the OSCAR currents, which is designed to represent a 30 m surface layer average.
The OSCAR currents have been validated with moored buoys, drifters, and
shipboard acoustic Doppler current profilers (Johnson et al., 2007).

171 Direct comparison

To compare possible velocity differences between the floating devices, we 172 selected every FAD and drifter pair that emitted a signal in near space and 173 time. Thus, for each FAD location and 24-hour time period, we searched 174 for a drifter within a 1/6 degree radius (~10 nm). If several drifters were 175 identified, we selected the device that was closest in time. A sensitivity 176 analysis, with time periods of 12 hours and 2.5 days (consistent with the 177 OSCAR temporal resolution), was then conducted. The correlation between 178 the corresponding zonal and meridional velocity components for the FAD and 179 drifter pairs was then considered using the Pearson's correlation coefficient 180 (Johnson et al., 2007). We then used major axis regression models (forced 181 through the origin) to assess the agreement between the two variables (Leg-182 endre and Legendre, 1998; Warton et al., 2006). This approach accounts for 183 the measurement errors in both variables. 184

185

186 Indirect comparison

This comparative analysis was then extended to the full data set by un-187 dertaking an indirect comparison of FAD and drifter velocities using satellite 188 measurements of near-surface current velocities. At each FAD and drifter po-189 sition, we linearly interpolated the OSCAR current data in time and space to 190 calculate the OSCAR velocities (Johnson et al., 2007; Dohan and Maximenko, 191 2010). To determine the correlation and agreement between the FADs and 192 OSCAR and drifters and OSCAR, we used the methodology described in the 193 previous section. This analysis was completed at both the basin and large 194 biogeographical province (Longhurst, 2007) scales to ensure that the different 195 oceanographic regimes of the Indian and Atlantic Oceans were represented. 196 The spatio-temporal autocorrelation of velocity values along the FAD and 197 drifter trajectories was accounted for by subsampling the data at values that 198 were close (5 days) and far above (15 days) the Lagrangian integral time 199 scale estimated for drifters in the Indian Ocean (i.e. 2-7 days; Peng et al., 200

201 2014). 202

²⁰³ Projection of FAD and drifter locations using OSCAR

The OSCAR velocities were then used to project the FAD and drifter lo-204 cations from one timestep to the next to compare their Lagrangian transport 205 in near-surface waters. We computed the distance d between the projected 206 location and the next observed location and the distance D between the 207 current location and next observed location to estimate the index d/D for 208 FADs and drifters. These indices were used to gauge the degree of departure 209 of each floating device from the OSCAR currents predictions (Berta et al., 210 2014; Yaremchuk et al., 2016). Index values that were larger than the 99%211 quantile values observed in the data sets were removed and then their dis-212 tributions were compared (i.e., for FADs and drifters) at both the basin and 213 (selected) large biogeographical province scales. 214 215

216 **Results**

At the basin scale, the velocity distributions of FADs and drifters were 217 different. In the Atlantic Ocean, the first quartile, median, and third quartile 218 values in the FAD and drifter velocity distributions were 11.45, 19.96, 32.8 cm 219 s^{-1} , and 9.15, 15.11, 24.36 cm s^{-1} , respectively. The velocities of both device 220 types were found to be higher in the Indian Ocean where these values were 221 21.58, 35.13, 54 cm s⁻¹ for FADs and 14.78, 24.5, 38.44 cm s⁻¹ for drifters. 222 At a regional scale, FAD and drifter velocities were similar in the ETRA, 223 NATR, and WTRA biogeographical provinces of the inter-tropical Atlantic 224 Ocean, but they differed in the GUIN province (Appendix Table A2). In that 225 province, the number of drifter locations was the lowest, more than an order 226 of magnitude lower than the number of FAD locations (Appendix Table A1). 227 Within the four provinces that make up most of the south-western Indian 228 Ocean, FAD velocities were substantially higher than drifter velocities (Ap-220 pendix Table A2). Differences in velocities between FADs and drifters were 230 attributed to differences in the spatio-temporal distribution between the two 231 types of devices. In the Atlantic Ocean, the FAD data were concentrated in 232 the central-eastern region (Fig. 2A) while the drifter data were more evenly 233 distributed, although the northern area showed the highest concentrations 234 (Fig. 2B). In the Indian Ocean, the FAD data were concentrated in the 235

central-western region (Fig. 2A) while the drifter data were more evenly 236 distributed over the entire basin (Fig. 2B). At a smaller, $1^{\circ} \times 1^{\circ}$ spatial 237 scale, the FADs and drifters showed very similar patterns of velocity in the 238 near-surface currents (Fig. 2C and D), revealing the major oceanographic 239 features of both the tropical Atlantic Ocean (the South Equatorial and the 240 North Brazil currents, the Equatorial countercurrent and the Guinea current) 241 and the Indian Ocean (Somali, North Madagascar, and Agulhas currents, the 242 Equatorial countercurrent and the South Equatorial current). 243

244

More than 18,000 pairs of FADs and drifters were detected across the 245 Atlantic (n = 4,146) and Indian (n = 14,556) Oceans (Fig. 3). For these 246 pairs, the zonal and meridional components of the FAD vs. drifter velocities 247 were found to be significantly and highly correlated with Pearson's correla-248 tion coefficients between 0.68 and 0.93 (Fig. 4). This result was found to be 240 robust to the time period considered for the definition of pairs of floating de-250 vices (Appendix Table A3). We also found several pairs in both oceans that 251 shared common trajectories over several weeks to months, e.g., two FADs 252 deployed in the Indian Ocean in 2013 traveled with two drifters during sev-253 eral months (Appendix Fig. A1). In the Indian Ocean, the velocity of FAD 254 and drifter pairs agreed remarkably well (Fig. 4 and Appendix Table A3). 255 In the Atlantic Ocean, however, small but consistent systematic differences 256 in the velocity components indicate that drifters move faster than FADs (2-257 37% higher velocity components, 10-21% higher overall velocity; Fig. 4 and 258 Appendix Table A3). 259

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The outcomes of the comparative analysis of FAD and drifter velocities 261 with OSCAR satellite current products further supports the case for using 262 FADs for monitoring ocean surface dynamics. Here, the regional spatial pat-263 terns in both FAD and drifter velocities were consistent with the remotely-264 sensed surface currents. The ocean currents inferred from the FAD and drifter 265 movement data for near-surface waters in both oceans were correlated in the 266 same way, and to the same extent, with the satellite currents in the large bio-267 geographical provinces. The correlation coefficients of velocity components 268 between FADs and OSCAR and drifters and OSCAR were generally very 269 similar (Fig. 5, Appendix Figs. A2-A5 and Appendix Table A4). After ac-270 counting for autocorrelation in the data, these relationships were still highly 271 significant (Appendix Table A5). However, the OSCAR currents appeared 272 to be slower than the currents derived from the *in-situ* data collected from 273



Figure 2: Spatial distribution of fish aggregating devices (FADs; A) and drifters (B) in the Atlantic and Indian Oceans within the defined geographical limits 60°W, 100°E longitude and 40°S, 30°N latitude. Density corresponds to the number of location points observed in each 1° × 1° grid cell for the time period 2008-2014. Mean of near-surface ocean currents (m s⁻¹) for the period 2008-2014, derived from FAD (C) and drifter (D) movements. Solid lines indicate boundaries between biogeographical provinces (Longhurst, 2007) (see Appendix Table A1 for acronyms).



Figure 3: Spatial distribution of FADs (red triangles) and drifters (blue crosses) pairs that occurred within a 10 nm radius during 24-hour periods in the Atlantic (n = 4,146) and Indian (n = 14,556) Oceans.

the floating devices, as indicated by the slopes of the relationships between the OSCAR currents and floating devices being lower than 1 in all cases but one (Appendix Fig. A6 and Appendix Table A4). At the biogeographical province scale, the large variability observed in these slopes (FADs: 0.2-0.9 and drifters: 0.4-1.2) shows that they are not representing the surface dynamics at the same spatio-temporal scale.

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The distributions of the OSCAR-projection error index d/D for FADs and drifters were almost identical across all biogeographical provinces (Appendix Fig. A7), with the notable exception in the south subtropical gyre province of the Indian Ocean (ISSG). Differences in spatial coverage explain this result, with FADs mostly occurring in the North of the ISSG province during the 2008-2014 period while drifters spanned the whole area (Fig. 2).

288 Discussion

We combined large data sets of remotely-sensed current speed with the GPS positions of thousands of satellite-tracked floating devices to show that the fish aggregating devices used in tuna fisheries and oceanographic drifters



Figure 4: Velocity comparisons between the FAD and drifter pairs (A) zonal component in the Atlantic Ocean; (B) meridional component in the Atlantic Ocean; (C) zonal component in the Indian Ocean; and (D) meridional component in the Indian Ocean. The solid line indicates the major axis regression model and the dashed line indicates the 1:1 isoline.



Figure 5: The comparison of correlation coefficients for the (A) zonal and (B) meridional components of velocity for the Ocean Surface Currents Analyses Real-time (OSCAR) versus fish aggregating devices (FADs) and OSCAR versus drifters in the selected Longhurst biogeographical provinces (see Appendix Table A1 for acronyms of the provinces and Fig. 2 for their location).

move similarly in near-surface ocean currents. This confirms that in tropical areas, the oceanographic information acquired by tuna fishermen could complement that gathered by the Global Ocean Observing System's drifter program. However, we highlighted some differences in the behaviour of FADs and drifters.

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While drifters are anchored at around 15 m below the surface, the FADs 298 subsurface structure composed of curtain or sausage nets can go down to 299 50-60 m in the Indian Ocean, 80 m in the Atlantic Ocean. These differences 300 in anchoring depth between the two types of floating devices, and between 301 FADs, locate them in different current layers. Indeed, we noted some speed 302 differences between the two types of floating devices, particularly in the At-303 lantic Ocean, which are likely related to differences in their drogue structures. 304 In the absence of strong winds, the geostrophic balance dominates the up-305 per ocean circulation. In this case, floating devices with different windage 306 and drogues at different depths, or even without drogue, move at similar 307 velocities. Conversely, higher and variable winds generate internal waves as 308 well as Ekman currents. The former modify the mixed layer depth whereas 309 the latter generate currents that quickly rotate with depth. In both cases, 310 floating devices with different drogue lengths will move with different ve-311 locities and often in different directions (Poulain et al., 2009). At smaller 312 scales, non-linear dynamics arising from wind-vorticity generate convergence 313 and divergence regions where floating devices drogued at various depth will 314 respond in different ways. 315

316

In the Indian Ocean, velocities in the Equatorial countercurrent where 317 many FADs occur have however been found to be relatively homogeneous 318 along a 0-60 m depth range (Gnanaseelan and Deshpande, 2017), showing 319 the same reversal pattern during monsoon periods. Depth homogeneous ve-320 locities were also reported along two modelled transects North of Madagascar 321 and off Tanzania (Manyilizu et al., 2016), within two areas of high FAD oc-322 currence. These conditions likely explain why velocities of drifter and FAD 323 pairs compare so well in the Indian Ocean, despite differences in their de-324 sign. By contrast, the eastern equatorial Atlantic Ocean is characterised 325 by the prominence of the Equatorial Undercurrent (EUC), a strong perma-326 nent eastward flow located just below the westward South Equatorial current 327 (Johns et al., 2014). FADs built and deployed in the Atlantic Ocean have 328 tails going down to 80 m, longer than in the other oceans (Franco et al., 329

2009), and at a depth where the core of the EUC is found along the equator
(Johns et al., 2014). These deep tails likely slow down the drift of the FADs
as compared to the shallow subsurface structure of the drifters, explaining
our results.

334

More generally and although the mechanisms of associative behavior of 335 tuna to FADs remain poorly understood (Fréon and Dagorn, 2000), tuna 336 fishermen consider that deeper tails increase the attraction of tunas by slow-337 ing down the FADs (Franco et al., 2009). Consequently, the depth of FAD 338 appendages has been increasing in recent years in all oceans (Murua et al., 339 2018). In the eastern Pacific Ocean for instance, data collected by observers 340 showed a substantial deepening of the net webbing from a median depth <10341 m in the early 1990s to about 30 m nowadays (Hall and Roman, 2017). How-342 ever, the progressive adoption of sausage nets and ropes in place of curtain 343 nets, aimed at reducing the entanglement of marine species, may incidentally 344 decrease the anchoring effect of the FAD tail appendage. In this study, our 345 data came from fishing companies that use very similar FAD designs made 346 of bamboo rafts and recycled fishing nets of similar lengths. More broadly, 347 information on the structural design of FADs and their components is now 348 being systematically collected through the fisheries observer programs run in 349 both oceans. This new information will be useful to determine the influence 350 of the subsurface currents on FAD drift. A comparison of the dispersion and 351 separation of concurrently deployed drifter and FAD clusters would also pro-352 vide insight into the extent to which design explains the observed differences 353 in speed between the two types of floating devices. 354

355

Given that the FAD data we used in this study is open access, we expect 356 that further analysis will be undertaken to fully validate the potential ap-357 plications of FAD data for oceanographers, and that the results of this work 358 will prompt long-term collaborations with the tuna fishing industry. The 359 quantity of information available to the scientific community would strongly 360 benefit from the release of data from other purse seine fishing companies op-361 erating in the Atlantic and Indian Oceans since the French purse seine fleet 362 only represented about 20% of the total purse seine catch in recent years. 363 Recent availability of FAD GPS positions in the western and central Pacific 364 Ocean shows a positive step in this direction (Escalle et al., 2017). It would 365 also be beneficial to apply the GDP's quality control procedures (Hansen and 366 Poulain, 1996; Lumpkin and Pazos, 2007) to the FAD data. This step may 367

provide useful information that is currently missing such as FAD location errors.

370

More broadly, the conspicuous character of global changes presents some 371 serious observational challenges. Effectively responding to these challenges 372 requires better integration across individual networks and multiple platforms. 373 to make the most of synergies between the different types of ocean observa-374 tions (Roemmich et al., 2010). The development of standards for metadata 375 and data formats, as well as access protocols (e.g., Web Services), has re-376 cently enhanced interoperability functions in information systems. Thus, 377 these standards are better able to merge and process heterogeneous data 378 sets stored in distributed infrastructures and promote integration across sci-379 entific disciplines (Reichman et al., 2011; Mooney et al., 2013; Robertson 380 et al., 2014). Data management systems should also include well-described 381 control procedures that aim to inform users about the best quality data sets 382 available (Roemmich et al., 2010). In oceanography, the recent introduction 383 of key standards contributes to this higher level of interoperability for phys-384 ical and chemical parameters delivered as gridded data (e.g. model outputs, 385 or satellite remote-sensing products) or time series of parameters retrieved 386 from platforms at sea (Hankin et al., 2010). Like the data collected through 387 citizen science initiatives (Lauro et al., 2014), the millions of data collected by 388 fishermen could substantially increase the spatio-temporal coverage of ocean 380 observations in a cost-efficient manner. Thus, the major contributions these 390 data sets could potentially make to the GOOS and GCOS calls for improved 391 collaboration with the fishing industry and the establishment of a system 392 that adequately acknowledges the contributors and fosters a data sharing 393 environment. 394

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416 Appendix

Province description	Code	Drifters	FADs
Australia-Indonesia Coastal Province	AUSW	$16,\!651$	6,106
Benguela Current Coastal Province	BENG	$6,\!431$	1,858
Brazil Current Coastal Province	BRAZ	$19,\!616$	1,399
Canary Coastal Province	CNRY	$13,\!085$	$21,\!399$
China Sea Coastal Province	CHIN	4,126	0
E. Africa Coastal Province	EAFR	$31,\!118$	175,733
E. India Coastal Province	INDE	10,248	237
Guianas Coastal Province	GUIA	$12,\!693$	8,946
Guinea Current Coastal Province	GUIN	8,009	234,069
NW Arabian Upwelling Province	ARAB	$23,\!400$	$367,\!690$
Red Sea, Persian Gulf Province	REDS	39	10
Sunda-Arafura Shelves Province	SUND	$6,\!872$	83
SW Atlantic Shelves Province	FKLD	189	0
W. India Coastal Province	INDW	$5,\!494$	2,401
Archipelagic Deep Basins Province	ARCH	$27,\!079$	554
Caribbean Province	CARB	2,515	174
Eastern Tropical Atlantic Province	ETRA	59,360	$780,\!874$
Indian Monsoon Gyres Province	MONS	146,705	$2,\!665,\!216$
Indian S. Subtropical Gyre Province	ISSG	$160,\!585$	162,963
N. Atlantic Tropical Gyral Province	NATR	$184,\!482$	63,101
South Atlantic Gyral Province	SATL	260, 199	47,029
Western Tropical Atlantic Province	WTRA	$57,\!414$	$230,\!438$
S. Subtropical Convergence Province	SSTC	$34,\!404$	805
Subantarctic Province	SANT	$1,\!603$	27
N. Atlantic Subtropical Gyral Province (East)	NASE	0	87

Table A1: Total number of fish aggregating device (FAD) and drifter observations collected in the Longhurst biogeographical provinces between 2008 and 2014. Selected provinces are shaded.

Table A2: The first quartile, median, and third quartile values (cm s⁻¹) from the fish aggregating device (FAD) and drifter velocity distributions in selected Longhurst biogeographical provinces of the Atlantic (upper part of the table) and Indian (lower part) Oceans (see Appendix Table A1 for acronyms of the provinces and Fig. 2 for their location).

Device	Province	1st quartile	Median	3rd quartile
Drifters	ETRA	11.62	19.65	31.45
FADs	ETRA	12.08	20.29	31.91
Drifters	GUIN	13.07	23.05	38.60
FADs	GUIN	8.91	15.98	28.49
Drifters	NATR	8.36	13.39	19.95
FADs	NATR	9.03	15.09	24.42
Drifters	WTRA	15.47	26.73	42.99
FADs	WTRA	15.58	27.39	44.18
Drifters	ARAB	14.22	23.79	39.19
FADs	ARAB	27.17	45.67	75.87
Drifters	EAFR	18.77	33.28	56.74
FADs	EAFR	22.63	36.72	55.12
Drifters	ISSG	13.70	22.20	33.39
FADs	ISSG	18.36	28.56	40.73
Drifters	MONS	17.00	27.89	43.64
FADs	MONS	21.27	34.51	52.65

Ocean	$deltaD_deg$	deltaT_day	$speed_component$	n	Corr	Slope	Slope_lower	Slope_upper
Indian	1/6	0.5	velocity	10,015	0.85	1	1	1.01
Indian	1/6	1	velocity	$14,\!556$	0.83	1.01	1	1.01
Indian	1/6	2.5	velocity	$25,\!956$	0.78	1.02	1.02	1.03
Atlantic	1/6	0.5	velocity	$2,\!842$	0.73	1.12	1.1	1.14
Atlantic	1/6	1	velocity	$4,\!146$	0.75	1.15	1.13	1.16
Atlantic	1/6	2.5	velocity	7,739	0.71	1.2	1.18	1.21
Indian	1/6	0.5	u	$10,\!015$	0.93	1	0.99	1
Indian	1/6	1	u	$14,\!556$	0.93	1	0.99	1.01
Indian	1/6	2.5	u	$25,\!956$	0.9	1.01	1	1.01
Atlantic	1/6	0.5	u	2,842	0.87	1.16	1.14	1.19
Atlantic	1/6	1	u	$4,\!146$	0.87	1.17	1.15	1.19
Atlantic	1/6	2.5	u	7,739	0.85	1.21	1.19	1.22
Indian	1/6	0.5	V	$10,\!015$	0.88	1.02	1.01	1.03
Indian	1/6	1	V	$14,\!556$	0.85	1.04	1.03	1.05
Indian	1/6	2.5	V	$25,\!956$	0.77	1.08	1.07	1.09
Atlantic	1/6	0.5	v	$2,\!842$	0.69	1.06	1.02	1.1
Atlantic	1/6	1	v	$4,\!146$	0.68	1.16	1.12	1.2
Atlantic	1/6	2.5	V	7,739	0.58	1.33	1.29	1.37

Table A3: The number of observations, correlation coefficients and slope of the velocity components for fish aggregating devices (FADs) versus drifters at different spatio-temporal buffers in the Atlantic and Indian Oceans.



Figure A1: Examples of long-associated drift across the Indian Ocean featuring fish aggregating device (FAD) buoy n°17179 (red triangles) and drifter n°109550 (blue crosses) on the left, FAD buoy n°16812 and drifter n°109364 on the right, sharing similar trajectories between August and November 2013.



Figure A2: The comparison of zonal velocities between fish aggregating devices (FADs) and Ocean Surface Currents Analyses Real-time (OSCAR) in the selected Longhurst biogeographical provinces of the Atlantic Ocean (top) and Indian Ocean (bottom). The solid line indicates the major axis regression model and the dashed line indicates the 1:1 isoline.



Figure A3: The comparison of zonal velocities between drifters and Ocean Surface Currents Analyses Real-time (OSCAR) in the selected Longhurst biogeographical provinces of the Atlantic Ocean (top) and Indian Ocean (bottom). The solid line indicates the major axis regression model and the dashed line indicates the 1:1 isoline.



Figure A4: The comparison of meridional velocities between fish aggregating devices (FADs) and Ocean Surface Currents Analyses Real-time (OSCAR) in the selected Longhurst biogeographical provinces of the Atlantic Ocean (top) and Indian Ocean (bottom). The solid line indicates the major axis regression model and the dashed line indicates the 1:1 isoline.



Figure A5: The comparison of meridional velocities between drifters and Ocean Surface Currents Analyses Real-time (OSCAR) in the selected Longhurst biogeographical provinces of the Atlantic Ocean (top) and Indian Ocean (bottom). The solid line indicates the major axis regression model and the dashed line indicates the 1:1 isoline.

Table A4: Summary of the major axis regression models fitted to the velocity components of the Ocean Surface Currents Analyses Real-time (OSCAR) measurements versus fish aggregating devices (FADs) and OSCAR versus drifters in the selected Longhurst biogeographical provinces of the Atlantic and Indian Oceans. Slope low. = 2.5% quantile value used as the lower limit of the regression slope estimate; Slope upp. = 97.5% quantile value used as the upper limit of the regression slope estimate. r = Pearson's correlation coefficient.

Device	Component	Province	Slope low.	Slope	Sloper upp.	r
FADs	Zonal	ETRA	0.760	0.762	0.764	0.64
FADs	Zonal	GUIN	0.689	0.695	0.702	0.38
FADs	Zonal	NATR	0.220	0.223	0.225	0.46
FADs	Zonal	WTRA	0.690	0.693	0.695	0.73
FADs	Zonal	ARAB	0.870	0.874	0.877	0.66
FADs	Zonal	EAFR	0.601	0.603	0.606	0.74
FADs	Zonal	ISSG	0.651	0.652	0.654	0.78
FADs	Zonal	MONS	0.730	0.731	0.731	0.76
Drifters	Zonal	ETRA	0.692	0.698	0.703	0.71
Drifters	Zonal	GUIN	0.550	0.568	0.587	0.56
Drifters	Zonal	NATR	0.419	0.422	0.424	0.59
Drifters	Zonal	WTRA	0.737	0.744	0.750	0.66
Drifters	Zonal	ARAB	0.741	0.753	0.764	0.64
Drifters	Zonal	EAFR	0.723	0.729	0.735	0.78
Drifters	Zonal	ISSG	0.620	0.623	0.625	0.77
Drifters	Zonal	MONS	0.766	0.769	0.773	0.77
FADs	Meridional	ETRA	0.506	0.510	0.514	0.26
FADs	Meridional	GUIN	0.674	0.684	0.693	0.29
FADs	Meridional	NATR	0.415	0.420	0.425	0.56
FADs	Meridional	WTRA	0.548	0.553	0.557	0.43
FADs	Meridional	ARAB	0.709	0.711	0.714	0.66
FADs	Meridional	EAFR	0.536	0.538	0.541	0.73
FADs	Meridional	ISSG	0.605	0.608	0.611	0.65
FADs	Meridional	MONS	0.665	0.666	0.668	0.53
Drifters	Meridional	ETRA	0.385	0.394	0.403	0.32
Drifters	Meridional	GUIN	1.094	1.175	1.264	0.31
Drifters	Meridional	NATR	0.352	0.355	0.357	0.56
Drifters	Meridional	WTRA	0.659	0.670	0.680	0.46
Drifters	Meridional	ARAB	0.634	0.643	0.652	0.68
Drifters	Meridional	EAFR	0.632	0.638	0.644	0.77
Drifters	Meridional	ISSG	0.540	0.543	0.545	0.71
Drifters	Meridional	MONS	0.682	0.688	0.694	0.49

Table A5: The number of observations and correlation coefficients of the velocity components of Ocean Surface Currents Analyses Real-time (OSCAR) versus fish aggregating devices (FADs) and OSCAR versus drifters for the entire dataset (n, Corr_u, Corr_v) and for the datasets subsampled every 5 days (n5, corr5_u, corr5_v) and 15 days (n15, corr15_u, corr15_v) in the Atlantic and Indian Oceans.

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Χ	Device	Ocean	n	Corr_u	Corr_v	n5	$corr5_u$	$corr5_v$	n15	corr15_u	corr1
1	FADs	Atlantic	$1,\!393,\!100$	0.62	0.31	66,228	0.57	0.27	$36,\!986$	0.53	0.24
2	FADs	Indian	$3,\!384,\!424$	0.75	0.58	$181,\!193$	0.72	0.56	$108,\!611$	0.70	0.55
3	Drifters	Atlantic	$634,\!297$	0.58	0.49	$32,\!303$	0.58	0.47	11,211	0.57	0.49
4	Drifters	Indian	458,065	0.74	0.63	$23,\!418$	0.75	0.64	$8,\!187$	0.77	0.65



Figure A6: The comparison of slopes of major axis regression models fitted to the (A) zonal and (B) meridional velocity data of the Ocean Surface Currents Analyses Real-time (OSCAR) versus fish aggregating devices (FADs) and OSCAR versus drifters in the selected Longhurst biogeographical provinces.



Figure A7: Relative frequency distributions of the index d/D that describe the prediction skill of fish aggregating devices (FADs; dashed red curves) and drifters (solid blue curves), with respect to the Ocean Surface Currents Analyses Real-time (OSCAR) velocities for the selected Longhurst biogeographical provinces, where d is the distance between the projected and observed location at the next time step and D is the distance between the current and next observed locations.

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