

# Highlights

## **Sources of marine debris for Seychelles and other remote islands in the western Indian Ocean**

Vogt-Vincent, N.S., Burt, A.J., Kaplan, D., Mitarai, S., Turnbull, L.A., Johnson, H.L.

- We use Lagrangian trajectory analysis to quantify sources of marine debris for Seychelles and other remote islands in the western Indian Ocean.
- Most terrestrial debris beaching at Seychelles comes from Indonesia, with contributions from India, Sri Lanka, Tanzania, Comoros, and Seychelles itself.
- Seychelles is at very high risk from debris of marine origin from fisheries and shipping lanes.
- Debris accumulation rates across Seychelles are likely strongly seasonal, and possibly amplified during positive Indian Ocean Dipole and El Niño phases.

# Sources of marine debris for Seychelles and other remote islands in the western Indian Ocean

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## Abstract

Vast quantities of marine debris have beached at remote islands in the western Indian Ocean such as Seychelles, but little is known about where this debris comes from. To identify these sources and temporal patterns in accumulation rate, we carried out global Lagrangian particle tracking experiments incorporating surface currents, waves, and variable windage, beaching, and sinking rates, taking into account both terrestrial (coastal populations and rivers) and marine (fisheries and shipping) sources of debris. Our results show that, whilst low-buoyancy terrestrial debris may originate from the western Indian Ocean (principally Tanzania, Comoros, and Seychelles), most terrestrial debris beaching at remote western Indian Ocean islands drifts from the eastern and northern Indian Ocean, primarily Indonesia and, to a lesser extent, India and Sri Lanka. Purse-seine fragments beaching at Seychelles are likely associated with fishing activity in the western Indian Ocean, but longline fragments may also be swept from the southeastern Indian Ocean. The entire of Seychelles is at very high risk from waste discarded from shipping routes transiting the Indian Ocean, and comparison with observations suggests that many bottles washing up on beaches may indeed originate from

these routes. Our analyses indicate that marine debris accumulation at Seychelles (and the Outer Islands in particular) is likely to be strongly seasonal, peaking during February-April, and this pattern is driven by local monsoonal winds. This seasonal cycle may be amplified during positive Indian Ocean Dipole phases and El-Niño events. These results underline the vulnerability of small island developing states to marine plastic pollution, and are a crucial first step towards improved management of the issue. The Lagrangian trajectories used in this study are available for download, and our analyses can be rerun under different parameters using the associated scripts.

*Keywords:* Marine debris, Indian Ocean, Seychelles, Plastic, Monsoon, Lagrangian

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## 1. Introduction

Marine plastic pollution is a significant environmental threat, both for marine ecosystems (Gall and Thompson, 2015), and the communities that depend on the ocean for sustenance, tourism, and other social and economic activities (Thompson et al., 2009; Werner et al., 2016). Only a small proportion of plastic thought to have entered the marine environment remains floating at the ocean surface (Cózar et al., 2014), with the vast majority sinking to deep sea sediments (Woodall et al., 2014) or beaching on coasts (Onink et al., 2021). Beached marine debris in particular is of great concern; coastal environments are highly productive and biodiverse so the accumulation of debris on coasts can be damaging to both marine and terrestrial organisms (e.g. Nelms et al., 2016; Bergmann et al., 2017), and is associated with significant economic costs (Newman et al., 2015). On some coastlines, much of the accumulated debris may be of local origin (e.g. Martinez-Ribes et al., 2007; Turrell, 2020). Elsewhere, however, particularly in the case of remote islands with minimal or no

14 local population, most debris accumulating on the coast may have been transported over  
 15 great distances by ocean currents, winds, and waves prior to beaching (van Sebille et al.,  
 16 2020). These islands, many of which belong to small island developing states, are faced  
 17 with the deeply inequitable situation of bearing the costs of removing waste they were not  
 18 responsible for generating, contrary to the “polluter pays” principle (OECD, 1975).

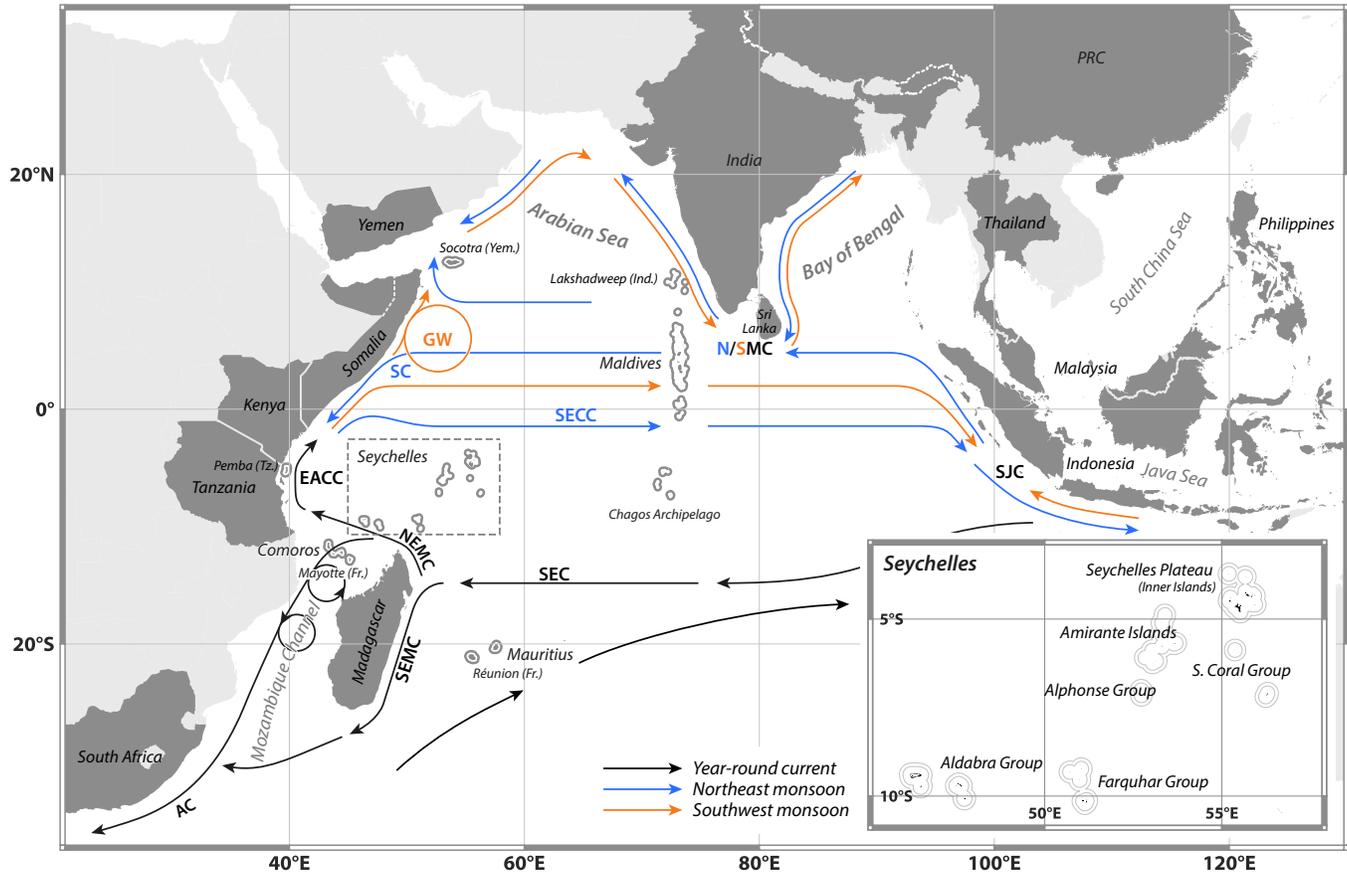


Figure 1: Map of the Indian Ocean, with key countries, island groups, and basins discussed in this study highlighted. Small islands are drawn with a halo for clarity. Arrows represent the major surface currents in the Indian Ocean, adapted from Schott et al. (2009). Black arrows represent currents that broadly occupy the same location year-round, whereas blue and orange arrows respectively represent currents during the northeast and southwest monsoon. *Inset*: The major island groups within Seychelles.

19 There are many small island developing states in the western Indian Ocean (Figure 1)  
 20 and, whilst marine plastic pollution is under-studied in this region in comparison to, for  
 21 instance, the North Atlantic and western Pacific, debris accumulation has been documented

22 in many of these remote island groups. Seychelles is one such small island developing state,  
23 spread across over 100 islands north of Madagascar, from the isolated Aldabra Group in the  
24 southwest, to the Inner Islands on the Seychelles Plateau in the northeast. Marine debris  
25 monitoring programmes have found large quantities of marine debris accumulating across  
26 the latitudinal and longitudinal range spanned by Seychelles, such as at Aldabra Atoll (Burt  
27 et al., 2020), Alphonse Island (Duhec et al., 2015), Cousine Island (Dunlop et al., 2020), and  
28 many others (Macmillan et al., 2022). Marine debris is primarily of terrestrial origin at some  
29 of these sites (e.g. Alphonse Island, Duhec et al. (2015)) whereas abandoned, lost, or oth-  
30 erwise discarded fishing gear (ALDFG) of marine origin dominates at others (e.g. Aldabra  
31 Atoll, Burt et al. (2020)).

32

33 Attribution of marine debris accumulating at these remote islands would be a positive step  
34 towards accountability and prevention, but this is challenging. Several studies have inferred  
35 the sources of beached debris based on intact labels on bottles (e.g. Duhec et al., 2015; Burt  
36 et al., 2020), but this method has historically been limited to small sample sizes, is biased  
37 against debris lacking intact labels due to degradation and/or biofouling, and cannot give  
38 representative provenance information for all types of marine debris, as transport pathways  
39 vary greatly with debris geometry and composition (Duhec et al., 2015; Maximenko et al.,  
40 2018).

41

42 Numerical models can also be used to predict the source of beaching debris by repre-  
43 senting debris as Lagrangian particles or Eulerian tracers. These simulations can be run  
44 forward-in-time, i.e. assuming knowledge of some input distribution of marine debris and

45 predicting where that debris is transported (e.g. Kaandorp et al., 2020; van der Mheen et al.,  
46 2020; Chassignet et al., 2021), or backward-in-time, i.e. simulating trajectories that lead to a  
47 site of interest and inferring debris sources based on where debris passed through in the past  
48 (e.g. Duhec et al., 2015; Stelfox et al., 2020). In the context of marine debris attribution for  
49 remote islands, backward-in-time simulations are more efficient as they must only compute  
50 the small subset of trajectories that end at the site of interest, reducing computational cost.  
51 However, there are significant limitations associated with the backward-in-time approach.  
52 For instance, it is not possible to implement parameterisations for subgrid-scale diffusivity.  
53 Even more significantly, since simulated backward trajectories comprise an unknown subset  
54 of all possible trajectories, there are fundamental limitations on the quantitative constraints  
55 that can be obtained on the sources of marine debris. Most studies using backward-in-time  
56 simulations are limited to qualitative predictions of debris sources based on assumptions of  
57 a fixed drift time (e.g. Duhec et al., 2015). van Duinen et al. (2022) used a Bayesian frame-  
58 work to quantify sources of debris for a beach in the Netherlands, but this approach still  
59 relies on assumptions on how long debris were adrift before beaching. For remote islands  
60 where potential sources of debris are many and distal, it is challenging to justify any *a priori*  
61 assumption for a drift time distribution. An innovative approach was used by Stelfox et al.  
62 (2020), who predicted the source fisheries for ghost gear accumulating in the Maldives based  
63 on backward-in-time simulations and constraints on drift time from biofouling. Unfortu-  
64 nately, these constraints are likely debris-type and site specific, and no such estimates exist  
65 in general for most remote islands.

66

67 In the absence of constraints on drift time, forward-in-time simulations are required to

68 provide quantitative, physically justifiable estimates for sources of marine debris. To date,  
69 sources of debris have been quantified for the Seychelles as part of two regional (van der  
70 Mheen et al., 2020) and global (Chassignet et al., 2021) studies. However, neither study  
71 registered a significant number of particles arriving at Seychelles, and they were therefore  
72 unable to make robust conclusions about sources of marine debris for remote islands. Both  
73 were large-scale studies focusing on major marine debris transport pathways, but this nev-  
74 ertheless highlights an important data gap, as well as a particular technical challenge for  
75 assessing sources of marine debris for small and remote locations.

76

77 As a result, whilst there are indications from bottle labels, no quantitative estimates exist  
78 for the relative importance of sources of debris for Seychelles, along with other remote island  
79 groups in the western Indian Ocean. Good constraints exist on source regions for one specific  
80 type of fishing gear, drifting Fish Aggregating Devices (dFADs), accumulating on Seychelles'  
81 beaches (Macmillan et al., 2022; Imzilen et al., 2021), but this has not been generalised  
82 to all marine-based sources of debris. In this study, we use large-scale Lagrangian forward  
83 simulations, forced by ocean currents, waves, and winds, generalisable to arbitrary sinking  
84 and beaching rates, to answer the following questions:

- 85 • Which countries are the most likely terrestrial sources of debris accumulating at Sey-  
86 chelles (and other western Indian Ocean islands), and how sensitive are these estimates  
87 to debris properties such as sinking rate and windage?
- 88 • If debris is generated at sea (from fisheries, ships, etc.), from which regions is there most  
89 risk of debris beaching at Seychelles, and can we therefore predict high-risk fisheries

90 and shipping channels?

- 91 • What are the physical drivers of marine debris accumulation at Seychelles, and are  
92 variations in accumulation rates (seasonal and inter-annual) predictable, allowing for  
93 more targeted cleanup efforts?

## 94 **2. Methods**

### 95 *2.1. Particle tracking*

96 To simulate the transport of marine debris, we carry out Lagrangian particle tracking us-  
97 ing OceanParcels (Lange and Seville, 2017; Delandmeter and van Seville, 2019). Particles are  
98 tracked for 10 years or until the end of 2019, with trajectories integrated using a fourth-order  
99 Runge-Kutta scheme and a time-step of 1 hour. Over large scales, buoyant marine debris  
100 is transported by surface currents, Stokes drift, and in the case of debris protruding above  
101 the sea surface, windage (van Seville et al., 2020), and all three processes are important in  
102 describing its dispersal (e.g. Duhec et al., 2015; Maximenko et al., 2018). We assume the  
103 force experienced by particles from the wind is parallel and proportional to surface winds,  
104 but note that this is a simplification compared to the real forces experienced by buoyant  
105 debris (Domon et al., 2012). We advect particles of terrestrial origin with 5 forcing scenar-  
106 ios: just surface currents (C0), surface currents + Stokes drift (CS0), and surface currents  
107 + Stokes drift + 1-3% windage (CS1-3). Particles of marine origin are advected using the  
108 same sets of forcing, plus 4% and 5% windage (CS4-5). We used the 1/12° CMEMS Global  
109 Ocean Physics Analysis GLORYS12V1 (Lellouche et al., 2021) for daily surface currents,  
110 1/5° Global Wave Reanalysis WAVERYS (Law-Chune, 2021) for three-hourly Stokes drift,  
111 and 1/4° three-hourly surface winds from ERA5 (Hersbach et al., 2020) (all 1993-2019).

112 All three forcing sets are provided by CMEMS, regridded to a regular grid. We applied a  
113 homogeneous lateral diffusivity of  $10 \text{ m}^2 \text{ s}^{-1}$  to particles, based on a typical value of the hori-  
114 zontal Smagorinsky diffusivity in the equatorial Indian Ocean diagnosed from GLORYS12V1  
115 ((Smagorinsky, 1963), Supplementary Figure 1) and in line with previous studies (Okubo,  
116 1971; Kaandorp et al., 2020). Further technical details on the treatment of particle tracking  
117 near the coasts are described in Supplementary Text 1.

118

## 119 *2.2. Particle sinking and beaching*

120 Marine debris is lost from the ocean surface through processes including beaching and  
121 sinking. These processes are complex and driven by small-scale physical and biological pro-  
122 cesses (van Sebille et al., 2020) and must therefore be parameterised in large-scale numerical  
123 models. Many models parameterise sinking as decay in the mass of debris represented by a  
124 particle (e.g. Kaandorp et al., 2020; Chassignet et al., 2021). Beaching is often parameterised  
125 by explicitly removing particles based on criteria, such as particles entering a land cell due  
126 to Stokes drift, wind and/or numerical error (e.g. Zhang et al., 2020; Cardoso and Caldeira,  
127 2021), particle stagnation (e.g. Seo and Park, 2020; Bosi et al., 2021), or as a stochastic  
128 process associated with some probability (e.g. van der Mheen et al., 2020; Onink et al., 2021).

129

130 An advantage with modelling beaching as a stochastic process is the ability to incorporate  
131 complex behaviour such as resuspension (Liubartseva et al., 2018; Onink et al., 2021) and,  
132 as understanding of the physics of beaching improves, stochastic parameterisations will be-  
133 come an increasingly valuable tool. However, as these parameterisations remove Lagrangian

134 particles from circulation (even if only temporarily), this can significantly reduce the number  
135 of particles representing floating debris in the model. This is a problem when attempting  
136 to quantify the sources of debris for small and remote islands: these islands are very small  
137 ‘targets’ and beaching events may be missed due to an insufficient number of particles, as  
138 was the case for Seychelles in the studies of van der Mheen et al. (2020) and Chassignet et al.  
139 (2021).

140

141 Instead, we assume that there is (i) a constant rate of debris removal through sink-  
142 ing,  $\mu_s$ , and (ii) a constant rate of debris removal through beaching,  $\mu_b^*$  when a particle is  
143 within a  $1/12^\circ$  coastal grid cell, and implement sinking and beaching offline through post-  
144 processing of the trajectory data. We store these beaching events for 18 sites within Sey-  
145 chelles (Aldabra, Assomption, Cosmoledo, Astove, Providence, Farquhar, Alphonse, Poivre,  
146 St Joseph, Desroches, Platte, Coëtivy, Mahé, Fregate, Silhouette, Praslin, Denis, and Bird).  
147 For our terrestrial-sourced debris experiments (section 2.3.1), we include an additional 9  
148 sites from the wider western Indian Ocean (Comoros, Mayotte [France], Lakshadweep [In-  
149 dia], Maldives, Mauritius, Réunion [France], Pemba [Tanzania], Socotra [Yemen], and the  
150 Chagos Archipelago). For brevity, we focus on Seychelles in this paper, specifically islands  
151 on the Seychelles Plateau, and the Aldabra Group as representative of the Outer Islands.  
152 Analyses and figures for other island groups that could not be included in this paper can be  
153 produced using the scripts in Supplementary Dataset 1.

154

155 By efficiently choosing which data to store during particle tracking simulations (see Sup-  
156 plementary Text 2), it is possible to compress all data required to reconstruct almost all

157 beaching events from the over  $2 \times 10^{11}$  particles used across all our simulations in <1TB,  
158 whilst allowing key parameters to be varied through postprocessing without having to rerun  
159 simulations.

160

### 161 *2.3. Debris sources*

162 We classify marine plastic debris into terrestrial sources (debris that entered the ocean  
163 from coastlines) and marine sources (debris that entered the ocean at sea). Due to the  
164 relatively poor constraints on the input distribution and magnitude of marine sources, we  
165 use different approaches to consider terrestrial and marine sources.

#### 166 *2.3.1. Terrestrial debris sources*

167 Debris can enter the ocean through rivers (transported from inland), as well as through  
168 direct coastal input from coastal populations through stormwater, sewage, or poor waste  
169 disposal (Mihai et al., 2022). For riverine debris input, we use the modelled midpoint annual  
170 estimates from Meijer et al. (2021), gridding the emissions from each river mouth to the  
171 nearest coastal cell on the  $1/12^\circ$  GLORYS12V1 grid (section 2.1). For direct coastal input,  
172 we base our estimates on modelled annual mismanaged plastic waste generation estimates  
173 from Lebreton and Andrady (2019). We degraded the resolution of this product to the GLO-  
174 RYS12V1 resolution, and then calculated emissions to the ocean by assuming that a fraction  
175  $f_i = f_c \cdot \exp\left[-\left(\frac{d_i}{L}\right)^2\right]$  of the mismanaged waste produced in a grid cell  $i$  enters the nearest  
176 coastal cell, where  $f_c$  is the maximum likelihood of mismanaged waste entering the ocean,  $d_i$   
177 is the distance of cell  $i$  from the coast, and  $L$  is a length scale over which direct coastal input  
178 to the ocean is significant. This parameterisation is based on the assumption that waste is

179 less likely to enter the ocean the further from the coast it is generated. Many previous stud-  
180 ies have used the alternative assumption, inherited from Jambeck et al. (2015), that a fixed  
181 fraction of mismanaged waste generated within 50km of the coast enters the ocean. Both  
182 of these parameterisations are somewhat arbitrary, but we believe that our assumptions are  
183 more appropriate.

184

185 We set  $L = 15\text{km}$  to reflect the length scale of a typical coastal city. The parameter  $f_c$   
186 is the main control on the ratio  $r$  of marine debris generation from coastal versus riverine  
187 sources. In the absence of good constraints on this parameter, we take  $f_c = 0.25$ , corre-  
188 sponding to a total flux of debris from coastal and riverine sources of  $3.1 \text{ Mt y}^{-1}$  and  $1.0$   
189  $\text{Mt y}^{-1}$  respectively ( $r = 3.1$ , between  $r = 1.9$  in Kaandorp et al. (2020) and  $r = 4.9$  in  
190 Lebreton et al. (2018)). The parameter  $f_c$  can, however, be modified during postprocessing  
191 and, if it becomes better constrained in the future, it is straightforward to regenerate our  
192 results for another value of  $f_c$ , or even an entirely different debris input distribution, using  
193 the trajectories in Supplementary Dataset 1.

194

195 To minimise the cost of simulations, we only considered coastal cells for countries that  
196 could reasonably act as a source of marine debris for islands in the western Indian Ocean,  
197 identified from a preliminary backward particle tracking experiment (Supplementary Text  
198 3, Supplementary Figure 2). Many coastal cells were associated with a very small flux of  
199 debris, so we removed the 7773 (of 20742) coastal cells with the smallest contributions, leaving  
200 99.99% of riverine plastic, and 99.9% of coastal plastic. An overview of the terrestrial sources  
201 of marine debris used in our experiments is shown in Figure 2.

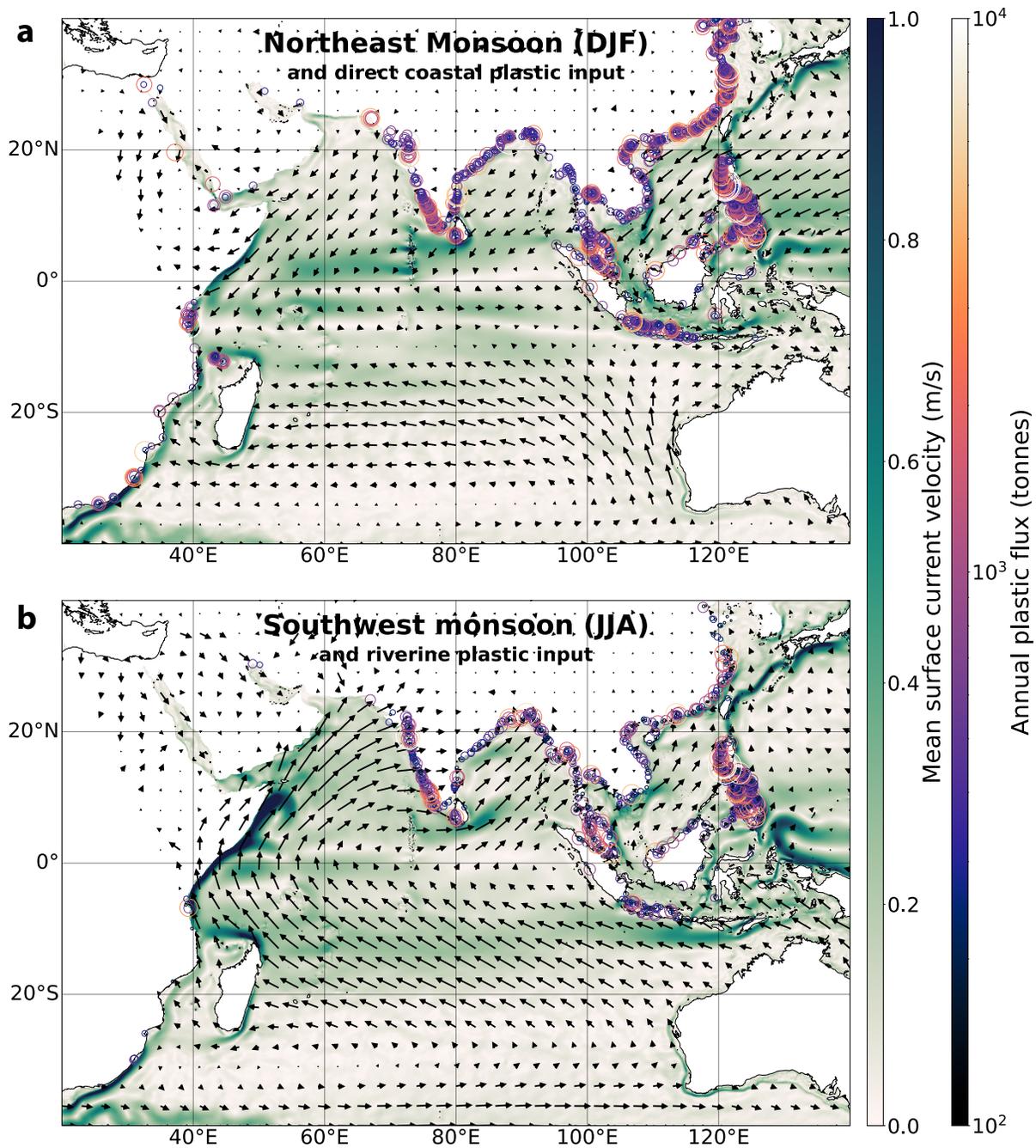


Figure 2: (a) Mean surface current speed (colours, Lellouche et al. (2021)) and surface winds (arrows, Hersbach et al. (2020)) for the northeast monsoon. Terrestrial sources of marine debris from direct coastal input, as used in our analyses, are overlaid, but note that this has no relation to the season. (b) Mean surface ocean current speed and surface winds for the southwest monsoon. Terrestrial sources of marine debris from riverine input are overlaid. This figure is zoomed to focus on the Indian Ocean region so does not include all sources of debris considered in the model; please see Supplementary Text 3 for a full list.

202 2.3.2. *Marine sources*

203 Global Fishing Watch uses tracking data from the automatic identification system (AIS),  
204 broadcast by large ships, to estimate fishing effort for tracked vessels (Kroodsma et al., 2018),  
205 which has been used as a proxy for ALDFG production in previous studies (e.g. Kaandorp  
206 et al., 2020). However, AIS coverage is poor in the Indian Ocean (Richardson, 2022). In-  
207 stead, we use publicly available Indian Ocean Tuna Commission effort data for purse-seines  
208 and longlines (provided on  $1^\circ$  and  $5^\circ$  grids respectively), which is well-studied and has been  
209 used extensively as an indicator of fishing activity, particularly in the case of purse-seines  
210 (Kaplan et al., 2014; Imzilen et al., 2022). We consider longline fisheries from Japan, Taiwan,  
211 and Korea only, as data from these countries is the most reliable in terms of spatial distribu-  
212 tion (Kaplan et al., 2014, Emmanuel Chassot (*personal communication*)). Since purse-seine  
213 and longline vessels lose gear at different rates (Kuczenski et al., 2022) with potentially dif-  
214 ferent behaviour in the water, we do not aggregate effort from these two fisheries, and instead  
215 consider them separately.

216

217 From these fishing effort data, we can generate ‘risk maps’, representing where debris from  
218 a particular fishery beaching at a particular site is most likely to come from. This cal-  
219 culation requires a matrix  $P_i(x, y, t_s, t_b)$  (for a particular debris class), giving the likeli-  
220 hood that debris beaches at site  $i$  in month  $t_b$ , given that it entered the ocean at  $(x, y)$   
221 in month  $t_s$ ; and a matrix  $E_j(x, y, t)$ , giving the fishing effort of fishery  $j$  at  $(x, y)$  in month  
222  $t$ . The relative flux  $f_{ij}(x, y)$  of fishery  $j$  debris from a point that *ever* beaches at site  
223  $i$  is given by  $f_{ij}(x, y) = \sum_{t_s=1}^{12} (E_j(x, y, t_s) \cdot \sum_{t_b=1}^{12} P_i(x, y, t_s, t_b))$ . We can normalise this  
224 relative flux by the total flux from all sources, to give the *risk*  $R_{ij}(x, y)$  to site  $j$  from

225 fishery  $i$  at location  $(x, y)$ ,  $R_{ij}(x, y) = \frac{f_{ij}(x, y)}{\sum^{x, y} f_{ij}(x, y)}$ . Along similar lines, we can also com-  
 226 pute a monthly climatology  $B_{ij}(t_b)$  for beaching rates from fishery  $j$  accumulating at site  $i$ ,  
 227  $B_{ij}(t_b) = \sum^{x, y} \left( \sum_{t_s=1}^{12} (E_j(x, y, t_s) \cdot P_i(x, y, t_s, t_b)) \right)$ .

228

229 Debris may also be discarded or lost at sea from shipping traffic, which was suggested  
 230 as a potentially significant source of debris for Alphonse Island, Seychelles by Duhec et al.  
 231 (2015). This debris source is challenging to quantify, but we have used AIS-based estimates of  
 232 shipping traffic intensity from Cerdeiro et al. (2020) as an indication of where major shipping  
 233 lanes in the Indian Ocean are.

## 234 *2.4. Seeding strategy*

### 235 *2.4.1. Terrestrial sources*

236 For each coastal cell  $i$  on the GLORYS12V1 grid, we split the annual flux of debris of  
 237 terrestrial origin  $F_i$  (as described in section 2.3.1) across 4 equally spaced releases per month,  
 238 for a total of 48 identical releases per year. The debris associated with each release was further  
 239 divided across  $n_i$  particles, such that the initial mass associated with a particular particle  
 240  $j$  released at cell  $i$  is  $M_j^0 = F_i/48n_i$ . We set  $n_i = \lceil c_1 \cdot \log_{10}[F_i] - c_2 \rceil^2$ , where  $c_1 = 16$  and  
 241  $c_2 = 18.4$  are arbitrary parameters chosen to distribute particles reasonably whilst keeping  
 242 computation tractable. We released 13.7 million terrestrial particles per release event, for a  
 243 total of 656 million per model year.

### 244 *2.4.2. Marine sources*

245 In each marine cell between 20°E-130°E and 40°S-30°N (excluding the Mediterranean)  
 246 we generated 36 particles per release. As for terrestrial sources, we released particles at four

247 equally spaced intervals per month, with 26.5 million particles per release event and a total  
248 of 1.27 billion particles per model year.

## 249 2.5. Debris Classes

250 In our model, the behaviour of marine debris in the ocean is set by the three parameters  
251  $\mu_s$  (sinking rate),  $\mu_b^*$  (beaching rate), and the forcing scenario. No one set of parameters will  
252 describe all marine debris, and constraints on all three are poor. We explore the sensitivity  
253 of model results to this parameter-space in section 3.3.1, but to provide concrete examples,  
254 we have defined four representative debris classes:

- 255 • **Class A:**  $1/\mu_b = 30\text{d}$ ,  $1/\mu_s = 30\text{d}$ , scenario **CS0**. Low volume mm-scale plastics with  
256 low (but positive) buoyancy and negligible exposure, e.g. **small plastic fragments,**  
257 **nurdles.**
- 258 • **Class B:**  $1/\mu_b = 30\text{d}$ ,  $1/\mu_s = 90\text{d}$ , scenario **CS1**. Moderate volume cm-scale plas-  
259 tics with moderate positive buoyancy and minor exposure, e.g. **bottle caps, small**  
260 **domestic items.**
- 261 • **Class C:**  $1/\mu_b = 30\text{d}$ ,  $1/\mu_s = 360\text{d}$ , scenario **CS3**. Moderate-large plastics with  
262 high positive buoyancy and moderate exposure, e.g. **beach sandals, bottles, foam**  
263 **sheets, buoyant nets.**
- 264 • **Class D:** (marine sources only)  $1/\mu_b = 30\text{d}$ ,  $1/\mu_s = 1800\text{d}$ , scenario **CS5**. Large  
265 plastics with very high positive buoyancy and high exposure, e.g. **fishing debris with**  
266 **buoys attached, robust empty bottles.**

267 To derive these classifications, we used guidance on windage coefficients from Duhec et al.  
268 (2015) and Domon et al. (2012), sinking rates from Fazey and Ryan (2016), and beaching rates  
269 from our own analysis (Supplementary Text 4, Supplementary Figures 3-4) and Kaandorp  
270 et al. (2020). However, we stress that windage coefficients and sinking rates for different types  
271 of marine debris remain poorly constrained and the classes we have defined are suggestions  
272 only. All trajectories computed for this study (and scripts required to reproduce beaching  
273 rates) are provided in Supplementary Dataset 1, so practitioners can recompute predictions  
274 for parameters of interest.

## 275 *2.6. Comparison with observations*

276 Burt et al. (2020) estimated the total mass of debris that accumulated on Aldabra Atoll  
277 (Seychelles), as well as countries of origin for a small sample of PET bottles. Quantitative  
278 source analyses have also been carried out for Alphonse, Coëtivy, Astove and Platte (The  
279 Ocean Project Seychelles, 2019; Dunlop et al., 2020). Finally, Macmillan et al. (2022) anal-  
280 ysed patterns of (satellite-tracked) drifting Fish Aggregating Device (dFAD) beaching events  
281 across Seychelles. We carried out a quantitative, side-by-side comparison of our analyses  
282 against the findings of these studies to identify limitations in both our approach, and these  
283 observational assessments of marine debris accumulation on remote western Indian Ocean  
284 islands.

### 285 3. Results and discussion

#### 286 3.1. Sources of debris for remote islands in the western Indian Ocean

##### 287 3.1.1. Debris of terrestrial origin

288 There is significant variation in the predicted source countries for debris beaching at the  
289 27 sites investigated in this study (Figure 3). These figures can be interpreted as the predicted  
290 likelihood of a fragment of marine debris originating from the source country and beaching  
291 at the target island (group), given that it has properties reflecting Class A, B, or C debris as  
292 defined above.

293 For Class A debris (Figure 3(a)), East Africa (predominantly Tanzania) is expected to  
294 be the largest source of marine debris for most of the Outer Islands of Seychelles, although  
295 Comoros is the dominant source for Aldabra and Assomption. For the Inner Islands on the  
296 Seychelles Plateau, most Class A debris is expected to come from within Seychelles, with the  
297 remainder sourced from East Africa. For sites in the central-northern Indian Ocean (Maldives  
298 and Lakshadweep), India and/or Sri Lanka are expected to be the principal sources of debris.  
299 Only the Chagos Archipelago is predicted to source most of its Class A debris from Indonesia.

300

301 For Class B debris (Figure 3(b)), a combination of longer residence time at the ocean  
302 surface (3 months), westward Stokes drift, and easterly winds allows Indonesia to begin to  
303 dominate the marine debris budget for much of the western Indian Ocean. Our analyses  
304 predict that Indonesia is responsible for over 50% of all Class B debris for all Outer Islands  
305 of Seychelles (and remains the dominant source for the Chagos Archipelago). Seychelles  
306 and Tanzania are still expected to be significant sources of debris within the inner islands

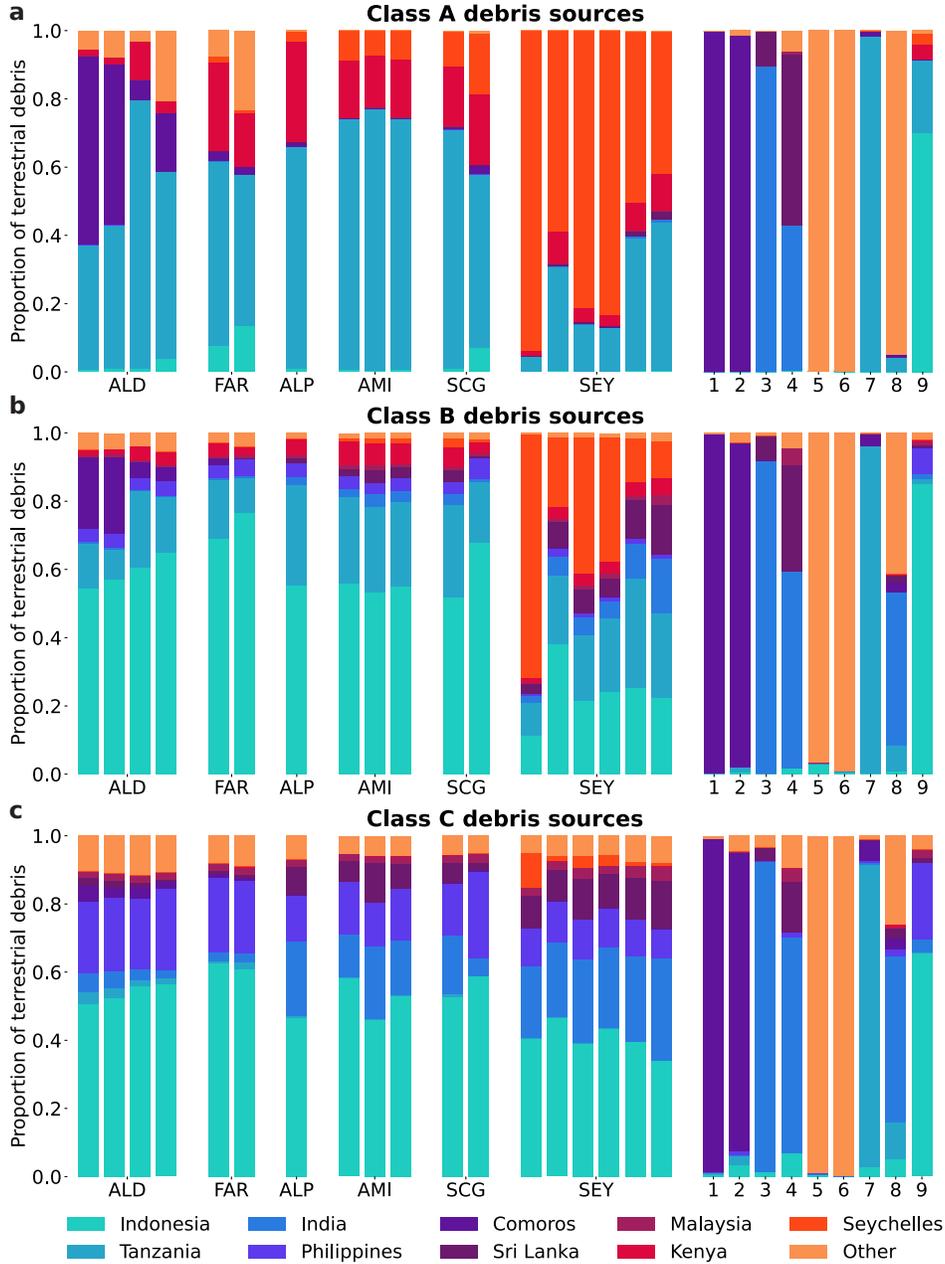


Figure 3: Sources of beaching (terrestrial) debris from all debris releases 1993-2014 for (a) Class A, (b) Class B, and (c) Class C debris. Sites from left to right: **Aldabra Group** (Aldabra, Assumption, Cosmoledo, Astove), **Farquhar Group** (Providence, Farquhar), **Alphonse Group** (Alphonse), **Amirante Islands** (Poivre, St Joseph, Desroches), **Southern Coral Group** (Platte, Coëtivy), **Seychelles Plateau** (Mahé, Fregate, Silhouette, Praslin, Denis, Bird); Comoros (**1**), Mayotte (**2**), Lakshadweep, India (**3**), Maldives (**4**), Mauritius (**5**), Réunion, France (**6**), Pemba, Tanzania (**7**), Socotra, Yemen (**8**), Chagos Archipelago (**9**). Nine source countries have been chosen; all other sources are grouped under ‘other’. For sites with significant proportions of Class A debris from ‘other’ countries, the largest ‘other’ sources are as follows: Astove (Madagascar); Farquhar (Madagascar); Mauritius (Mauritius); Réunion (Réunion); Socotra (Yemen). For Class B debris: Mauritius (Mauritius); Réunion (Réunion); Socotra (Yemen). For Class C debris: Mauritius (South Africa and Mauritius); Réunion (Réunion and South Africa); Socotra (Yemen and Pakistan).

307 of Seychelles (particularly Mahé, the main population centre of Seychelles), but substantial  
308 proportions are also predicted to originate from Indonesia and, in the case of islands in the  
309 northernmost Seychelles Plateau (Denis and Bird islands), India and Sri Lanka. India and  
310 Sri Lanka are expected to still act as the main sources of debris for the relatively nearby  
311 island groups of Lakshadweep and Maldives, but the lower sinking rate and contributions  
312 from winds and waves during the northeast monsoon also results in these countries becoming  
313 significant sources of debris for Socotra, previously dominated by local sources from Yemen.

314

315 Finally, Class C debris (Figure 3(c)) beaching across Seychelles (and the Chagos Archipelago)  
316 is expected to originate almost entirely from the northern and eastern Indian Ocean. Indone-  
317 sia is still expected to be the largest single source country, but a significant proportion is  
318 swept from Philippines and, in the case of more northerly islands, India and Sri Lanka. Sey-  
319 chelles and East Africa are not significant sources of Class C debris for any sites in Seychelles.  
320 Our analyses also suggest that Mauritius and Réunion, dominated by local sources for less-  
321 buoyant classes of debris, receive significant quantities of Class C debris from South Africa  
322 (57% and 36% respectively).

323

324 We can also extract the predicted drift time distribution for debris accumulating at our  
325 study sites (shown for Aldabra in Supplementary Figure 5). Unsurprisingly, the more buoy-  
326 ant debris classes have a broader range of drifting times, where drifting times are stratified  
327 by the oceanographic distance of source countries from Aldabra. For instance, for Class C  
328 debris accumulating at Aldabra, debris arriving from Comoros and Tanzania have generally  
329 only been at sea for 1-2 months, whereas debris arriving from Indonesia has been at sea for at

330 least 6 months, with a small proportion exceeding 2 years. However, an important conclusion  
331 is that the distribution of drift times is complex and multimodal. Although Lagrangian back-  
332 tracking is considerably less computationally expensive than the approach used in this study,  
333 van Duinen et al. (2022) were required to make an *a priori* assumption for the drift time  
334 distribution of debris accumulating at their site of interest. These drift time distributions  
335 for Aldabra highlight that assuming a uniform age distribution of beaching debris is not an  
336 appropriate assumption for remote islands.

337

338 As further discussed in section 3.2, there is significant temporal variability in accumulation  
339 rates at many of these remote sites, particularly for Class A debris. However, recomputing  
340 Figure 3 for subsets of the full time-series suggests that our source attribution is robust for  
341 almost all sites (Supplementary Text 5).

### 342 3.1.2. Debris of marine origin

343 As with the terrestrial case, the probability of debris lost or discarded at sea eventually  
344 beaching at Seychelles strongly depends on the physical properties of the debris, and where  
345 it entered the ocean (Figure 4). Incoming Class A debris beaching at Aldabra (Figure 4(a))  
346 is sourced from a relatively narrow latitudinal band, due to primarily zonal currents around  
347 Aldabra. The Class A risk region for the Aldabra Group is almost entirely eastward of the  
348 island group, as these islands are in the path of a powerful westward-flowing ocean cur-  
349 rent (the North Madagascar Current). In contrast, the Class A risk map for the Seychelles  
350 Plateau (Supplementary Figure 8) is centred on the plateau due to the monsoonal reversal  
351 of prevailing zonal currents around the island group (Schott et al., 2009).

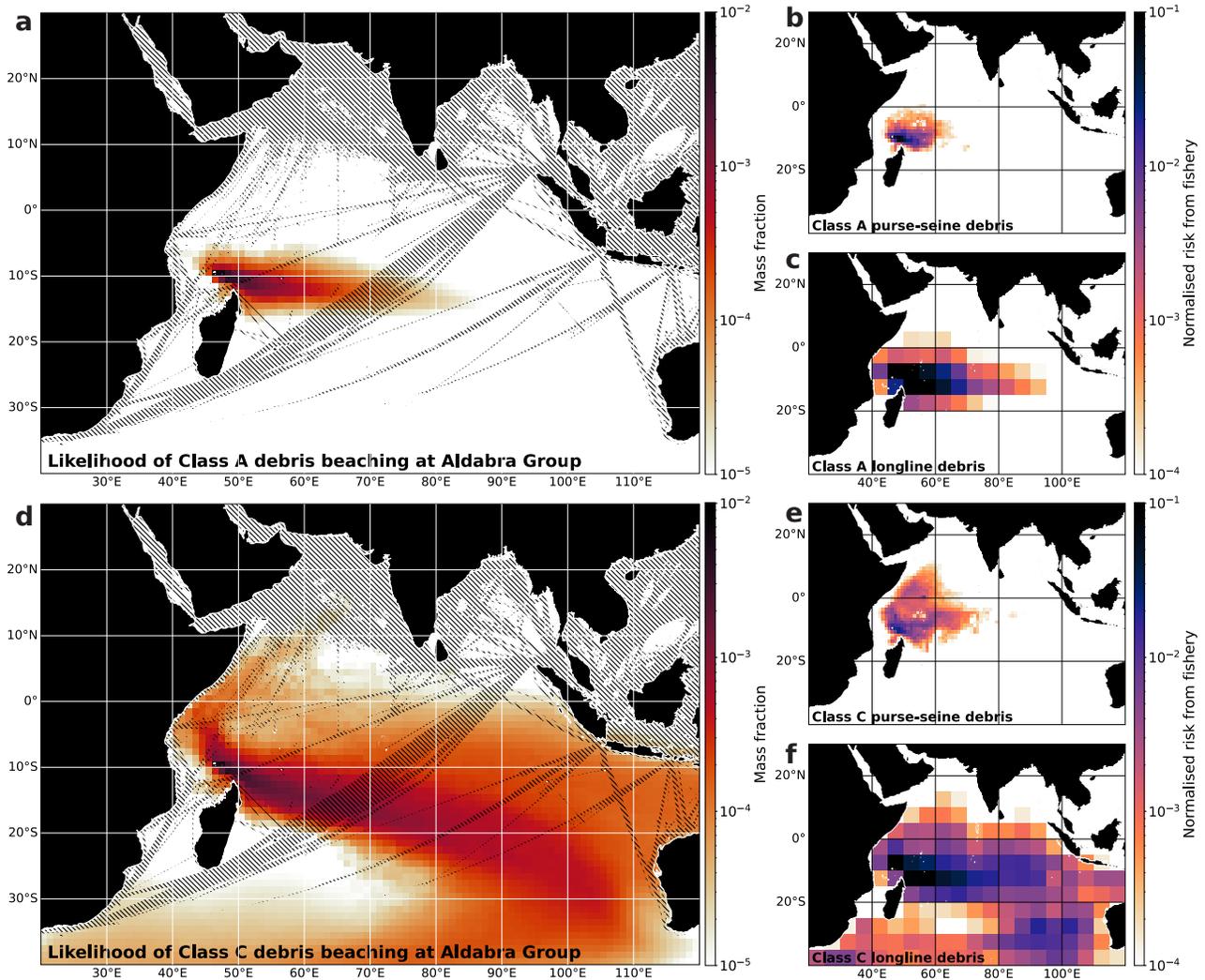


Figure 4: Risk map for Aldabra Group for (a) Class A debris and (d) Class C debris, showing the fraction of debris initialised per marine grid cell that beaches within the Aldabra Group. Hatching shows shipping corridors with the most intense traffic from Jan 2015 to Feb 2021 (Cerdeiro et al., 2020). Risk map for (b-c) Class A debris and (e-f) Class C debris from (b,e) purse-seines and (c,f) longlines ( $R_{ij}(x, y)$  from section 2.3.2). Corresponding plots for the Seychelles Plateau can be found in Supplementary Figures 8-11. Note the logarithmic scales in all panels.

353 With a significantly longer residence time at the ocean surface, and greater propulsion  
 354 due to windage, the risk maps for Class C debris (Figure 4(d)) cover a much greater area  
 355 than for Class A debris. For both the Aldabra Group and Seychelles Plateau, debris from  
 356 much of the tropical Indian Ocean has a non-negligible chance of beaching at one of these

357 island groups. Although much debris in the Indonesian archipelagic seas and further afield  
358 is removed through beaching within the narrow straits of the Indonesian Throughflow, the  
359 sheer quantity of mismanaged waste generated in Indonesia and Philippines allows a signifi-  
360 cant quantity to leak into the Indian Ocean.

361

362 The right-hand side panels in Figure 4 show predictions of the relative risk to the Aldabra  
363 Group of ALDFG associated with purse-seine and longline fisheries. In the case of purse-seine  
364 debris, due to the concentration of fishing effort around the Seychelles, our analyses suggest  
365 that most debris originates from the western Indian Ocean. The seas around the Outer  
366 Islands of Seychelles are associated with the highest risk, but our analyses suggest that for  
367 Class C purse-seine debris, there is still a non-negligible risk from fishing activity to the north  
368 and east of the Seychelles Plateau (Figure 4(e)). In contrast to purse-seine fisheries, effort  
369 associated with longline fisheries is more broadly distributed around the Indian Ocean. As a  
370 result, the footprint of the potential source region is much larger than for purse-seines. In the  
371 case of longline ALDFG behaving as Class C debris, whilst the highest risk regions are still  
372 in the southwestern Indian Ocean (around Seychelles and eastern Madagascar), debris could  
373 reasonably be sourced from as far afield as the southeastern Indian Ocean, west of Australia  
374 (Figure 4(f)). This suggests that a significant proportion of ALDFG beaching at Seychelles  
375 could originate from outside the Seychelles EEZ, particularly in the case of longline debris.

376

377 Finally, Figure 4 also shows that there is significant overlap between major high seas  
378 shipping lanes (hatching in Figure 4), and high risk regions for Seychelles. Even in the case  
379 of short-lived Class A debris, the major shipping lanes linking the Bay of Bengal and South

380 China Sea to the Atlantic pass within the high risk zone for the Aldabra Group. For Class C  
381 debris, most of the major shipping lanes in the Indian Ocean pass through regions associated  
382 with a high risk of beaching for both the Inner and Outer Islands of Seychelles, including  
383 Atlantic-bound connections from the Middle East and Java Sea, as well as those originating  
384 from the Bay of Bengal and South China Sea.

### 385 *3.2. Variability and drivers of beaching marine debris*

386 Despite the monthly input of terrestrial debris remaining constant in our analyses, there  
387 is substantial temporal variability in beaching rate predicted for remote islands. Figure 5(a)  
388 shows the mass of Class A debris beaching at Aldabra (Aldabra Group) and Praslin (Sey-  
389 chelles Plateau) per month from 1995 (two years after the first debris release) to 2014 (the  
390 last release year for terrestrial debris). Although the average accumulation rate for Class  
391 A debris at Praslin is substantially higher than for Aldabra, the monthly accumulation rate  
392 at Aldabra varies over 6 orders of magnitude and its peak (in 1995) exceeds any month at  
393 Praslin. These patterns are a result of the different principal sources of Class A debris for  
394 Aldabra and Praslin (Figure 3(a)). In the case of Praslin, most Class A debris is sourced  
395 from within Seychelles, largely from Mahé (around 50km away). The transport pathway from  
396 source to sink for Class A debris beaching at Praslin is therefore short ( $< 2$  weeks), which  
397 allows debris from within Seychelles to consistently beach at Praslin before sinking, with  
398 less of an opportunity for seasonal variations in ocean currents or eddy variability to disrupt  
399 this pathway. In contrast, most Class A debris beaching at Aldabra originates in Comoros  
400 and Tanzania, both of which are hundreds of kilometres away and are connected to Aldabra  
401 through low probability connections (Figure 4(a)). As a result, Aldabra sees almost no Class

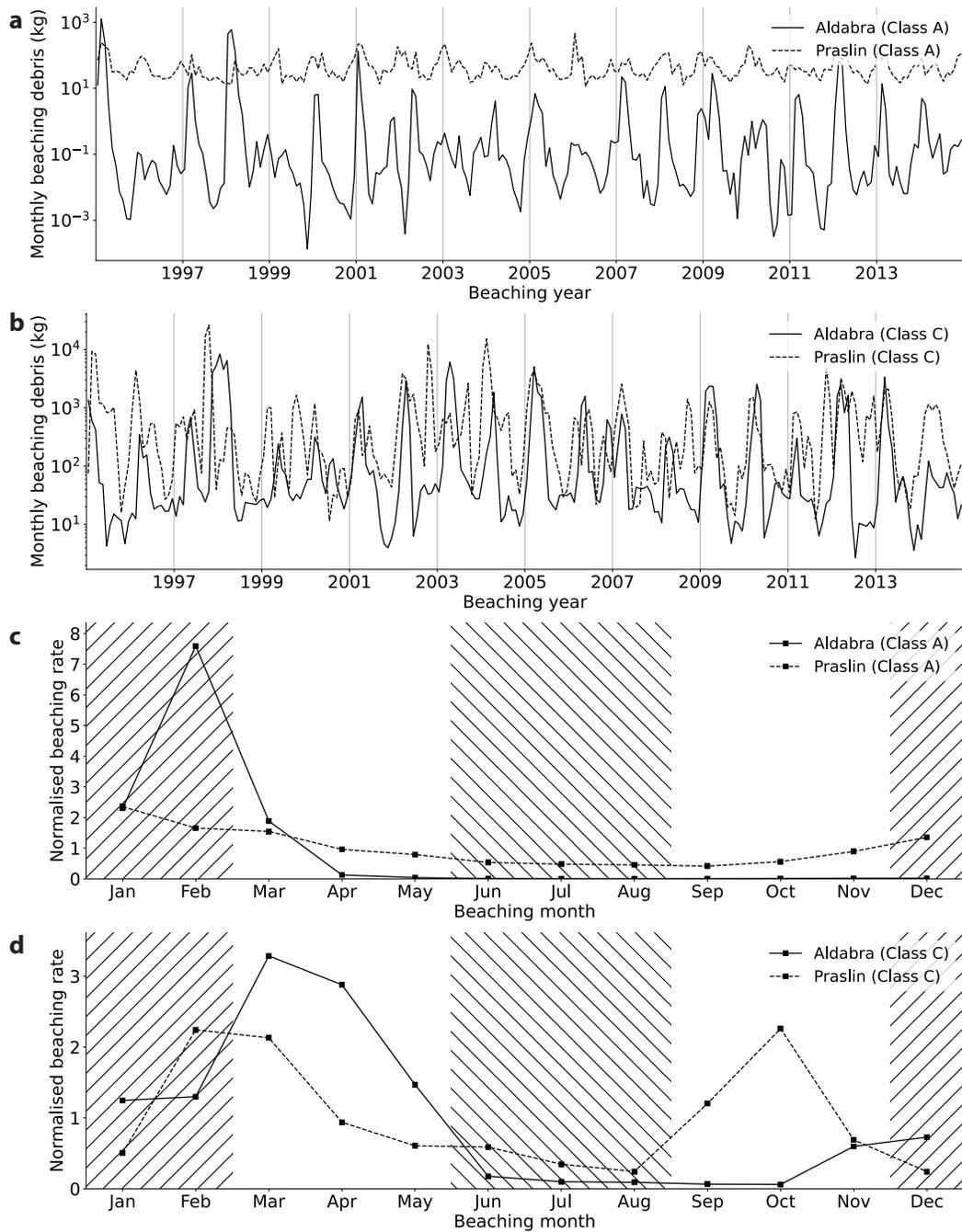


Figure 5: (a)-(b) Monthly beaching rate from 1995-2014 at Aldabra (Aldabra Group) and Praslin (Seychelles Plateau) assuming all terrestrial debris is (a) Class A and (b) Class C. (c)-(d) Monthly beaching rate averaged across 1995-2014 at Aldabra and Praslin (normalised by the annual mean) for (c) Class A and (d) Class C debris. The hatching indicates the approximate timing of the northeast monsoon ( $\sim$  December to February) and southwest monsoon ( $\sim$  June to August).

402 A debris beaching in most months, but if an eddy happens to direct a filament of Class A  
 403 debris towards Aldabra, a large amount of debris may beach in a short period of time. This

404 prediction is similar to patterns of ‘pulsed recruitment’ predicted for the long-distance larval  
405 dispersal of some marine organisms (e.g. Siegel et al., 2008).

406

407 In contrast, *both* Praslin and Aldabra see a similar level of variability in beaching rates  
408 for Class C debris (Figure 5(b)). Most Class C debris beaching at both islands originates  
409 from distal sources in southeast Asia and, in the case of Praslin, south Asia. Class C debris  
410 arrives at both islands through long-distance transport pathways, and there is therefore am-  
411 ple opportunity for these transport pathways to be controlled by stochastic, eddy-induced  
412 variability. The variability in accumulation rate at Aldabra is lower for Class C debris than  
413 for Class A, possibly because the wider geographic distribution of sources and greater time  
414 available for mixing ‘smooths out’ the distribution of marine debris in the ocean. Never-  
415 theless, monthly beaching rates for both islands are predicted to vary across three orders of  
416 magnitude, with most debris arriving during short periods of high accumulation rate.

### 417 *3.2.1. Seasonal variability*

418 Given this enormous variability in beaching rate, it is useful to understand whether beach-  
419 ing rate varies entirely stochastically, or whether there is some predictability (which could  
420 help with the organising of beach clean-ups and other marine debris management activities).  
421 In particular, prevailing winds and many currents in the Indian Ocean change direction  
422 following the monsoons<sup>1</sup>. These monsoons have previously been suggested to control the  
423 partitioning of debris between the southern and northern Indian Ocean (van der Mheen  
424 et al., 2020). Figures 5(c)-(d) show the monthly accumulation rate for Class A and C debris

---

<sup>1</sup>In this study, we refer to the monsoons around December to February, and June to August, as the northeast and southwest monsoons, respectively, in line with Schott et al. (2009).

425 arriving at Aldabra and Praslin, averaged over the interval 1995-2014. For Class A debris  
426 (Figure 5(c)), almost all debris beaches at Aldabra between January and March, i.e. the  
427 end of the northeast monsoon. The sharpness of this peak is partially due to extreme events  
428 in 1995 and 1998 (Figure 5(a)), but this seasonal cycle remains robust even without these  
429 years. Comoros is the largest source of Class A debris for Aldabra but ordinarily, debris  
430 from Comoros is swept into the Mozambique Channel and away from Aldabra. Rapid debris  
431 transport from Comoros to Aldabra relies on a relatively uncommon pathway in which de-  
432bris is entrained into eddies in the northern Mozambique Channel and transported towards  
433 Madagascar, before entering the North Madagascar Current upstream of Aldabra, and sub-  
434sequently beaching. This pathway is improbable during the southwest monsoon as a strong  
435 North Madagascar Current Backeberg and Reason (2010) results in debris rapidly beaching  
436 along the east African coast. As a result, transport from Comoros to Aldabra is generally only  
437 feasible during the northeast monsoon and subsequent intermonsoon. There is some seasonal  
438 variability at Praslin, with higher Class A beaching rates during the northeast monsoon,  
439 but considerably less than Aldabra. During the northeast monsoon, the South Equatorial  
440 Countercurrent shifts towards the south near the Seychelles Plateau (Schott et al., 2009),  
441 facilitating the eastward transport of debris from the highly populated island of Mahé to-  
442wards Praslin. Conversely, the South Equatorial Countercurrent shifts to the north during  
443 the southwest monsoon, and debris is more likely to be transported westward from Mahé  
444 due to the northwestward Stokes drift over the Seychelles Plateau at this time. Indeed, the  
445 seasonal pattern for Class A debris beaching at Silhouette Island, west of Mahé, is in exact  
446 antiphase to the pattern at Praslin (see Supplementary Table 1).

447

448 In the case of Class C debris, both Aldabra and Praslin see a peak in beaching rate dur-  
449 ing the late northeast monsoon and subsequent intermonsoon (Figure 5(d)). In the case of  
450 Aldabra, this peak is due to debris from Indonesia, whereas the peak at Praslin is due to  
451 debris arriving from India and Sri Lanka. Instead, Praslin has a second Class C beaching  
452 peak following the southwest monsoon, which is driven by debris from Indonesia. However,  
453 this peak is deceptive. Contrary to Aldabra, which has a clearly defined peak attributable  
454 to Indonesia at approximately the same time in almost all model years, the time-mean peak  
455 attributable to Indonesia at Praslin in Figure 5(d) is actually driven by a small number of  
456 outlier events, most significantly in 1997. Although the time-integral source attribution data  
457 presented in Figure 3 are generally robust with respect to simulation timespan, it is clear  
458 through inspection that interpreting temporal variability in beaching rate is not as straight-  
459 forward.

460  
461 Alternatively, we can observe that, in most years, log-transformed beaching rates are  
462 dominated by a single clear sinusoidal peak at most sites we considered (e.g. Figure 5(a)).  
463 By analysing beaching rates in the frequency domain and extracting the phase of the compo-  
464 nent with a period of 1 year, we can estimate during which season beaching rates *consistently*  
465 peak. This is summarised for Class C debris in Table 1 (corresponding tables for Class A and  
466 Class B debris are given in the Supplementary Tables 1-2). To verify whether the assumption  
467 of a single clear beaching rate peak per year is valid, we computed the correlation between  
468 the actual (log) beaching rate, and the idealised beaching rate using only the annual compo-  
469 nent of the Fourier spectrum. This correlation was significant ( $p < 0.01$ , taking into account  
470 autocorrelation within both the modelled and seasonal time-series (Bretherton et al., 1999))

471 for almost all islands considered.

472

473 Table 1 suggests that the seasonality of beaching rates across Seychelles is actually in  
474 phase for Class C debris of terrestrial origin, with a significant peak predicted in March or  
475 April for almost all Seychellois islands (i.e. the end of the northeast monsoon and the sub-  
476 sequent intermonsoon). This peak shifts slightly earlier in the year for less buoyant classes,  
477 but remains during the northeast monsoon for Class A and Class B debris beaching at  
478 most islands in Seychelles. The strength of this seasonality (quantified by the ratio of the  
479 beaching rate during the highest and lowest three months), however, is considerably larger for  
480 the Outer Islands of Seychelles, particularly for the Aldabra, Farquhar and Alphonse Groups.

481

482 We can gain further insight into the physical drivers of this seasonality by repeating the  
483 same spectral analysis for all source grid cells in the marine-release experiments. Plotted  
484 in Figure 6(a) is the correlation between the beaching time at the Aldabra Group of debris  
485 released across the Indian Ocean, and the seasonal cycle identified for that cell from the  
486 Fourier Spectrum. Although our analyses suggest that Indonesia is the dominant source of  
487 Class C debris for Aldabra, Figure 6(a) shows that debris beaching at the Aldabra Group is  
488 significantly correlated ( $p < 0.01$ ) with the seasonal cycle for most source regions across the  
489 Indian Ocean, as well as the Indonesian archipelagic and Chinese marginal seas. The phase  
490 of this seasonal cycle is given in Figure 6(b), revealing that this entire region is perfectly in  
491 phase. This may be surprising, as the drift time to the Aldabra Group varies considerably  
492 across the Indian Ocean. If the seasonality of Class C debris beaching at the Aldabra Group  
493 depended on remote forcing (i.e. currents, winds and waves at the debris source region, or

Beaching site	Seasonal cycle peak	Seasonality strength
Aldabra	March	35.1
Assomption	March	36.7
Cosmoledo	March	35.1
Astove	March	35.2
Providence	March	48.3
Farquhar	March	47.4
Alphonse	March	39.0
Poivre	April	13.6
St Joseph	March	10.9
Desroches	April	8.7
Platte	March	6.3
<i>Coëtivy</i>	<i>March</i>	<i>20.1</i>
Mahé	March	5.2
Fregate	April	6.3
Silhouette	April	10.5
Praslin	March	8.0
Denis	April	5.7
Bird	April	6.9
Comoros	December	1.9
Mayotte	January	18.7
Lakshadweep	February	92.4
Maldives	February	10.7
<i>Mauritius</i>	<i>August</i>	<i>1.9</i>
<i>Réunion</i>	<i>November</i>	<i>1.3</i>
Pemba	January	4.4
Socotra	March	12.7
Chagos Archipelago	September	7.9

Table 1: Class C debris beaching rate seasonal peak, and strength of the seasonal cycle (1995-2014), based on the phase of the component of the Fourier spectrum with period 1 year. The strength of the seasonal cycle is the ratio of the mean beaching rate during the three months with the highest beaching rate, and the three months with the lowest beaching rate. All time series correlated significantly with idealised cycle ( $p < 0.01$ ) aside from sites in *italics*.

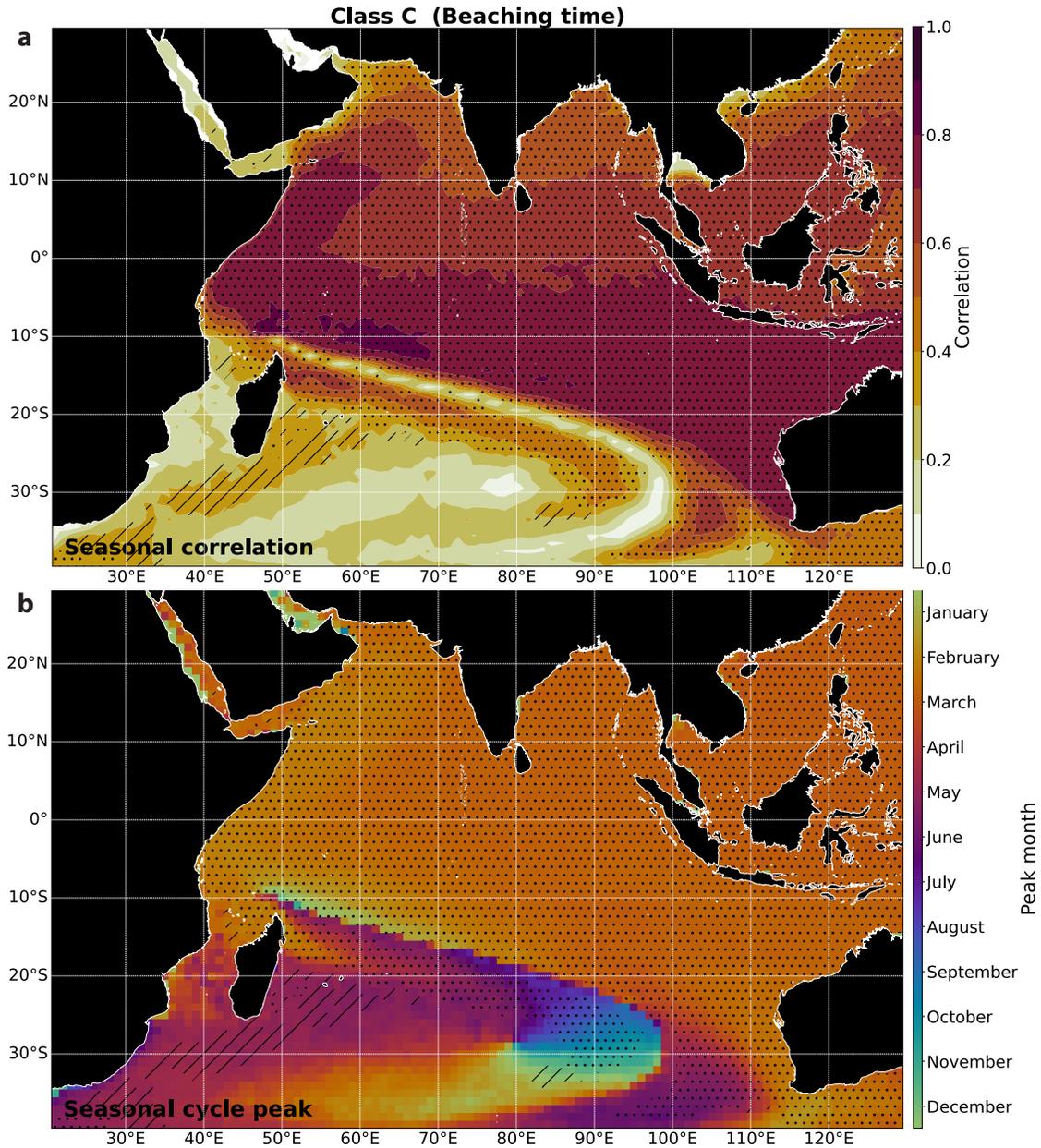


Figure 6: (a) Correlation between (log-transformed) time-series of debris beaching at the Aldabra Group from each cell, and the idealised seasonal cycle extracted from the Fourier spectrum. Shading indicates that the time-series in a cell correlates with the seasonal cycle significantly,  $p < 0.01$  (dotted) and  $p < 0.05$  (hatched), taking into account autocorrelation within both the modelled and seasonal time-series (Bretherton et al., 1999). (b) Phase of the seasonal cycle extracted from the Fourier spectrum, in terms of the seasonal cycle peak.

494 along its transport path), we would expect considerable spatial heterogeneity in Figure 6(b).

495

496 Instead, this figure demonstrates that the seasonality of Class C debris beaching at the

497 Aldabra Group is dominated by *local* forcing, specifically the monsoonal variation in the  
498 winds. During the northeast monsoon, winds around Aldabra are relatively weak (Figure  
499 2(a)) and westward zonal surface currents are proportionately more important. As a re-  
500 sult, debris arriving at Aldabra during the northeast monsoon is sourced, on sub-seasonal  
501 timescales, east of Aldabra, from the southern tropical Indian Ocean. Conversely, during the  
502 southwest monsoon, strong southeasterly winds blow over the Aldabra Group (Figure 2(b))  
503 and the source region for Aldabra (on sub-seasonal timescales) shifts to the *southeast* of  
504 Aldabra, in the southern *subtropical* Indian Ocean. Crucially, since winds over the southern  
505 Indian Ocean never have a strong northerly component (Figure 2), there is *no efficient path-*  
506 *way* for Class C marine debris to reach the subtropical southern Indian Ocean from Indonesia  
507 (or other south(east) Asian sources), and therefore no route to Aldabra. As a result, it is  
508 improbable for Class C debris from the eastern or northern Indian Ocean to reach Aldabra  
509 during the southwest monsoon. This wind-driven mixing barrier can be clearly seen in Fig-  
510 ure 6(b) as the sharp phase discontinuity extending southeastwards from the Aldabra Group.

511

512 In this way, the monsoonal winds over the Aldabra Group act as a debris ‘switch’, al-  
513 ternating the principal debris source between the southwestern Indian Ocean (with minimal  
514 debris sources), and the remainder of the basin. The dominance of winds over the seasonal-  
515 ity of beaching at the Aldabra Group remains valid for Class B debris, but not for Class A  
516 debris (0% windage), where the seasonality instead appears to be dominated by the strength  
517 and position of the North Madagascar and South Equatorial Currents. The phase of the  
518 seasonal cycle with respect to the Seychelles Plateau (Supplementary Figure 17) is similar  
519 to the Aldabra Group, but due to the more northerly position of the Inner Islands, winds

520 associated with the southwest monsoon do not have as extreme a blocking effect as with the  
521 Aldabra Group. Additionally, as hinted at by the greater spatial heterogeneity in Supple-  
522 mentary Figure 17, remote forcing may play a greater role for debris beaching at the Inner  
523 Islands. For instance, there is a fairly direct transport pathway from India and Sri Lanka  
524 to the Seychelles Plateau during the northeast monsoon due to the northeasterly winds and  
525 westward-flowing Northeast Monsoon Current south of India, whereas these winds and cur-  
526 rents reverse during the southwest monsoon.

527

528 At some remote islands, such as Aldabra, most beaching debris is actually related to  
529 fishing activities rather than terrestrial input (Burt et al., 2020). As a result, the seasonal  
530 patterns identified for Aldabra may not necessarily be the same for fishing-related debris  
531 due to the very different input distribution to debris from the coasts. However, by comput-  
532 ing monthly beaching rates for ALDFG ( $B_{ij}(t_b)$  from section 2.3.2), we find that, although  
533 peaks are not perfectly aligned with predictions for debris of terrestrial origin, purse-seine  
534 and longline associated debris beaching at the Aldabra Group will still likely peak during the  
535 northeast monsoon or subsequent intermonsoon, and fall to a minimum during the south-  
536 west monsoon (Figure 7(a)-(b)). As a result, although there may not be a clearly defined  
537 peak of debris accumulation at Aldabra in March as suggested by Table 1, we would still  
538 expect debris accumulation to be significantly enhanced during the northeast monsoon and  
539 subsequent intermonsoon, as compared to the southwest monsoon. For fishery-related debris  
540 accumulating at sites across the Seychelles Plateau, our analyses suggest that the seasonal  
541 cycle would be similar, but slightly broader and shifted later in the year. This may be due  
542 to the more central position of the Seychelles Plateau with respect to intensive fisheries in

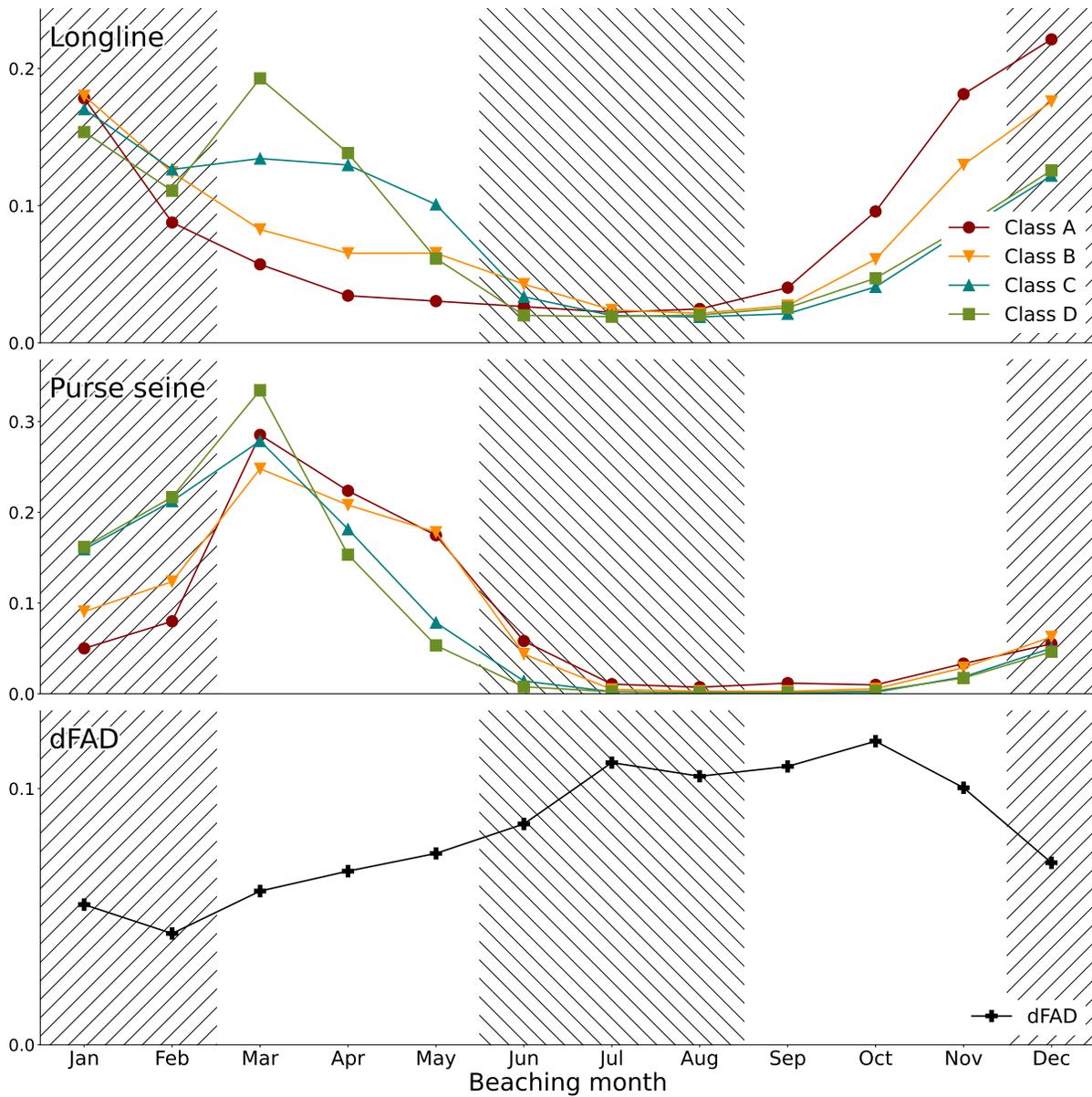


Figure 7: (a)-(b) Monthly beaching rate from 1995-2012 at the Aldabra Group for debris related to (a) longlines and (b) purse-seines, for Class A-D debris. (c) Predicted monthly beaching rate of dFADs, assuming they are not affected by winds or Stokes drift, i.e. follow physical scenario **C0** (Imzilen, 2021). Supplementary Figure 19 is the analogous plot for the Seychelles Plateau.

543 the western Indian Ocean, as well as the seasonality of fishing activities in the region, which  
 544 is incorporated into these analyses.

545

546 *3.2.2. Interannual variability*

547 Although our analyses suggest that temporal variability in beaching rates at remote is-  
548 lands in the western Indian Ocean is dominated by seasonal variability from the monsoons,  
549 there is still considerable interannual variability. This is most extreme in the case of the  
550 short-lived Class A debris, where for some islands the majority of beached debris arrived  
551 during a small number of debris pulses. However, even in the case of the more predictable  
552 and long-lived Class C debris, inspection of Figure 5(b) demonstrates that substantial year-  
553 to-year variability remains. Northerly wind anomalies across the southern Indian Ocean  
554 are associated with IOD and ENSO events (Yu et al., 2005) and, as described in Section  
555 3.2.1, the meridional component of winds over the southern Indian Ocean associated with  
556 the monsoons appear to be driving the seasonal cycle in beaching rates across Seychelles.  
557 We may therefore expect IOD and ENSO phases to amplify the seasonal cycle simulated  
558 for Seychelles, amplifying northeast monsoon beaching rates for debris from southeast Asia  
559 during positive phases, and further suppressing southwest monsoon beaching rates during  
560 negative phases.

561

562 To test this, we passed time-series of marine debris beaching rates through a low-pass  
563 filter with a cutoff frequency of 1.25 years, to remove intra-annual variability from the signal.  
564 We then carried out a lagged correlation of the filtered time-series against the Dipole Mode  
565 Index (DMI), an IOD index based on SST gradients across the equatorial Indian Ocean,  
566 and NINO3.4, an ENSO index based on mid-Pacific SST. Figure 8(a) shows an analogue of  
567 Figure 6(a) based on correlations of Class C debris beaching rates at the Aldabra Group with  
568 DMI. Although correlations are unsurprisingly lower than for the seasonal cycle, interannual

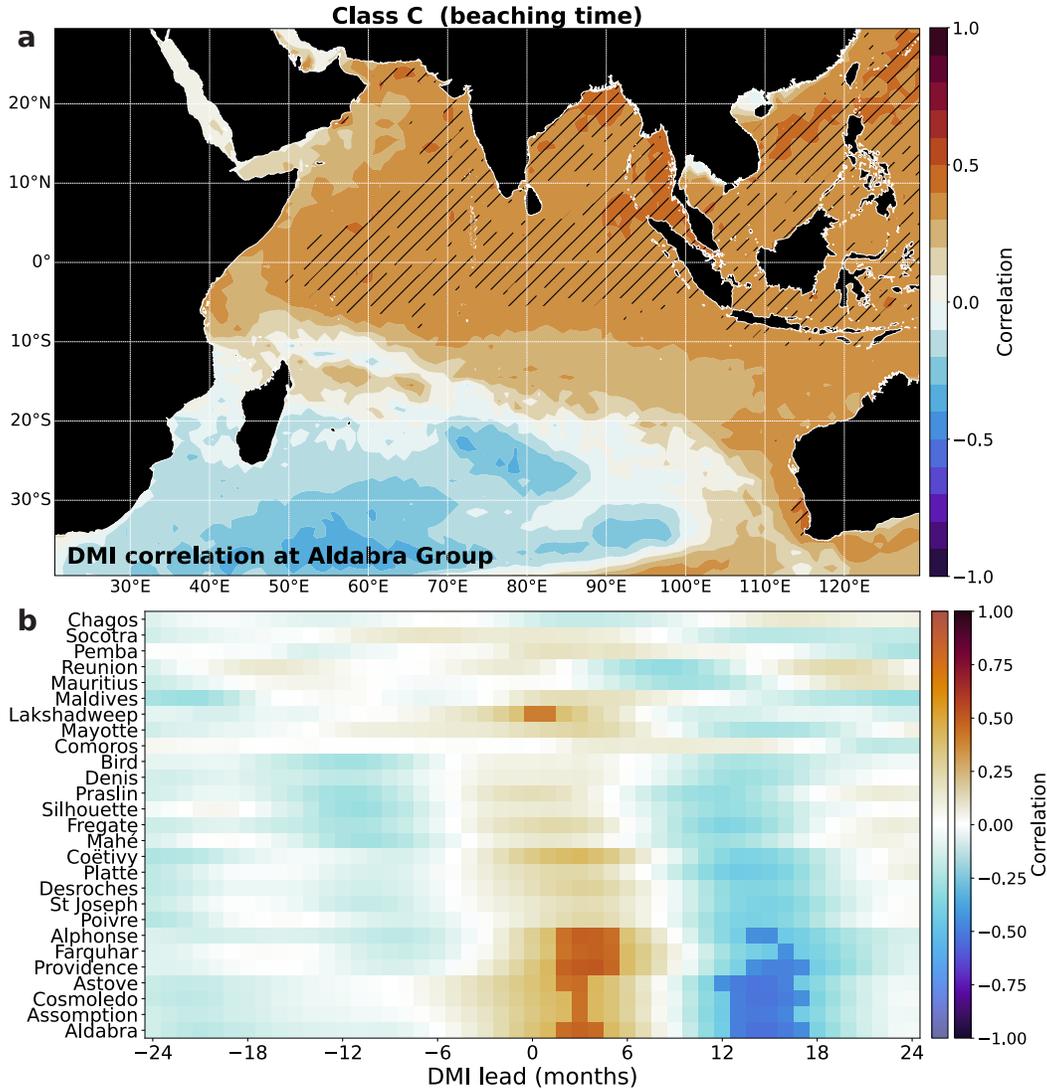


Figure 8: (a) Correlation between (log-transformed) time-series of debris beaching at the Aldabra Group from each cell, and the IOD Dipole Mode Index (DMI). Shading indicates that the time-series in a cell correlates with DMI significantly,  $p < 0.01$  (dotted) and  $p < 0.05$  (hatched), taking into account autocorrelation within both the modelled and DMI time-series (Bretherton et al., 1999). (b) Correlation between the (log-transformed) time-series of debris beaching at each site investigated in this study, and DMI, as a function of DMI lead time (months). Correlations significant to  $p < 0.01$  are shown in bolder colours (the second colour bar).

569 variability in Class C beaching rates at Aldabra are correlated with DMI for source sites  
 570 across much of the north and northeastern Indian Ocean. Additionally, the spatial pattern  
 571 of these correlations strongly resembles the pattern in Figure 6(a) from the seasonal cycle,  
 572 supporting the hypothesis that the IOD may amplify the seasonal cycle through modulation

573 of meridional winds in the southern Indian Ocean. Figure 8(b) shows correlations between  
574 the total Class C beaching rate at all sites considered in this study, and DMI, as a function  
575 of DMI lead time. DMI correlates significantly with beaching rates at islands in the Aldabra,  
576 Farquhar, and Alphonse groups, which is expected as these are the same island groups that  
577 saw the most dramatic modulation by the seasonal cycle. DMI also correlates most strongly  
578 with beaching rates with a lead time of a few months, which supports the hypothesis that the  
579 IOD modulates the seasonal cycle as the monsoonal winds also lead peak Class C beaching  
580 rates (the seasonal cycle peaks in Table 1 within Seychelles generally occur just after the  
581 northeast monsoon, during the subsequent intermonsoon).

582

583 Correlation with the NINO3.4 index actually returns higher correlation coefficients com-  
584 pared to DMI, which is consistent with the partial correlations with the surface wind field  
585 given in Yu et al. (2005), as ENSO appears to be associated with stronger meridional wind  
586 anomalies closer to Aldabra. However, due to the longer autocorrelation timescale within the  
587 NINO3.4 time-series, the correlation of the NINO3.4 index with beaching rates at our study  
588 sites was not significant ( $p > 0.01$ ).

589

### 590 *3.3. Comparison with observations*

#### 591 *3.3.1. Marine debris accumulation at Aldabra*

592 The two parameters in our analyses describing the sinking rate and beaching rate,  $\mu_s$   
593 and  $\mu_b^*$ , are highly uncertain, particularly  $\mu_s$ . Fazey and Ryan (2016) estimated sinking  
594 timescales for polyethylene (LDPE and HDPE) fragments ranging from 0.5-5cm in size and,

595 whilst the statistical model used in their study is not identical to the statistical model used for  
596 our sinking parameterisation, they estimated sinking timescales on the order of 17-66 days.  
597 Kaandorp et al. (2020) predicted a slightly higher sinking timescale of  $1/\mu_s = 81$  days based  
598 on an inverse model incorporating observations of floating debris in the Mediterranean Sea.  
599 Koelmans et al. (2017) predicted an effective removal timescale of marine debris from the  
600 ocean surface (through fragmentation into microplastics) on the order of months, based on  
601 mass-balance arguments and observations of floating debris. However, Lebreton et al. (2019)  
602 argued that observations of the age distribution of debris in the North Pacific subtropical  
603 gyre are inconsistent with rapid sinking rates, instead suggesting that observations are more  
604 consistent with low sinking rates and rapid scavenging of debris at coastlines through beach-  
605 ing.

606

607 There are very few observational estimates for marine debris beaching rates. Dunlop  
608 et al. (2020) carried out repeat beach surveys at Cousine Island, Seychelles, from 2003-2019,  
609 and estimated accumulation rates. However, they calculated accumulation rate in terms  
610 of number of items rather than mass, so these results cannot be directly compared to our  
611 model output. However, Burt et al. (2020) carried out a five-week clean-up on Aldabra, Sey-  
612 chelles, and estimated that 513.4 tonnes of debris had accumulated on the island, of which  
613 87.3 tonnes was terrestrial in origin. Annual emissions of marine debris into the ocean have  
614 increased over time, but our numerical model assumes constant annual debris emissions at  
615 2015 levels. We estimate that the 87.3 tonnes of terrestrial debris that has accumulated at  
616 Aldabra corresponds to an annual beaching rate of around 2.9-5.3 tonnes per year, assuming  
617 no losses (see Supplementary Text 6).

618

619 Calculating the average annual beaching rate at Aldabra across  $\mu_b^*$ - $\mu_s$  parameter space  
620 (from 1999-2014 to allow a longer spin-up for lower values of  $\mu_s$ ) reveals (1) that the beaching  
621 rate at Aldabra is insensitive to beaching rate in the interval  $1/\mu_b^* \in [5, 60]$ , and (2) that the  
622 inferred average bulk beaching flux at Aldabra is most consistent with  $100\text{d} < 1/\mu_s < 400\text{d}$ ,  
623 depending on the windage coefficient (Supplementary Figures 20-22). This is not to suggest  
624 that all marine debris has a sinking rate in this range ( $\mu_s$  is a variable which will likely depend  
625 on debris composition, geometry, and biofouling rates), but it does indicate that most debris,  
626 by mass, is likely to have a sinking rate on the order of months to a year. This is consistent  
627 with the findings of Fazey and Ryan (2016), Kaandorp et al. (2020), and Koelmans et al.  
628 (2017). As a result, Class A debris is probably not going to represent a significant fraction  
629 of debris beaching at Aldabra by mass. Additionally, it is also unlikely that most debris  
630 beaching at Aldabra has a sinking timescale of multiple years, since we would expect a  
631 significantly greater mass of terrestrial debris to have accumulated on Aldabra if this were  
632 the case.

### 633 *3.3.2. Temporal variability of marine debris beaching across Seychelles*

634 Drifting Fish Aggregating Devices (dFADs) are buoyant drifters used primarily by purse-  
635 seine fisheries to aggregate tuna. The majority of these dFADs are tracked remotely using  
636 satellite-transmitting GPS-equipped buoys and as a result, dFADs are one of the only types  
637 of marine debris that can be tracked directly from source to sink Imzilen et al. (2021).  
638 Macmillan et al. (2022) identified over 3000 dFAD beaching events across Seychelles, and  
639 analysed beaching rates and seasonality. This provides a useful test-case for our trajectory

640 analysis, but the physics of dFAD transport do not correspond well to any of our marine  
641 debris classes A-C; the long drogue attached to dFADs reduces the effects of windage and  
642 Stokes drift, and a previous study found that the incorporation of windage reduces the skill  
643 score of dFAD trajectory prediction (Imzilen, 2021). As a result, we define ‘dFADs’ as a new  
644 Class of marine debris with  $\mu_s = 1800\text{d}$  (dFADs are large, buoyant, and non-biodegradable),  
645  $\mu_s = 30\text{d}$ , and physical scenario **C0** (surface currents only. Imzilen (2021)). We compute  
646 the predicted seasonal distribution of dFAD beachings based on the methodology described  
647 in section 3.2.1, taking into account the seasonality of dFAD deployments. Our simulations  
648 reproduce a relatively muted seasonal cycle of dFAD deployments at Aldabra (Figure 7(b))  
649 and a pronounced peak in dFAD beaching rates within the Seychelles Plateau during the  
650 intermonsoon following the northeast monsoon, both of which correspond well to observations  
651 (Isla MacMillan, *personal communication*).

### 652 *3.3.3. Sources of debris at remote islands in the western Indian Ocean*

#### 653 **Aldabra (Seychelles)**

654 In addition to quantifying the total mass of debris on Aldabra, Burt et al. (2020) also iden-  
655 tified the origin of 45 PET bottles with intact labels. In Table 2, we compare the predicted  
656 distribution of countries of origin for Class C debris beaching at Aldabra, to the distribution  
657 of countries of manufacture for intact PET bottles found at Aldabra.

658

659 For several countries of origin, there is agreement between the two datasets. Of the 5 largest  
660 sources of debris predicted by the model, bottles were found on Aldabra from 3 (Indonesia,  
661 India, and South Africa). Indonesia was the largest source of (Class C) debris in our model,

Origin	This study (%)	Burt et al. (2020) (%)
Indonesia	50.6	13.3
Philippines	21.2	-
India	5.3	6.7
South Africa	5.0	2.2
Comoros	4.4	-
Tanzania	3.7	-
Sri Lanka	2.4	-
Timor-Leste	1.7	-
Malaysia	1.6	6.7
Thailand	0.5	8.9
China	<0.1	46.7
Singapore	<0.1	4.4

Table 2: Distribution of countries of origin or manufacture from this study (based on Class C debris) and the sample of 45 PET bottles with intact labels from Burt et al. (2020). Only countries associated with at least 1% of accumulated debris (this study) or at least 1 bottle (Burt et al., 2020) are included.

662 and was the second largest country of manufacture in the sample from Aldabra. However,  
663 there are also some significant differences. This in itself is not unexpected. For instance,  
664 the sample size (45) of PET bottles in Burt et al. (2020) is small, and the sample is likely  
665 biased against bottles with longer drift times, as only bottles with intact labels could have  
666 their country of manufacture identified. Additionally, the country of manufacture of a bot-  
667 tle is not necessarily the same as the country where a bottle entered the ocean. However,  
668 the particular countries associated with model-observation disagreement provide interesting  
669 insights into the sources of debris for Aldabra.

670

671 The most obvious discrepancy between the two datasets is China. In our analysis,  
672 China was responsible for a negligible proportion of Class C debris accumulating at Aldabra  
673 (<0.1%), but was responsible for the manufacture of almost half of all bottles actually found  
674 on Aldabra. Although our Class C debris may be an imperfect representation of the physics

675 driving PET bottle transport, no realistic combination of  $\mu_b^*$ ,  $\mu_s$ , or physical scenario results  
676 in a significant flux of marine debris from China to Aldabra. More likely is an explanation  
677 suggested by Duhec et al. (2015), that a large proportion of labelled items from Asia accu-  
678 mulating at beaches in Seychelles were thrown overboard or lost from shipping activities in  
679 the vicinity of Seychelles. Indeed, this is strongly supported by Figure 4(d), which shows  
680 that Aldabra is directly downstream of the extremely busy shipping lanes linking SE Asia  
681 to the Atlantic. This same explanation could account for the number of bottles found on  
682 Aldabra from Thailand and Singapore, both of which were more than an order of magnitude  
683 more abundant in the cleanup than our predictions based on trajectory analysis. Shipping  
684 lanes aside, another possibility is that some waste entering the ocean from countries such as  
685 Indonesia was manufactured abroad. This could be due to the export of goods for sale and/or  
686 the export of waste. Indonesia is a major waste importer, but the main export partners are  
687 in Europe and the Americas (Greenpeace East Asia, 2019), so this cannot account for the  
688 discrepancies in Table 2. We do not have data on the proportion of bottled drinks sold in  
689 Indonesia (or other identified source countries) which are foreign imports, but imports would  
690 have to account for almost all PET bottles sold in these countries to explain the discrepancies  
691 in Table 2. We therefore suggest that disposal at sea is the most likely explanation for the  
692 discrepancies we have found.

693

694 Disposal at sea cannot, however, explain the under-representation of bottles from Philip-  
695 pines amongst PET bottles found at Aldabra relative to our predictions. We suggest the most  
696 likely explanation is that the value for  $\mu_b^*$  diagnosed from drifters based on a global dataset  
697 is inappropriate for the complex archipelagic coastline and bathymetry around Philippines,

698 resulting in our analyses underestimating the beaching rate for debris of Philippine origin.

699

### 700 **Alphonse (Seychelles)**

701 Dunlop et al. (2020) carried out a short-term marine debris monitoring program in 2013 at  
702 Alphonse and attempted to identify the country of origin for plastic and glass bottles (and  
703 caps) with intact labels. Dunlop et al. (2020) found that 75% of labelled items originated  
704 from Southeast Asia (primarily Indonesia and Thailand, although two glass bottles were  
705 found from Philippines), with 13% originating from East Asia (mainly China). Our model  
706 predicts that 46.5% of Class C debris beaching at Alphonse should originate from Indonesia  
707 and 13.5% from Philippines so, as with Aldabra, there is general agreement that a large pro-  
708 portion of beaching debris of terrestrial origin at Alphonse originates from Southeast Asia.  
709 As with Aldabra, Dunlop et al. (2020) found significantly more bottles of Chinese origin than  
710 predicted by our analysis, supporting their conclusion that these bottles were likely lost at  
711 sea relatively close to Alphonse. One interesting discrepancy is that our trajectory analysis  
712 predicts 30.0% of Class C debris at Alphonse originated from India or Sri Lanka, whereas  
713 Dunlop et al. (2020) did not identify any bottles from either of these countries. A difference  
714 in transport time cannot explain this difference, as the mean transport time from India and  
715 Sri Lanka to Alphonse should be less than that for debris of Indonesian origin.

716

### 717 **Outer Islands of Seychelles**

718 Based on a sample of 189 labels found on four islands in Seychelles (Alphonse, Coëtivy,  
719 Astove, and Platte), The Ocean Project Seychelles (2019) found that 49% of labels originated  
720 from SE Asia and specifically noted that the most common countries of origin were Indonesia

721 (26.5%), Mauritius (12%), and Malaysia (10.2%). This is broadly in line with the findings  
722 of the other debris monitoring programmes in Seychelles, although one exception is the large  
723 proportion of debris originating from Mauritius. Mauritius sits just to the south of the  
724 bifurcation point of the Southern Equatorial Current as it splits into the North Madagascan  
725 Current and Southeast Madagascan Current (Voldsund et al., 2017) and, as a result, debris  
726 from Mauritius is unlikely to be transported towards Seychelles, even accounting for the  
727 effects of winds. Given that no other studies assessing sources of debris in Seychelles noted  
728 a large proportion of debris from Mauritius ( $< 4\%$  at Alphonse (Duhec et al., 2015), and  
729 no mention at Aldabra (Burt et al., 2020)), it is possible that these items from Mauritius  
730 instead originated from nearby ships.

#### 731 **4. Conclusions and implications for conservation**

732 Environmental conservation NGOs have been burdened with the task of cleaning up vast  
733 quantities of marine debris arriving on coastlines across Seychelles and other small island  
734 developing states. Observations have suggested that most of these states are not responsi-  
735 ble for the bulk of debris accumulating on their shores, but limited quantitative data are  
736 available on sources, hindering management of the issue through source interventions and  
737 pursuing the ‘polluter pays principle’. We have provided the first quantitative estimates for  
738 the sources of marine debris (both terrestrial and marine in origin) across Seychelles, as well  
739 as other remote islands in the western Indian Ocean.

740

741 Our analyses suggest that Seychelles is a hotspot for marine debris accumulation from  
742 around the Indian Ocean. We estimate that a large proportion of debris beaching at Sey-

743 chelles has drifted from southeast Asia (principally Indonesia) and, in the case of the Inner  
744 Islands, south Asia (primarily India and Sri Lanka). Since debris drifting from sources such as  
745 Indonesia will have been at sea for at least six months, this also increases the risk of invasive  
746 species and pathogen introductions through rafting from the eastern and northern Indian  
747 Ocean. These results emphasise the scale of the challenge facing small island developing  
748 states such as Seychelles, and underlines the need for multilateral discussions around waste  
749 management. Smaller debris fragments may originate from East Africa (mainly Tanzania)  
750 and from within Seychelles itself, although these are unlikely to account for most beaching  
751 debris by mass, particularly for the Outer Islands of Seychelles. Our results suggest that  
752 Seychelles as a whole is at very high risk from debris that has been lost from ships transiting  
753 the Indian Ocean, and that most debris accumulating at Seychelles from Malaysia, Thailand  
754 and, in particular, China, is likely associated with these shipping corridors. This prediction  
755 could be used to initiate discussions with shipping companies to reduce these sources of ma-  
756 rine pollution. We have also found that abandoned, lost or otherwise discarded fishing gear  
757 has a high probability of beaching within Seychelles, directly polluting island ecosystems.  
758 Beaching purse-seine fragments are likely associated with fishing activity around Seychelles,  
759 but longline fragments could feasibly drift from fisheries across the southern Indian Ocean.  
760 Greater enforcement by regional governments of MARPOL Annex V (Marine Environment  
761 Protection Committee, 2017), forbidding the discharge of fishing gear and other plastics at  
762 sea, would reduce these sources of pollution, particularly for Aldabra.

763

764 We have also found that there is likely to be significant predictability in marine debris  
765 accumulation rates across Seychelles, primarily from a strong seasonal cycle controlled by the

766 monsoons. For classes of debris experiencing a significant push from the winds, our analysis  
767 suggests debris from terrestrial sources and fisheries are most likely to beach at Seychelles  
768 (but most significantly the Aldabra, Farquhar, and Alphonse Groups) during the northeast  
769 monsoon and subsequent intermonsoon. Beach clean-ups should ideally take place after peak  
770 beaching (i.e. May to June for much of Seychelles) to reduce the likelihood of beached plas-  
771 tics breaking down into smaller unmanageable fragments and impacting ecosystems. We have  
772 also proposed a mechanism by which ENSO and the IOD may modulate this seasonal cycle,  
773 and have presented some evidence to suggest that marine debris beaching rates at the more  
774 southerly island groups within Seychelles may be greater during and following positive IOD  
775 phases. These predictions may be helpful for practitioners deciding when to carry out beach  
776 cleanup operations.

777

778       There is reasonable agreement between our predictions, and the limited quantitative ob-  
779 servations of marine debris that are available from across Seychelles. Key discrepancies with  
780 observations have also highlighted the importance of shipping lanes as a source of marine  
781 debris for remote western Indian Ocean islands. Nevertheless, it is important to remember  
782 that our trajectory analysis relies on a large number of poorly constrained parameters. There  
783 is an urgent need for further studies on the rate of marine macrodebris fragmentation, bio-  
784 fouling, and sinking. Despite the number of marine debris modelling studies incorporating  
785 windage into simulations and acknowledging the important role it plays in determining drift  
786 trajectories, there are limited publicly available estimates of appropriate windage coefficients  
787 for common classes of marine debris. Additionally, this windage coefficient will likely change  
788 with time, as debris loses buoyancy and/or fragments. Finally, considerable uncertainty re-

789 mains in the input function of marine debris into the ocean. Nevertheless, a strength of this  
790 study is that our results can be easily recomputed for different combinations of sinking rate,  
791 beaching rate, and windage so, if improved constraints in the future demonstrate that our  
792 classification of debris (into our four classes A-D) is inappropriate, it will be straightforward  
793 to recompute results with the dataset and scripts provided in the Supplementary Data.

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811 special island and whose work inspired this research project.

## 812 **Data Availability Statement**

813 All data and scripts required to reproduce the figures in the main text are archived  
814 at the British Oceanographic Data Centre (link)<sup>2</sup>, with the exception of dFAD deploy-  
815 ment data due to a confidentiality agreement. Requests for access to dFAD deployment  
816 and tracking data should be addressed directly to the Ob7 pelagic ecosystem observatory  
817 (<https://www.ob7.ird.fr/>) using the following email address: adm-dblp@ird.fr.

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<sup>2</sup>Note: these data are currently undergoing archival, which will be complete by the time this paper is published. This link is static, and will direct to the full dataset once archival is complete.

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