



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

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### ABSTRACT

Vast quantities of debris are beaching at remote islands in the western Indian Ocean. We carry out marine dispersal simulations incorporating currents, waves, winds, beaching, and sinking, for both terrestrial and marine sources of debris, to predict where this debris comes from. Our results show that most terrestrial debris beaching at these remote western Indian Ocean islands drifts from Indonesia, India, and Sri Lanka. Debris associated with fisheries and shipping also poses a major risk. Debris accumulation at Seychelles is likely seasonal, peaking during February–April. This pattern is driven by monsoonal winds and may be amplified during positive Indian Ocean Dipole and El-Niño events. Our results underline the vulnerability of small island states to marine plastic pollution, and are a crucial step towards improved management of the issue. The trajectories used in this study are available for download, and our analyses can be rerun under different parameter choices.

### 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 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, 2020a). Elsewhere, however, particularly in the case of remote islands with minimal or no population, debris accumulating on the coast may have

been transported over great distances by ocean currents, winds, and waves (van Sebille et al., 2020). These islands, many of which belong to small island developing states, are faced with the deeply inequitable situation of bearing the costs of removing waste they were not responsible for generating, contrary to the “polluter pays” principle (OECD, 1975).

There are many small island developing states in the western Indian Ocean (Fig. 1) and, whilst marine plastic pollution is under-studied in this region relative to, for instance, the North Atlantic and Western Pacific, debris accumulation has been documented in many of these remote island groups. Seychelles is one such small island developing state, spread across over 100 islands north of Madagascar, from the isolated Aldabra Group in the southwest, to the Inner Islands on the Seychelles Plateau in the northeast. Marine debris monitoring programmes have found large quantities of debris accumulating across the latitudinal and longitudinal range spanned by Seychelles, such as at Aldabra Atoll (Burt et al., 2020), Alphonse Island (Duhec et al., 2015), Cousine Island (Dunlop et al., 2020), and many others (Macmillan et al.,

*Abbreviations:* ALDFG, Abandoned, Lost or otherwise Discarded Fishing Gear; dFAD, Drifting Fish Aggregating Device; IOD, Indian Ocean Dipole; ENSO, El Niño - Southern Oscillation.

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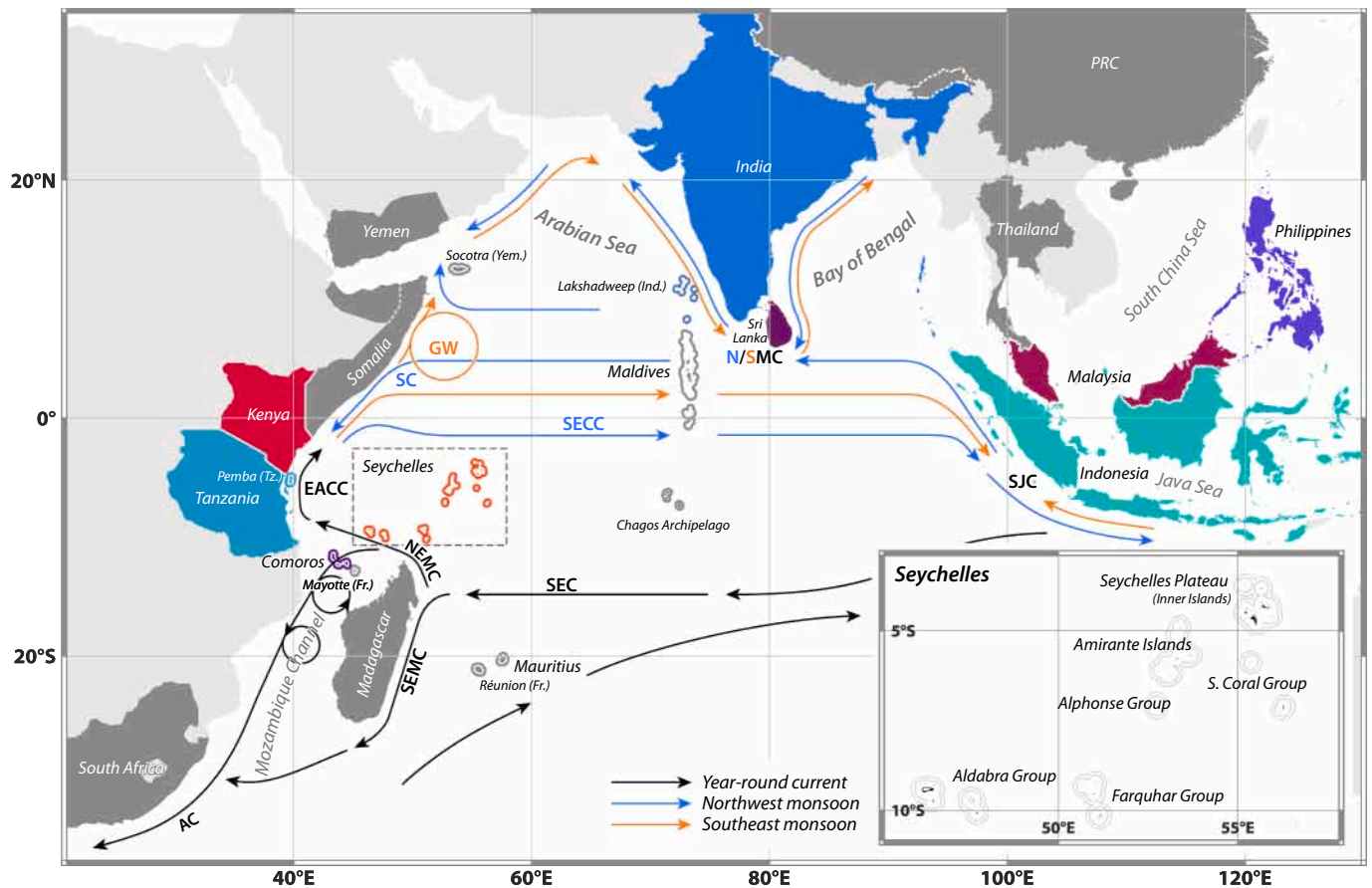
2022). Marine debris is primarily of terrestrial origin at some of these sites (e.g. Alphonse Island, [Duhec et al. \(2015\)](#)) whereas abandoned, lost, or otherwise discarded fishing gear (ALDFG) of marine origin dominates at others (e.g. Aldabra Atoll, [Burt et al. \(2020\)](#)).

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

Numerical models can also be used to predict the source of beaching debris by representing debris as Lagrangian particles or Eulerian tracers. These simulations can be run forward-in-time, i.e. assuming knowledge of some input distribution of marine debris and predicting where that debris is transported (e.g. [Kaandorp et al., 2020](#); [van der Mheen et al., 2020](#); [Chassignet et al., 2021](#)), or backward-in-time, i.e. simulating trajectories that lead to a site of interest and inferring debris sources based on where debris passed through in the past (e.g. [Duhec et al., 2015](#); [Stelfox et al., 2020](#)). In the context of marine debris attribution for remote islands, backward-in-time simulations are more efficient as they must only compute the small subset of trajectories that end at the site of interest, reducing computational cost. However, there are significant limitations associated with the backward-in-time approach. For instance, it is not possible to implement parameterisations for subgrid-

scale diffusivity (although see [Isobe et al. \(2009\)](#)). Even more significantly, since simulated backward trajectories comprise an unknown subset of all possible trajectories, there are fundamental limitations on the quantitative constraints that can be obtained on the sources of marine debris. Most studies using backward-in-time simulations are limited to qualitative predictions of debris sources based on assumptions of a fixed drift time (e.g. [Duhec et al., 2015](#)). [van Duinen et al. \(2022\)](#) used a Bayesian framework to quantify sources of debris for a beach in the Netherlands, but this approach still relies on assumptions on how long debris was adrift before beaching. For remote islands where potential sources of debris are many and distal, it is challenging to justify any a priori assumption for a drift time distribution. An innovative approach was used by [Stelfox et al. \(2020\)](#), who predicted the source fisheries for ghost gear accumulating in the Maldives based on backward-in-time simulations and constraints on drift time from biofouling. Unfortunately, these constraints are likely debris-type and site specific, and no such estimates exist in general for most remote islands.

In the absence of constraints on drift time, forward-in-time simulations are required to provide quantitative, physically justifiable estimates for sources of marine debris. To date, sources of debris have been quantified for the Seychelles as part of two regional ([van der Mheen et al., 2020](#)) and global ([Chassignet et al., 2021](#)) studies. However, neither study registered a significant number of particles arriving at Seychelles, and they were therefore unable to make robust conclusions about sources of marine debris for remote islands. Both were large-scale studies focusing on major marine debris transport pathways, but this nevertheless highlights an important data gap, as well as a particular



**Fig. 1.** Map of the Indian Ocean, with key countries, island groups, and basins highlighted. The remote islands discussed in this study are drawn with a halo for clarity. The colours are the same as those used in [Fig. 4](#). 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 northwest and southeast monsoons. *Inset:* The major island groups within Seychelles. (For interpretation of the references to colour in this figure, the reader is referred to the web version of this article.)

technical challenge for assessing sources of marine debris for small and remote locations.

As a result, whilst there are indications from bottle labels, no quantitative estimates exist for the relative importance of sources of debris for Seychelles, along with other remote island groups in the western Indian Ocean. Good constraints exist on source regions for one specific type of fishing gear, drifting Fish Aggregating Devices (dFADs), accumulating on Seychelles' beaches (Macmillan et al., 2022; Imzilen et al., 2021), but this has not been generalised to all marine-based sources of debris. The primary purpose of this study is to provide environmental practitioners working with remote western Indian Ocean islands with a first quantitative estimate for debris sources, and to make available a product that can be adapted to future improved constraints on marine debris physics. We use large-scale Lagrangian forward simulations, forced by ocean currents, waves, and winds, generalisable to arbitrary sinking and beaching rates, to answer the following questions:

- Which countries are the most likely terrestrial sources of debris accumulating at Seychelles (and other western Indian Ocean islands), and how sensitive are these estimates to debris properties such as sinking rate and windage?
- If debris is generated at sea (from fisheries, ships, etc.), from which regions is there most risk of debris beaching at Seychelles, and can we therefore predict high-risk fisheries and shipping channels?
- What are the physical drivers of debris accumulation at Seychelles, and are variations in accumulation rates (seasonal and interannual) predictable, allowing for more targeted cleanup efforts?

All datasets generated by this study, as well as associated documentation, results that are not discussed in detail in the main text, and code to repeat analyses under different parameters, can be found in the Supplementary Materials.

## 2. Methods

### 2.1. Particle tracking

To simulate the transport of marine debris, we carry out Lagrangian particle tracking using OceanParcels (Lange and Sebille, 2017; Delandmeter and van Sebille, 2019). Particles are tracked for 10 years or until the end of 2019, with trajectories integrated using a fourth-order Runge-Kutta scheme and a time-step of 1 h. Over large scales, buoyant marine debris is transported by surface currents, Stokes drift, and in the case of debris protruding above the sea surface, windage (van Sebille et al., 2020). All three processes are important in describing its dispersal (e.g. Duhec et al., 2015; Maximenko et al., 2018). We assume the force experienced by particles from the wind is parallel and proportional to surface winds, but note that this is a simplification compared to the real forces experienced by buoyant debris (Domon et al., 2012). We advect particles of terrestrial origin with 5 forcing scenarios: just surface currents (C0), surface currents + Stokes drift (CS0), and surface currents + Stokes drift +1–3 % windage (CS1–3). Particles of marine origin are advected using the same sets of forcing, plus 4 % and 5 % windage (CS4–5). We use the 1/12° CMEMS Global Ocean Physics Analysis GLORYS12V1 (Lellouche et al., 2021) for daily surface currents, 1/5° Global Wave Reanalysis WAVEYRS (Law-Chune, 2021) for three-hourly Stokes drift, and 1/4° three-hourly surface winds from ERA5 (Hersbach et al., 2020) (all 1993–2019). Surface current speed and surface wind vectors for the northwest and southeast monsoons are shown in Fig. 2.<sup>1</sup> We apply a homogeneous lateral diffusivity of 10 m<sup>2</sup> s<sup>-1</sup> to particles to represent the diffusive effects of sub-grid scale surface currents, based on

<sup>1</sup> In this study, we refer to the monsoons around December to February, and June to August, as the northwest and southeast monsoons, respectively, in line with terminology used in Seychelles

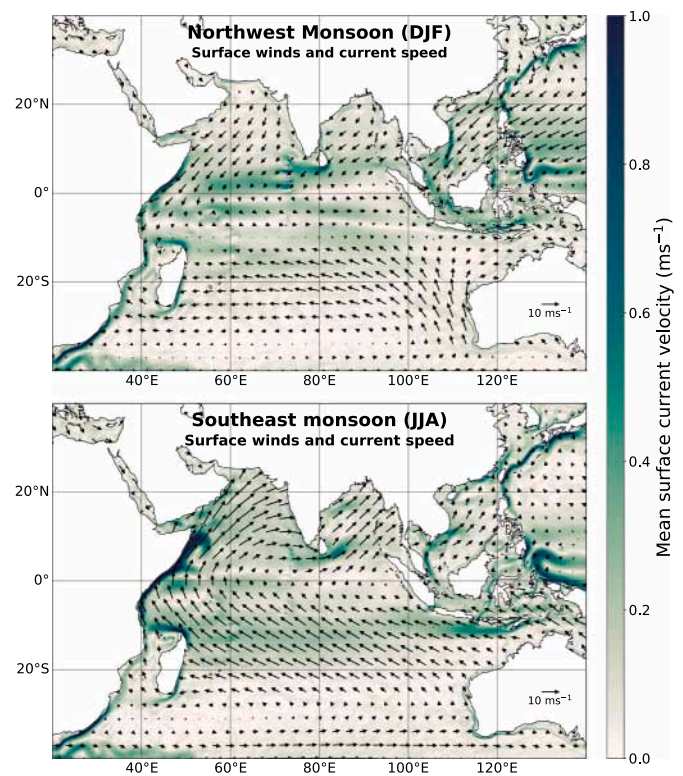


Fig. 2. Mean surface current speed (colours, Lellouche et al. (2021)) and surface winds (arrows, Hersbach et al. (2020)) as used in our analyses for the northwest monsoon (top) and southeast monsoon (bottom).

a typical value of the horizontal Smagorinsky diffusivity in the equatorial Indian Ocean diagnosed from GLORYS12V1 (Smagorinsky (1963), Supplementary Fig. 1) and in line with previous studies (Okubo, 1971; Kaandorp et al., 2020). Further technical details on the treatment of particle tracking near the coasts are described in Supplementary Text 1.

### 2.2. Particle sinking and beaching

Marine debris is lost from the ocean surface through processes including beaching and sinking. These processes are complex and driven by small-scale physical and biological processes (Critchell and Lambrechts, 2016; van Sebille et al., 2020) and must therefore be parameterised in large-scale numerical models. Many models parameterise sinking as decay in the mass of debris represented by a particle (e.g. Kaandorp et al., 2020; Chassignet et al., 2021). Although more sophisticated implementations exist (e.g. Turrell, 2020b), beaching is often parameterised by explicitly removing particles based on criteria, such as particles entering a land cell due to Stokes drift, wind and/or numerical error (e.g. Zhang et al., 2020; Cardoso and Caldeira, 2021), particle stagnation (e.g. Seo and Park, 2020; Bosi et al., 2021), or as a stochastic process associated with some probability (e.g. van der Mheen et al., 2020; Onink et al., 2021).

An advantage with modelling beaching as a stochastic process is the ability to incorporate complex behaviour such as resuspension (Liubartseva et al., 2018; Onink et al., 2021) and, as understanding of the physics of beaching improves, stochastic parameterisations will become an increasingly valuable tool. However, as these parameterisations remove Lagrangian particles from circulation (even if only temporarily), this can significantly reduce the number of particles representing floating debris in the model. This is a problem when attempting to quantify the sources of debris for small and remote islands: these islands are very small ‘targets’ and beaching events may be missed due to an insufficient number of particles, as was the case for Seychelles in the

studies of van der Mheen et al. (2020) and Chassignet et al. (2021).

Instead, we assume that there is (i) a constant rate of debris removal through sinking,  $\mu_s$ , and (ii) a constant rate of debris removal through beaching,  $\mu_b^*$  when a particle is within a  $1/12^\circ$  coastal grid cell, and implement sinking and beaching offline through postprocessing of the trajectory data. These two parameters are highly uncertain, particularly  $\mu_s$ . Fazez and Ryan (2016) estimated sinking timescales for small polyethylene (LDPE and HDPE) fragments and, whilst the statistical model used in their study is not identical to our sinking parameterisation, they estimated sinking timescales on the order of 17–66 days. Kaandorp et al. (2020) predicted a slightly higher sinking timescale of  $1/\mu_s = 81$  days based on an inverse model incorporating observations of floating debris in the Mediterranean Sea. Koelmans et al. (2017) predicted an effective removal timescale of marine debris from the ocean surface (through fragmentation into microplastics) on the order of months, based on mass-balance arguments and observations of floating debris. However, Lebreton et al. (2019) argued that observations of the age distribution of debris in the North Pacific subtropical gyre are inconsistent with rapid sinking rates, instead suggesting that observations are more consistent with low sinking rates and rapid scavenging of debris at coastlines through beaching. Since these parameters are so uncertain, it is very useful to be able to modify these parameters through postprocessing, rather than having to rerun expensive particle tracking analyses.

We store ‘beaching events’, defined as the mass of debris beaching from a particle whilst that particle is continuously within coastal grid cells of a beaching site. In this study, we store results for 27 beaching sites, representing islands or island groups. 18 of these sites are within Seychelles (Aldabra, Assomption, Cosmoledo, Astove, Providence, Farquhar, Alphonse, Poivre, St Joseph, Desroches, Platte, Coëtivy, Mahé, Fregate, Silhouette, Praslin, Denis, and Bird). For the terrestrial-sourced debris experiments only, we included an additional 9 sites from the wider western Indian Ocean (Comoros, Mayotte [France], Lakshadweep [India], Maldives, Mauritius, Réunion [France], Pemba [Tanzania], Socotra [Yemen], and the Chagos Archipelago). A grid file with the labelled grid cells for all sites can be found in Supplementary Dataset 3. For brevity, we focus on Seychelles in this paper, specifically islands on the Seychelles Plateau, and the Aldabra Group as representative of the Outer Islands. We focus on these two island groups as the population of Seychelles almost entirely resides within the Seychelles Plateau, whilst Aldabra is a World Heritage Site and the site of a major observational marine debris study (Burt et al., 2020). Together, these groups span the geographical range of Seychelles. Analyses and figures for other island groups that could not be included in this paper can be produced using the Supplementary Datasets (Vogt-Vincent, 2022; Vogt-Vincent and Johnson, 2022a,b).

By efficiently choosing which data to store during particle tracking simulations (see Supplementary Text 2), it is possible to compress all data required to reconstruct almost all beaching events from the over  $2 \times 10^{11}$  particles used across all our simulations in  $\sim 1.2$  TB, whilst allowing key parameters to be varied through postprocessing.

### 2.3. Debris sources

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

#### 2.3.1. Terrestrial sources

Debris can enter the ocean through rivers (transported from inland), as well as through direct coastal input from coastal populations through stormwater, sewage, or poor waste disposal (Mihai et al., 2022). For riverine debris input, we use the modelled midpoint annual estimates from Meijer et al. (2021), gridding the emissions from each river mouth

to the nearest coastal cell on the  $1/12^\circ$  GLORYS12V1 grid (Section 2.1). For direct coastal input, we base our estimates on modelled annual mismanaged plastic waste generation estimates from Lebreton and Andrady (2019). We degrade the resolution of this product to the GLORYS12V1 resolution, and then calculate emissions to the ocean by assuming that a fraction  $f_i = f_c \cdot \exp\left[-\left(\frac{d_i}{L}\right)^2\right]$  of the mismanaged waste produced in a grid cell  $i$  enters the nearest coastal cell, where  $f_c$  is the maximum likelihood of mismanaged waste entering the ocean,  $d_i$  is the distance of cell  $i$  from the coast, and  $L$  is a length scale over which direct coastal input to the ocean is significant. This parameterisation is based on the assumption that waste is less likely to enter the ocean the further from the coast it is generated. Many previous studies have used the alternative assumption, inherited from Jambeck et al. (2015), that a fixed fraction of mismanaged waste generated within 50 km of the coast enters the ocean. Both of these parameterisations are somewhat arbitrary, but we believe that our assumptions are more appropriate.

We set  $L = 15$  km as a conservative value to reflect the length scale of a typical coastal city, and to reduce the risk of ‘double counting’ debris entering rivers from coastal cities (counted as riverine debris in Meijer et al. (2021)). The parameter  $f_c$  is the main control on the ratio  $r$  of marine debris generation from coastal versus riverine sources. In the absence of good constraints on this parameter, we take  $f_c = 0.25$ , corresponding to a total flux of debris from coastal and riverine sources of  $3.1 \text{ Mt. y}^{-1}$  and  $1.0 \text{ Mt. y}^{-1}$  respectively ( $r = 3.1$ , between  $r = 1.9$  in Kaandorp et al. (2020) and  $r = 4.9$  in Lebreton et al. (2018)). The parameter  $f_c$  can, however, be modified during postprocessing and, if it becomes better constrained in the future, it is straightforward to regenerate our results for another value of  $f_c$ , or even an entirely different debris input distribution, using the Supplementary Datasets.

To minimise the cost of simulations, we only consider coastal cells for countries that could reasonably act as a source of marine debris for islands in the western Indian Ocean, identified from a preliminary backward particle tracking experiment (Supplementary Text 3, Supplementary Fig. 2). Many coastal cells are associated with a very small flux of debris, so we remove the 7773 (of 20,742) coastal cells with the smallest contributions, leaving 99.99 % of riverine plastic, and 99.9 % of coastal plastic. An overview of the terrestrial sources of debris used in our experiments is shown in Fig. 3.

#### 2.3.2. Marine sources

Ship tracking data from the automatic identification system (AIS), broadcast by large ships, has been used as a proxy for fishing effort and therefore ALDFG production in previous studies (e.g. Kaandorp et al., 2020). However, AIS coverage is poor in the Indian Ocean (Richardson, 2022). Instead, we use publicly available Indian Ocean Tuna Commission (IOTC) effort data for purse-seines and longlines (provided on  $1^\circ$  and  $5^\circ$  grids respectively, Fig. 3), which is well-studied and has been used extensively as an indicator of fishing activity, particularly in the case of purse-seines (Kaplan et al., 2014; Imzilen et al., 2022). Effort is quantified as fishing time (purse-seines) and hooks (longlines). We consider longline fisheries from Japan, Taiwan, and Korea only, as data from these countries is the most reliable in terms of spatial distribution (Kaplan et al., 2014, Emmanuel Chassot (*personal communication*)). Since purse-seine and longline vessels lose gear at different rates (Kuczenski et al., 2022) with potentially different behaviour in the water, we do not aggregate effort from these two fisheries, and instead consider them separately.

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

## 2.4. Seeding strategy

### 2.4.1. Terrestrial sources

For each coastal cell  $i$  on the GLORYS12V1 grid, we split the annual flux of debris of terrestrial origin  $T_i$  (as described in Section 2.3.1) across 4 equally spaced releases per month, for a total of 48 identical releases per year. Releasing a uniform number of particles per coastal cell would be computationally inefficient as the mass represented by each particle would vary over 5 orders of magnitude. However, scaling particle number linearly with debris flux would result in extremely low numbers of particles representing weak debris sources, which may still be locally important. As a compromise, we divide the debris associated with each release across  $n_i$  particles, such that the initial mass associated with a particular particle  $j$  released at cell  $i$  is  $M_j^0 = T_i/48n_i$ . We set  $n_i = \lceil [c_1 \cdot \log_{10}[F_i] - c_2]^2 \rceil$ , where  $c_1 = 16$  and  $c_2 = 18.4$  are arbitrary parameters chosen to distribute particles reasonably whilst keeping computation tractable, and  $\lceil \dots \rceil$  is the ceiling function. We release 13.7 million terrestrial particles per release event, for a total of 656 million per model year.

### 2.4.2. Marine sources

In each marine cell on the GLORYS21V1 grid between 20°E-130°E and 40°S-30°N (excluding the Mediterranean) we generate 36 particles per release. We release particles at four equally spaced intervals per month, with 26.5 million particles per release event and a total of 1.27 billion particles per model year.

## 2.5. Quantifying beaching debris

### 2.5.1. Terrestrial sources

Since each particle representing terrestrial debris has an explicit initial mass associated with it (Section 2.4.1), there is also a mass associated with each beaching event. We therefore compute a list of beaching events  $(i, j, m_b, t_b)$  for each set of parameters, where  $i$  is the beaching site,  $j$  is the country of origin,  $m_b$  is the mass beached (kg), and  $t_b$  is the beaching month. We then bin  $m_b$  as a terrestrial-sourced debris flux matrix  $F_T^j(t_b)$  for each set of parameters, giving the total mass beaching at  $i$  from  $j$  in month  $t_b$ .

### 2.5.2. Marine sources

We firstly compute a list of beaching events  $(i, x, y, m_b, t_s, t_b)$  for each set of parameters, where  $i$  is the beaching site,  $(x, y)$  is the input location,

$m_b$  is the ‘mass’ beached,  $t_s$  is the input month, and  $t_b$  is the beaching month. Importantly, we compute  $m_b$  as a *fraction* of debris generated at  $(x, y, t_s)$ . From this list, we then define a marine-sourced debris flux matrix  $F_M^i(x, y, t_s, t_b)$ , giving the mass (fraction) of debris beaching at  $i$  and month  $t_b$ , given that it entered the ocean at  $(x, y)$  and time  $t_s$ . We further compute the marine-sourced debris source flux  $F_{Ms}^i(x, y, t_s) = \sum^{t_b} F_M^i$  and marine-sourced debris beaching flux  $F_{Mb}^i(x, y, t_b) = \sum^{t_s} F_M^i$ . Note that  $F_{Ms}^i(x, y, t_s)$  is equivalent to the likelihood of debris entering the ocean at  $(x, y, t_s)$  ever beaching at  $i$ , and we call this quantity the *risk*. We also compute monthly climatological versions of all the matrices (subscript  $c$ ) by summing over all years, e.g.  $F_{Mc}^i(x, y, t_s^*, t_b^*) = 1 \dots 12$ .

Assuming marine debris input by fishery  $j$  is proportional to fishing effort  $E_j(x, y, t_s^*)$ , the relative flux  $f_j^i(x, y)$  of fishery  $j$  debris that ever beaches at site  $i$  is given by  $f_j^i(x, y) = \sum_{t_s=1}^{12} (E_j(x, y, t_s^*) \cdot F_{Msc}^i(x, y, t_s^*))$ . We can normalise this relative flux by the total flux from all sources, to give the *realised risk*  $R_j^i(x, y)$  to site  $j$  from fishery  $i$  at location  $(x, y)$ ,  $R_j^i(x, y) = \frac{f_j^i(x, y)}{\sum^{x, y} f_j^i(x, y)}$ , which is the likelihood that fishery  $j$  debris beaching at site  $i$  entered the ocean at  $(x, y)$ . Along similar lines, we also compute a monthly climatology  $B_j^i(t_b^*)$  for beaching rates from fishery  $j$  at site  $i$ ,  $B_j^i(t_b^*) = \sum^{x, y} \left( \sum_{t_s=1}^{12} (E_j(x, y, t_s^*) \cdot F_{Mc}^i(x, y, t_s^*, t_b^*)) \right)$ .

### 2.5.3. Seasonality

To test whether seasonality exists in a time-series  $f(t)$ , we define the operator  $\mathcal{F}[f(t)]$ , which returns a sine wave with the phase and wavelength of the annual component of  $f(t)$  (based on its Fourier spectrum). We determine that there is a significant seasonal component to  $f(t)$  if  $\text{corr}[\mathcal{F}[f(t)], f(t)]$  is significant ( $p < 0.01$ , taking into account autocorrelation within both time-series (Bretherton et al., 1999)).

## 2.6. Debris classes

In our model, the behaviour of marine debris in the ocean is set by the three parameters  $\mu_s$  (beaching rate), and the forcing scenario (C0 or CS0–5). No one set of parameters will describe all marine debris, and constraints on all three are poor. We simplistically explore the sensitivity of model results to this parameter-space in Section 3.3.1, but to provide concrete examples, we define four representative debris classes in Table 1.

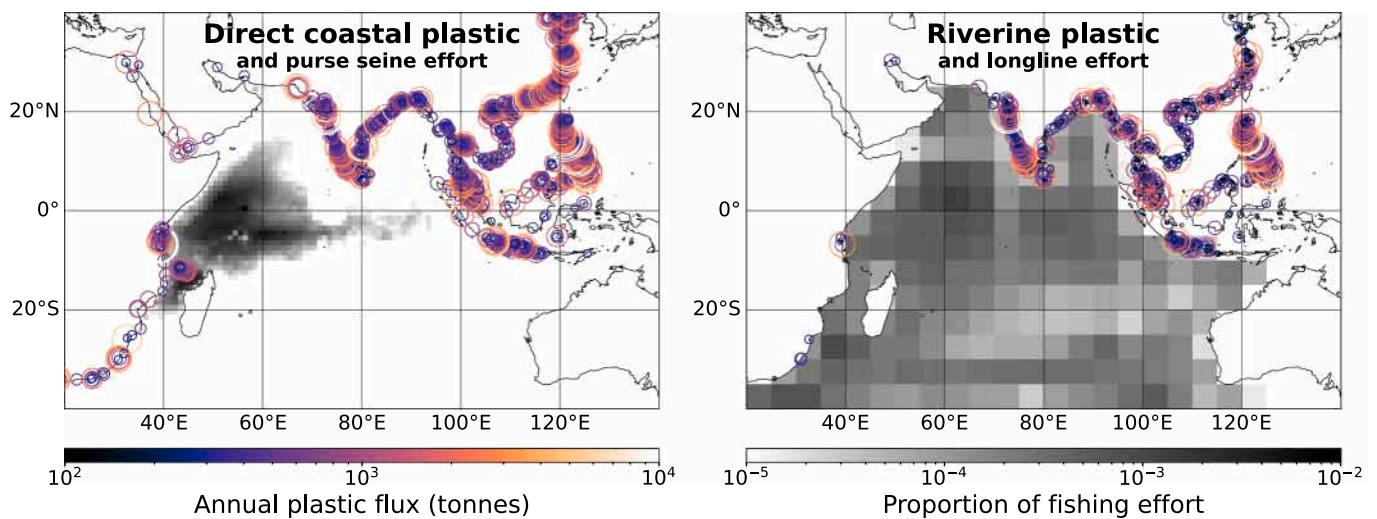


Fig. 3. Left: Terrestrial sources of debris from direct coastal input (as described in Section 2.3.1, coloured and scaled by the annual plastic flux per grid cell), and proportion of purse seine fishing effort in the Indian Ocean (fishing hours, as described in Section 2.3.2). Right: Terrestrial sources of debris from riverine input, and proportion of longline fishing effort in the Indian Ocean (hooks).

To derive these classifications, we use guidance on windage coefficients from Duhec et al. (2015) and Domon et al. (2012), sinking rates from Fazey and Ryan (2016), and beaching rates from our own analysis (Supplementary Text 4, Supplementary Figs. 3–4) and Kaandorp et al. (2020). However, we stress that windage coefficients and sinking rates for different types of marine debris remain poorly constrained, and the classes we have defined are suggestions only. Practitioners can recompute predictions for parameters of their own interest using the Supplementary Datasets.

## 2.7. Comparison with observations

Burt et al. (2020) estimated the total mass of debris that accumulated on Aldabra Atoll (Seychelles), as well as countries of origin for a small sample of PET bottles. Quantitative source analyses have also been carried out for Alphonse, Coëtivy, Astove and Platte (Duhec et al., 2015; The Ocean Project Seychelles, 2019). Finally, Macmillan et al. (2022) analysed patterns of (satellite-tracked) drifting Fish Aggregating Device (dFAD) beaching events across Seychelles. We carry out a quantitative, side-by-side comparison of our analyses against the findings of these studies in Section 3.3.

## 3. Results and discussion

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

#### 3.1.1. Debris of terrestrial origin

There is significant variation in the predicted source countries for debris beaching at the 27 sites investigated in this study (Fig. 4). These figures can be interpreted as the predicted likelihood of a fragment of marine debris originating from the source country and beaching at the target island (group), given that it has properties reflecting Class A, B, or C debris as defined in Table 1.

For Class A debris (Fig. 4(a)), East Africa (predominantly Tanzania) is expected to be the largest source of marine debris for most of the Outer Islands of Seychelles, although Comoros is the dominant source for Aldabra and Assomption. For the Inner Islands on the Seychelles Plateau, most Class A debris is expected to come from within Seychelles, with the remainder sourced from East Africa. For sites in the central-northern Indian Ocean (Maldives and Lakshadweep), India and/or Sri Lanka are expected to be the principal sources of debris. Only the Chagos Archipelago is predicted to source most of its Class A debris from Indonesia.

For Class B debris (Fig. 4(b)), a combination of longer residence time at the ocean surface (3 months) and easterly winds allows Indonesia to begin to dominate the marine debris budget for much of the western Indian Ocean. Our analyses predict that Indonesia is responsible for over 50 % of all Class B debris for all Outer Islands of Seychelles (and remains the dominant source for the Chagos Archipelago). Seychelles and Tanzania are still expected to be significant sources of debris within the inner islands of Seychelles (particularly Mahé, the main population centre of Seychelles), but substantial proportions are also predicted to originate from Indonesia and, in the case of islands in the northernmost Seychelles Plateau (Denis and Bird islands), India and Sri Lanka. India and Sri Lanka are expected to still act as the main sources of debris for the relatively nearby island groups of Lakshadweep and Maldives, but

the lower sinking rate and contributions from winds and waves during the northwest monsoon also results in these countries becoming significant sources of Class B debris for Socotra, dominated by local sources from Yemen for Class A debris.

Finally, Class C debris (Fig. 4(c)) beaching across Seychelles (and the Chagos Archipelago) is expected to originate almost entirely from the northern and eastern Indian Ocean. Indonesia is still expected to be the largest single source country, but a significant proportion is swept from the Philippines and, in the case of more northerly islands, India and Sri Lanka. Seychelles and East Africa are not significant sources of Class C debris for any sites in Seychelles. Our analyses also suggest that Mauritius and Réunion, dominated by local sources for less-buoyant classes of debris, receive significant quantities of Class C debris from South Africa (57 % and 36 % respectively).

In general, the distribution of terrestrial sources is sensitive to the sinking rate, but relatively insensitive to the beaching rate (see Section 3.3.1). The distribution of terrestrial sources is sensitive to both Stokes drift and windage, although the presence or absence of Stokes drift appears to be particularly important in partitioning sources between the eastern and western Indian Ocean for the outer islands of Seychelles (Supplementary Figs. 5–7).

We can also extract the predicted drift time distribution for debris accumulating at our study sites (shown for Aldabra in Supplementary Fig. 8). Unsurprisingly, the more buoyant debris classes have a broader range of drifting times, where drifting times are stratified by the oceanographic distance of source countries from Aldabra. For instance, for Class C debris accumulating at Aldabra, debris arriving from Comoros and Tanzania have generally only been at sea for 1–2 months, whereas debris arriving from Indonesia has been at sea for at least 6 months, with a small proportion exceeding 2 years. However, the distribution of drift times is complex and multimodal. Although Lagrangian backtracking is considerably less computationally expensive than the approach used in this study, van Duinen et al. (2022) were required to make an a priori assumption for the drift time distribution of debris accumulating at their site of interest. These drift time distributions for Aldabra highlight that assuming a uniform age distribution of beaching debris is not an appropriate assumption for remote islands.

As further discussed in Section 3.2, there is significant temporal variability in accumulation rates at many of these remote sites, particularly for Class A debris. However, recomputing Fig. 4 for subsets of the full time-series suggests that our source attribution is robust for almost all sites (Supplementary Text 5).

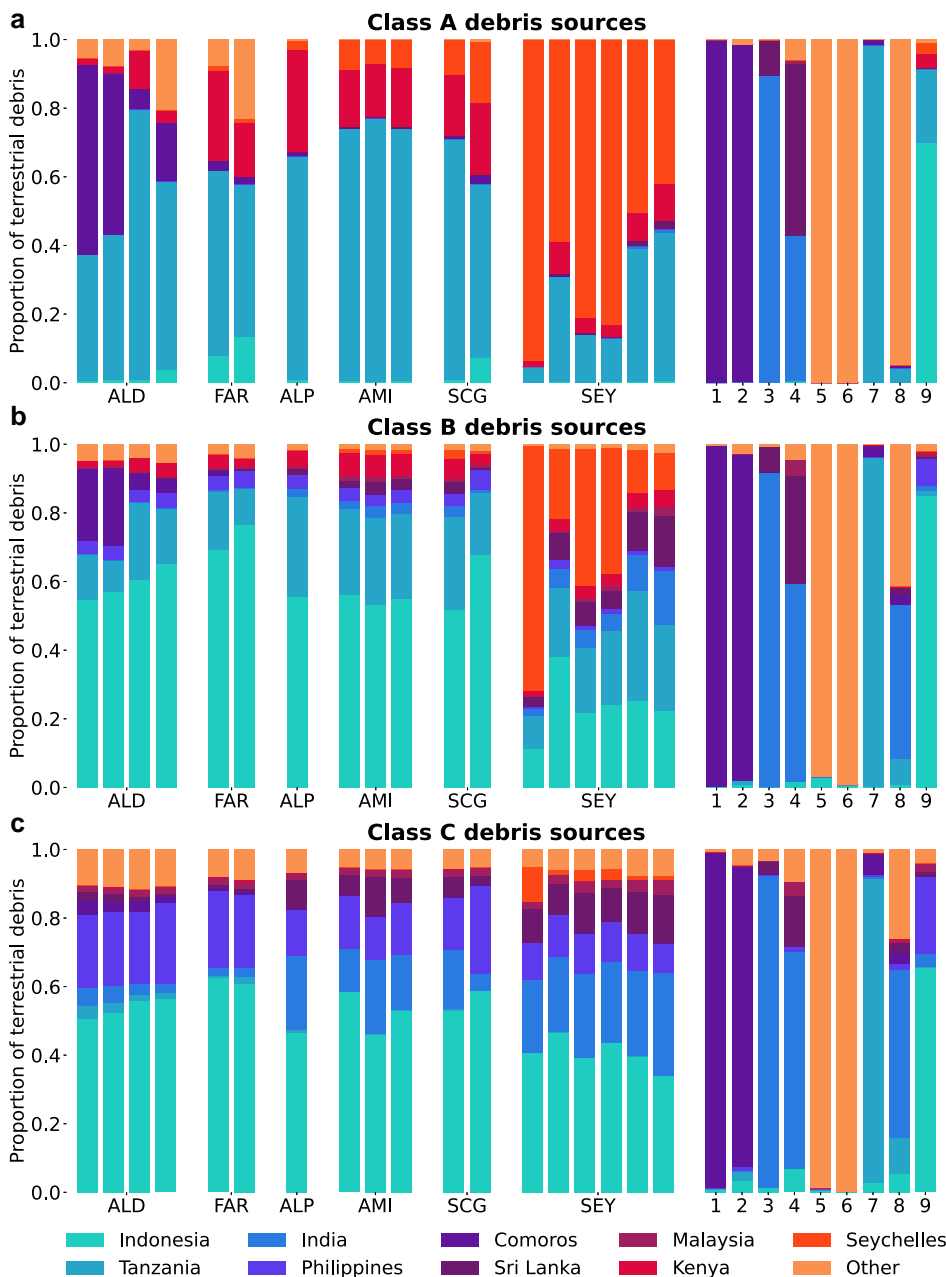
#### 3.1.2. Debris of marine origin

As with the terrestrial case, the probability of debris lost or discarded at sea eventually beaching at Seychelles strongly depends on the physical properties of the debris, and where it entered the ocean. Fig. 5(a) and (d) show the time-mean risk  $\bar{F}_{Ms}(x,y)$  of marine debris to the Aldabra Group, i.e. the mean likelihood that debris entering the ocean at  $(x,y)$  ever beaches at the Aldabra Group. Incoming Class A debris beaching at Aldabra (Fig. 5(a)) is sourced from a relatively narrow latitudinal band, due to primarily zonal currents around Aldabra. The Class A risk region for the Aldabra Group is almost entirely eastward of the island group, as these islands are in the path of a powerful westward-flowing ocean current (the North Madagascar Current). In contrast, the Class A risk map for the Seychelles Plateau (Supplementary Fig. 11) is centred on the

**Table 1**

Representative classes of debris used in this study, defined by beaching rate  $1/\mu_b^*$  (days), sinking rate  $1/\mu_s$  (days), physical scenario, buoyancy, exposure to the winds, and possible real-life examples of the representative class. All debris classes have positive buoyancy.

	$1/\mu_b$ (d)	$1/\mu_s$ (d)	Scenario	Buoyancy	Exposure	Examples
Class A	30	30	CS0	Low	Negligible	mm-scale plastic fragments, nurdles
Class B	30	90	CS1	Med	Low	Bottle caps, small domestic items
Class C	30	360	CS3	High	Med	Beach sandals, bottles, foam sheets, buoyant nets
Class D	30	1800	CS5	Very high	High	ALDFG with buoys, robust empty bottles



**Fig. 4.** 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, Assomption, 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).

plateau due to the monsoonal reversal of prevailing zonal currents around the island group (Schott et al., 2009).

With a significantly longer residence time at the ocean surface, and greater propulsion due to windage, the risk maps for Class D debris (Fig. 5(d)) cover a much greater area than for Class A debris. For both the Aldabra Group and Seychelles Plateau, debris from most of the tropical Indian Ocean has a non-negligible chance of beaching at one of these island groups. Although much debris in the Indonesian archipelagic seas and further afield is removed through beaching within the narrow straits of the Indonesian Throughflow, the sheer quantity of mismanaged waste generated in Indonesia and the Philippines allows a significant quantity to leak into the Indian Ocean.

The right-hand side panels in Fig. 5 show predictions of the realised risk to the Aldabra Group of ALDFG associated with purse-seine and longline fisheries. In the case of purse-seine debris, due to the concentration of purse-seine fishing effort around the Seychelles, our analyses suggest that most debris originates from the western Indian Ocean. In

contrast to purse-seine fisheries, effort associated with longline fisheries is more broadly distributed around the Indian Ocean. As a result, the footprint of the potential source region is much larger than for purse-seines. In the case of longline ALDFG behaving as Class D debris, debris could reasonably be sourced from as far afield as the southeastern Indian Ocean, west of Australia (Fig. 5(f)), largely due to the broader areal footprint of longline fisheries in the Indian Ocean. This suggests that a significant proportion of ALDFG beaching at Seychelles could originate from outside the Seychelles EEZ, particularly in the case of longline debris.

Finally, Fig. 5 also shows that there is significant overlap between major shipping lanes (shading in Fig. 5), and high risk regions for Seychelles. Even in the case of short-lived Class A debris, the major shipping lanes linking the Bay of Bengal and South China Sea to the Atlantic pass within the high risk zone for the Aldabra Group. For Class D debris, most of the major shipping lanes in the Indian Ocean pass through regions associated with a high risk of beaching for both the

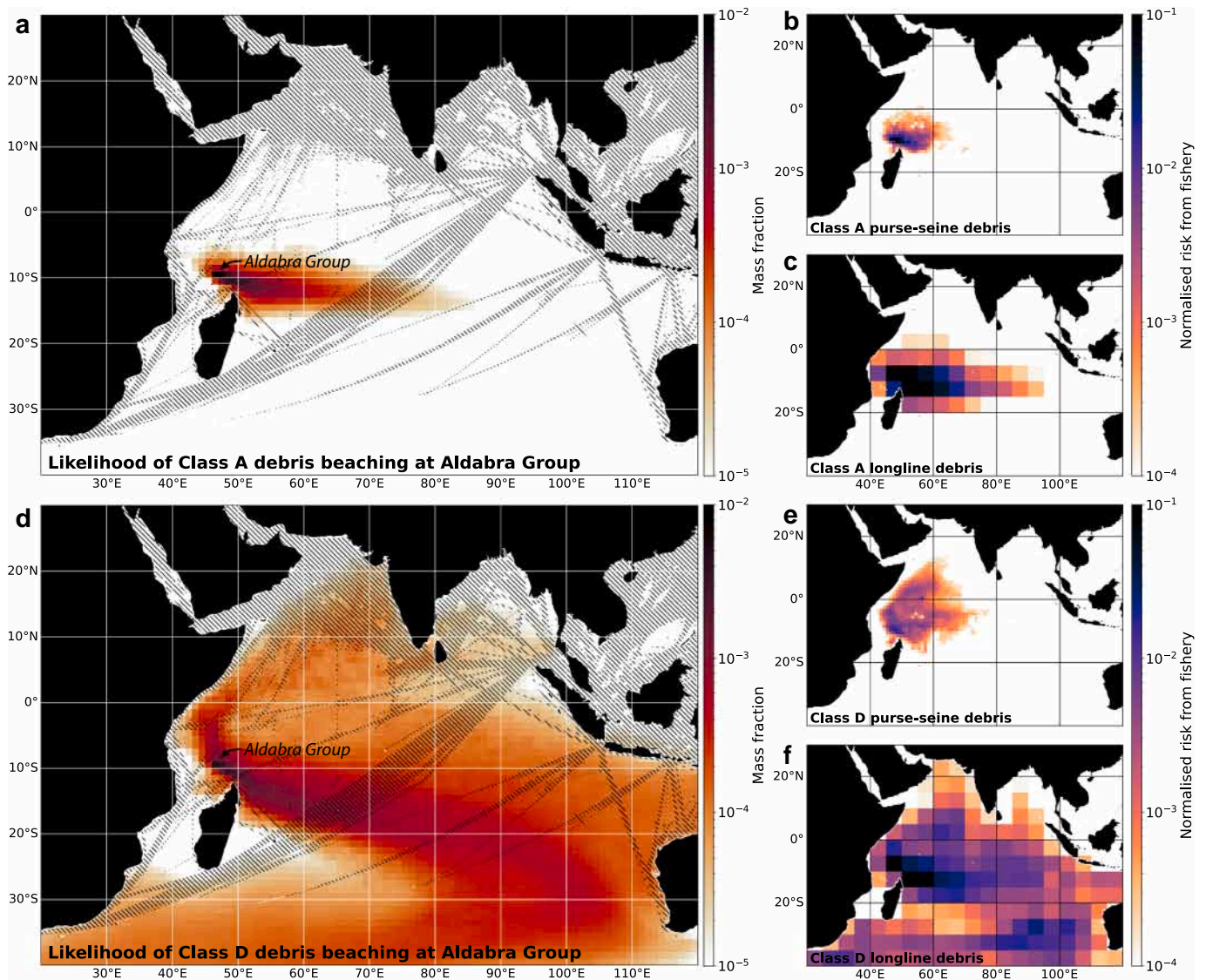
Inner and Outer Islands of Seychelles, including Atlantic-bound connections from the Middle East and Java Sea, as well as those originating from the Bay of Bengal and South China Sea.

### 3.2. Variability and drivers of beaching debris

Despite the monthly input of terrestrial debris remaining constant in our analyses, there is substantial temporal variability in beaching rate predicted for remote islands. Fig. 6(a) shows the mass of Class A debris beaching at the Aldabra Group and Seychelles Plateau per month from 1995 (two years after the first debris release) to 2014 (the last release year for terrestrial debris). Although the average accumulation rate for Class A debris at the Seychelles Plateau is substantially higher than for the Aldabra Group, the monthly accumulation rate at the Aldabra Group varies over 6 orders of magnitude. These patterns are a result of the different principal sources of Class A debris for each island group (Fig. 4 (a)). In the case of the Seychelles Plateau, most Class A debris is sourced from within Seychelles, largely from Mahé (<100km from most islands). The transport pathways from source to sink for Class A debris beaching within the Seychelles Plateau are therefore short (<2 weeks), with less

of an opportunity for seasonal variations in ocean currents or eddy variability to disrupt this pathway. In contrast, most Class A debris beaching at the Aldabra Group originates in Comoros and Tanzania, both of which are hundreds of kilometres away and are connected through low probability connections (Fig. 5(a)). As a result, the Aldabra Group sees almost no Class A debris beaching in most months, but if an eddy happens to direct a filament of Class A debris towards Aldabra, a large amount of debris may beach in a short period of time. This prediction is similar to patterns of ‘pulsed recruitment’ predicted for the long-distance larval dispersal of some marine organisms (e.g. Siegel et al., 2008).

In contrast, the Aldabra Group and Seychelles Plateau see a similar level of variability in beaching rates for Class C debris (Fig. 6(b)). Most Class C debris beaching at both island groups originates from distal sources in southeast Asia and, in the case of the Seychelles Plateau, south Asia. Class C debris arrives at both groups through long-distance transport pathways, and there is therefore ample opportunity for these transport pathways to be controlled by stochastic, eddy-induced variability. The variability in accumulation rate at the Aldabra Group is lower for Class C debris than for Class A, possibly because the wider



**Fig. 5.** Time-mean risk  $\bar{F}_{Ms}(x, y)$  for the Aldabra Group for (a) Class A debris and (d) Class D debris. Hatching shows shipping corridors with the most intense traffic from Jan 2015 to Feb 2021 (Cerdeiro et al., 2020). Realised risk  $R_i^j(x, y)$  for (b–c) Class A debris and (e–f) Class D debris from (b,e) purse-seines and (c,f) longlines. Corresponding plots for the Seychelles Plateau and other debris classes can be found in Supplementary Figs. 9–14. Note the logarithmic scales in all panels.



geographic distribution of sources and greater time available for mixing ‘smooths out’ the distribution of marine debris in the ocean. Nevertheless, monthly beaching rates for both groups are predicted to vary across around three orders of magnitude, with most debris arriving during short periods of high accumulation rate.

3.2.1. Seasonal variability from terrestrial sources

Given this enormous variability in beaching rate, it is useful to understand whether beaching rate varies entirely stochastically, or whether there is some predictability (which could help with the organising of beach clean-ups and other management activities). In

particular, prevailing winds and many currents in the Indian Ocean change direction following the monsoons, which have previously been suggested to control the partitioning of debris between the southern and northern Indian Ocean (van der Mheen et al., 2020). Fig. 6(c)–(d) show the monthly accumulation rate for Class A and C debris arriving at the Aldabra Group and Seychelles Plateau, averaged over 1995–2014. For Class A debris (Fig. 6(c)), almost all debris beaches at the Aldabra Group between January and March, i.e. the end of the northwest monsoon. Comoros is a major source of Class A debris for the Aldabra Group but ordinarily, debris from Comoros is swept into the Mozambique Channel and away from Aldabra. Rapid debris transport from Comoros to

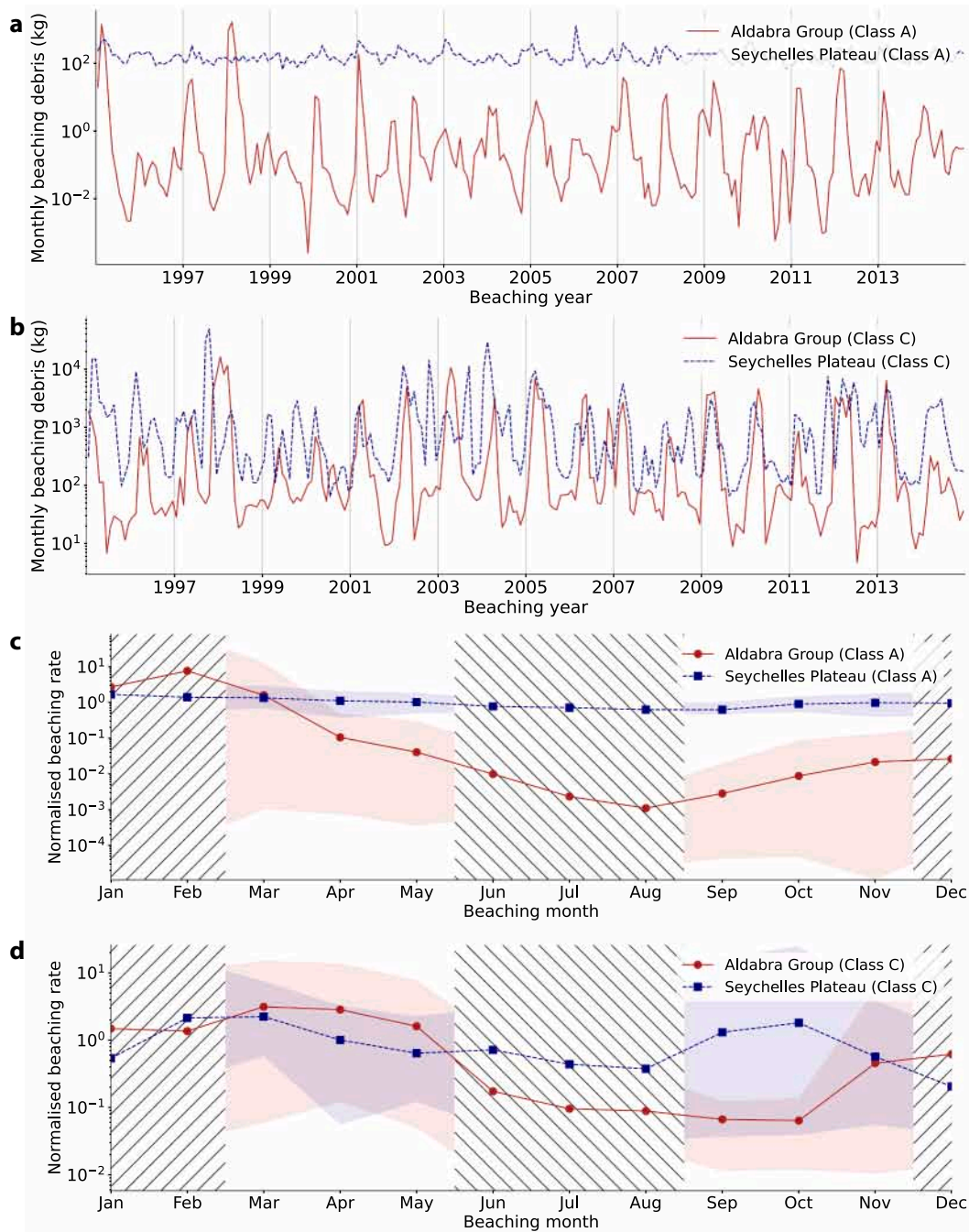


Fig. 6. (a)–(b) Monthly beaching rate  $F_T^d(t_b)$  from 1995–2014 at the Aldabra Group and Seychelles Plateau assuming all terrestrial debris is (a) Class A and (b) Class C. (c)–(d). Monthly beaching rate averaged across 1995–2014 ( $F_T^d(t_b^*)$ ) at the Aldabra Group and Seychelles Plateau (normalised by the annual mean) for (c) Class A and (d) Class C debris. The hatching indicates the approximate timing of the northwest monsoon (December to February) and southeast monsoon (June to August). The shading represents the range in monthly beaching rates.

Aldabra relies on a relatively uncommon pathway in which debris is entrained into eddies in the northern Mozambique Channel and transported towards Madagascar, before entering the North Madagascar Current upstream of Aldabra, and subsequently beaching (Supplementary Animation 1). This pathway is improbable during the southeast monsoon as a strong North Madagascar Current (Backeberg and Reason, 2010) results in debris rapidly beaching along the east African coast. As a result, transport from Comoros to Aldabra is generally only feasible during the northwest monsoon and subsequent intermonsoon. In contrast, there is very little seasonal variability in Class A beaching rates across the Seychelles Plateau. There is some seasonality at individual islands, however. For instance, during the northwest monsoon, the South Equatorial Countercurrent shifts towards the south near the Seychelles Plateau (Schott et al., 2009), facilitating the eastward transport of debris from the highly populated island of Mahé towards Praslin. Conversely, the South Equatorial Countercurrent shifts to the north during the southeast monsoon, and debris is more likely to be transported westward from Mahé due to the northwestward Stokes drift over the Seychelles Plateau at this time. Indeed, the seasonal pattern for Class A debris beaching at Silhouette Island, west of Mahé, is in exact antiphase to the pattern at Praslin (see Supplementary Table 1).

In the case of Class C debris (Fig. 6(d)), the Aldabra Group again sees a strong seasonal cycle, with a clear minimum in beaching rate during the southeast monsoon, and a peak following the northwest monsoon. The seasonal cycle at the Seychelles Plateau is weaker than for the Aldabra Group, with two peaks broadly aligned with the intermonsoons, with the former driven by debris arriving from south Asia, and the latter by debris arriving from southeast Asia. However, the second peak associated with debris from southeast Asia is driven by a small number of large beaching ‘pulses’, and it is not clear from Fig. 6 whether this seasonal cycle is robust. Alternatively, we can observe that, in most years, log-transformed beaching rates are dominated by a single clear sinusoidal peak at most sites we considered. By analysing (log) beaching rates in the frequency domain and extracting the phase of the component with a period of 1 year (see Section 2.5.3), we can estimate during which season beaching rates consistently peak. This is summarised for Class C debris in Table 2 (corresponding tables for Class A and Class B debris are given in the Supplementary Tables 1–2). These seasonal cycles are significant for almost all islands considered.

Table 2 suggests that the seasonality of beaching rates across Seychelles is actually in phase for Class C debris of terrestrial origin, with a significant peak predicted in March or April (i.e. the end of the northwest monsoon and the subsequent intermonsoon) for almost all islands in Seychelles. This peak shifts slightly earlier in the year for less buoyant classes, but remains during the northwest monsoon for Class A and Class B debris beaching at most islands in Seychelles. The strength of this seasonality (quantified by the ratio of the beaching rate during the highest and lowest three months), however, is considerably larger for the Outer Islands of Seychelles, particularly for the Aldabra, Farquhar and Alphonse Groups.

### 3.2.2. Seasonal variability from marine sources

We gain further insight into the physical drivers of this seasonality by repeating this spectral analysis for marine-sourced debris  $F_{Mb}^i(x, y, t_b)$  (Section 2.5.2), i.e. the beaching rate at site  $i$  and time  $t_b$  given a uniform rate of debris generation at  $(x, y)$ . This reveals whether debris entering the ocean at a uniform rate is more likely to beach at  $i$  in certain parts of the year. Plotted in Fig. 7(a) is the correlation in time between the log of the marine-sourced debris beaching flux,  $\log(F_{Mb}^i)$ , and the annual component,  $\mathcal{F}(\log(F_{Mb}^i))$ . Although our analyses suggest that Indonesia is the dominant source of Class C debris for Aldabra, Fig. 7(a) shows that debris generated across most of the Indian Ocean is significantly ( $p < 0.01$ ) more likely to beach at Aldabra in certain seasons. The phase of the seasonal cycle  $\mathcal{F}(\log(F_{Mb}^i))$ , given in Fig. 7(b), represents when in the year debris is most likely to beach at the Aldabra Group, for a uniform

rate of debris production. Importantly, the entire region is perfectly in phase. This may be surprising, as the drift time to the Aldabra Group varies considerably across the Indian Ocean. If the seasonality of Class C debris beaching at the Aldabra Group depended on remote forcing (e.g. currents, winds and waves at the debris source region, or along its transport path), we would expect considerable spatial heterogeneity in Fig. 7(b).

Instead, this figure demonstrates that the seasonality of Class C debris beaching at the Aldabra Group is dominated by local forcing, specifically the monsoonal variation in the winds. During the northwest monsoon, winds around Aldabra are relatively weak (Fig. 2) and westward zonal surface currents are proportionately more important. As a result, debris arriving at Aldabra during the northwest monsoon is sourced, on sub-seasonal timescales, east of Aldabra, from the southern tropical Indian Ocean. Conversely, during the southeast monsoon, strong southeasterly winds blow over the Aldabra Group (Fig. 2) and the source region for Aldabra (on sub-seasonal timescales) shifts to the southeast of Aldabra, in the southern subtropical Indian Ocean. Since winds over the southern Indian Ocean never have a strong northerly component (Fig. 2), there is no efficient pathway for Class C marine debris to reach the subtropical southern Indian Ocean from Indonesia (or other south(east) Asian sources), and therefore no route to Aldabra. As a result, it is improbable for Class C debris from the eastern or northern Indian Ocean to reach Aldabra during the southeast monsoon. This wind-driven mixing barrier can be clearly seen in Fig. 7(b) as the sharp phase discontinuity extending southeastwards from the Aldabra Group.

In this way, the monsoonal winds over the Aldabra Group act as a debris ‘switch’ for highly buoyant debris, alternating the principal debris source between the southwestern Indian Ocean (with minimal debris sources), and the remainder of the basin. The dominance of winds in determining the seasonality of beaching at the Aldabra Group remains valid for Class B debris (Supplementary Fig. 16), but not for Class A debris (Supplementary Fig. 15), where the seasonality instead appears to be dominated by the strength and position of the North Madagascar and South Equatorial Currents. The phase of the Class C seasonal cycle with respect to the Seychelles Plateau (Supplementary Fig. 20) is similar to the Aldabra Group, but due to the more northerly position of the Inner Islands, winds associated with the southeast monsoon do not have as extreme a blocking effect as with the Aldabra Group. Additionally, as hinted at by the greater spatial heterogeneity in Supplementary Fig. 20, remote forcing may play a greater role for debris beaching at the Inner Islands.

At some remote islands, such as Aldabra, most beaching debris is actually related to fishing activities rather than terrestrial input (Burt et al., 2020). As a result, the seasonal patterns identified for Aldabra may not necessarily be the same for fishing-related debris as terrestrial debris due to the very different input distributions. Monthly beaching rates for ALDFG are given by the quantity  $B_b^i(t_b^*)$  (Section 2.5.2), and are plotted in Fig. 8(a)–(b). This shows that, although peaks are not perfectly aligned with predictions for debris of terrestrial origin, purse-seine and longline associated debris beaching at the Aldabra Group will still likely peak during the northwest monsoon or subsequent intermonsoon, and fall to a minimum during the southeast monsoon. Therefore, although there may not be a clearly defined peak of debris accumulation at Aldabra in March as suggested by Table 2, we would still expect debris accumulation to be significantly enhanced during the northwest monsoon and subsequent intermonsoon, as compared to the southeast monsoon. For fishery-related debris accumulating at sites across the Seychelles Plateau, our analyses suggest that the seasonal cycle would be similar to the Aldabra Group, but slightly broader and shifted later in the year (Supplementary Fig. 22). This may be due to the more central position of the Seychelles Plateau with respect to intensive fisheries in the western Indian Ocean, as well as the seasonality of fishing activities in the region, which is incorporated into these analyses.

It is also possible to investigate the seasonality of the marine-sourced

**Table 2**

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

Beaching site	Seasonal cycle peak	Seasonality strength
Aldabra	March	35.1
Assumption	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

debris source flux or ‘risk’ ( $F_{Ms}^i(x, y, t_s)$ , see Section 2.5.2), i.e. the likelihood that debris entering the ocean at  $(x, y)$  during month  $t_s$  eventually beaches at site  $i$  (Supplementary Figs. 23–30). This analysis demonstrates that the risk to remote islands of marine debris generation is greater at some parts of the year than others. For instance, our analysis suggests that Class C ALDFG generated by fishing activities throughout most of the Indian Ocean is most likely to beach at the Aldabra Group if it enters the ocean from August to December (Supplementary Fig. 25), indicating that these may be months during which fishing activity is most at risk of generating harmful ALDFG. However, in contrast to the beaching flux, the phase of the seasonal cycle associated with risk is more spatially heterogeneous, indicating that both local and remote forcing drives risk.

### 3.2.3. Interannual variability

Although our analyses suggest that temporal variability in beaching rates at remote islands in the western Indian Ocean is dominated by seasonal variability from the monsoons, there is still considerable interannual variability (6(a)–(b)). Northerly wind anomalies across the southern Indian Ocean are associated with IOD and ENSO events (Yu et al., 2005) and, as described in Section 3.2.1, the meridional component of winds over the southern Indian Ocean associated with the monsoons appears to be driving the seasonal cycle in beaching rates across Seychelles. We may therefore expect IOD and ENSO phases to amplify the seasonal cycle for highly buoyant debris beaching at Seychelles, increasing northwest monsoon beaching rates for debris from southeast Asia during positive phases, and further suppressing southeast monsoon beaching rates during negative phases.

To test this, we pass marine-sourced debris beaching flux  $F_{Mb}^i(x, y, t_b)$  through a low-pass filter with a cutoff frequency of 1.25 years, to remove intra-annual variability from the signal. We then carry out a lagged correlation of the filtered time-series against the Dipole Mode Index (DMI), an IOD index based on SST gradients across the equatorial Indian Ocean, and NINO3.4, an ENSO index based on mid-Pacific SST. Fig. 9(a)

shows an analogue of Fig. 7(a) based on correlations of Class C debris beaching rates at the Aldabra Group with DMI. Although correlations are unsurprisingly lower than for the seasonal cycle, interannual variability in Class C beaching rates at Aldabra are correlated with DMI for source sites across much of the north and northeastern Indian Ocean. Additionally, the spatial pattern of these correlations strongly resembles the pattern in Fig. 7(a) from the seasonal cycle, supporting the hypothesis that the IOD may amplify the seasonal cycle through modulation of meridional winds in the southern Indian Ocean. Fig. 9(b) shows correlations between the total Class C beaching rate at all sites considered in this study, and DMI, as a function of DMI lead time. DMI correlates significantly with beaching rates at islands in the Aldabra, Farquhar, and Alphonse groups, which is expected as these are the same island groups that saw the most dramatic modulation by the seasonal cycle. DMI also correlates most strongly with beaching rates with a lead time of a few months, which supports the hypothesis that the IOD modulates the seasonal cycle as the monsoonal winds also lead peak Class C beaching rates (the seasonal cycle peaks in Table 2 within Seychelles generally occur just after the northwest monsoon, during the subsequent intermonsoon).

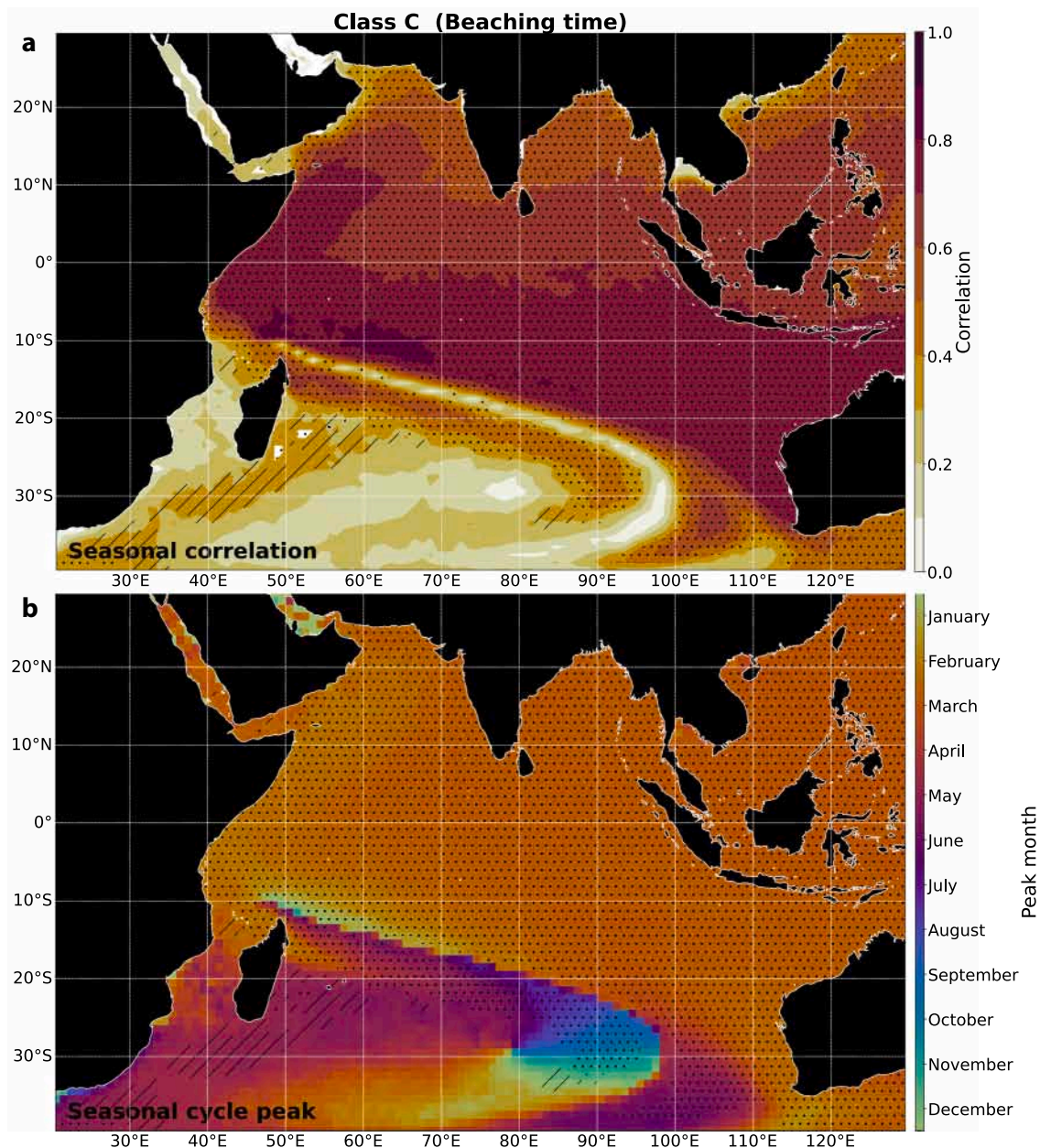
Correlation with the NINO3.4 index returns higher correlation coefficients compared to DMI, which is consistent with the partial correlations with the surface wind field given in Yu et al. (2005), as ENSO appears to be associated with stronger meridional wind anomalies closer to Aldabra. However, due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

### 3.3. Comparison with beached debris observations

#### 3.3.1. Debris accumulation at Aldabra

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

Calculating the average annual beaching rate at Aldabra across  $\mu_b^* - \mu_s$  parameter space (from 1999 to 2014 to allow a longer spin-up for lower values of  $\mu_s$ ) reveals (1) that the beaching rate at Aldabra is insensitive to the parameter  $\mu_b^*$  in the range  $1/\mu_b^* \in [5, 60]$ , and (2) that the inferred average bulk beaching flux at Aldabra is most consistent with  $100d < 1/\mu_s < 400d$ , depending on the windage coefficient (Supplementary Figs. 31–33). This is not to suggest that all marine debris has a sinking rate in this range ( $\mu_s$  is a variable which depends on debris composition, geometry, and biofouling rates), but it does indicate that most debris, by mass, is likely to have a sinking rate on the order of months to a year. This is consistent with the findings of Fazey and Ryan (2016), Kaandorp et al. (2020), and Koelmans et al. (2017). As a result, Class A debris is unlikely to represent a significant fraction of debris beaching at Aldabra by mass. Additionally, it is also unlikely that most debris beaching at Aldabra has a sinking timescale of multiple years, since we would expect a significantly greater mass of terrestrial debris to have accumulated on Aldabra if this were the case.



**Fig. 7.** (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 (Section 2.5.3). 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. Corresponding plots for other debris classes and the Seychelles Plateau can be found in Supplementary Figs. 15–21.

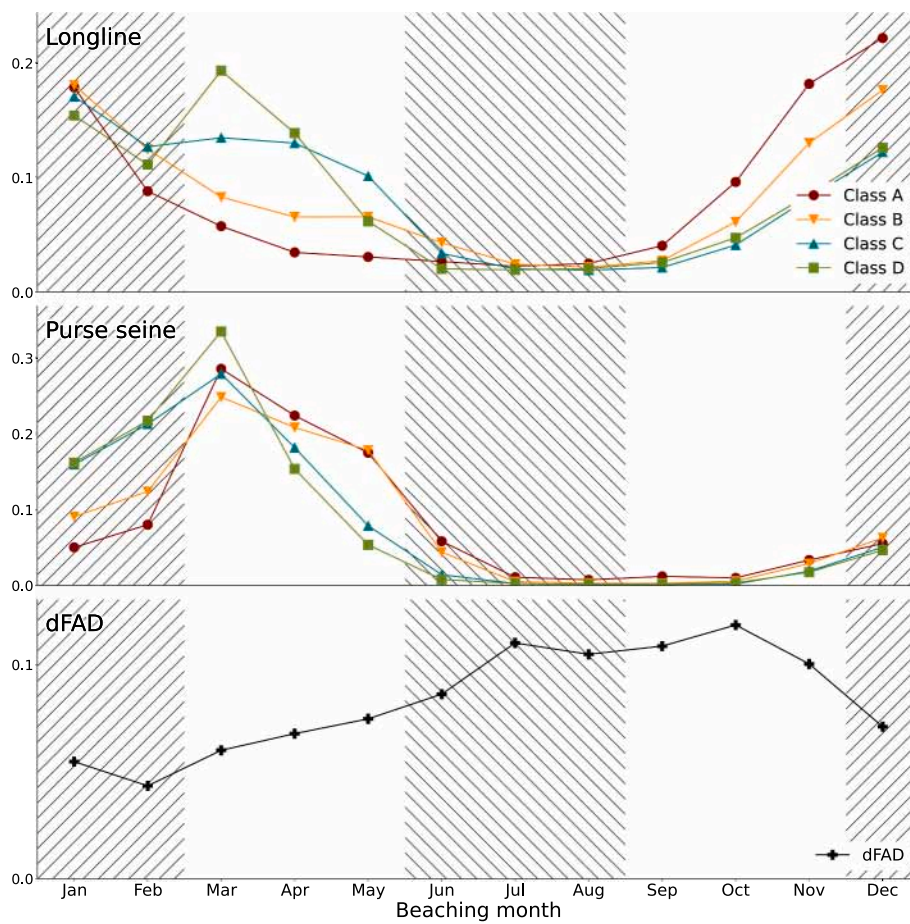
### 3.3.2. Temporal variability of drifting Fish Aggregating Devices across Seychelles

Drifting Fish Aggregating Devices (dFADs) are buoyant drifters used primarily by purse-seine fisheries to aggregate tuna. The majority of these dFADs are tracked remotely using satellite-transmitting GPS-equipped buoys and, as a result, dFADs are one of the only types of marine debris that can be tracked directly from source to sink Imzilen et al. (2021). Macmillan et al. (2022) identified over 3000 dFAD beaching events across Seychelles, and analysed beaching rates and seasonality. This provides a useful test-case for our trajectory analysis, but the physics of dFAD transport do not correspond well to any of our marine debris classes A-C due to their long drogue. As a result, we define ‘dFADs’ as a new Class of marine debris with  $\mu_s = 1800d$  (dFADs are large, buoyant, and non-biodegradable),  $\mu_b^* = 20d$  (based on observed

beaching rates of dFADs, see Supplementary Fig. 4), and physical scenario C0 (surface currents only, Imzilen et al. (2021)). We compute the predicted seasonal distribution of dFAD beachings based on the methodology described in Section 3.2.2, taking into account the seasonality of dFAD deployments. Our simulations reproduce a relatively muted seasonal cycle of dFAD beachings at Aldabra (Fig. 8(c)) and a pronounced peak in dFAD beaching rates within the Seychelles Plateau during the intermonsoon following the northwest monsoon (Supplementary Fig. 22), both of which correspond well to observations (Isla MacMillan, *personal communication*).

### 3.3.3. Countries of origin for debris beaching across Seychelles

**3.3.3.1. Aldabra (Seychelles).** In addition to quantifying the total mass



**Fig. 8.** (a)–(b) Monthly beaching rate from 1995 to 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 CO (Imzilen, 2021). Supplementary Fig. 22 is the analogous plot for the Seychelles Plateau.

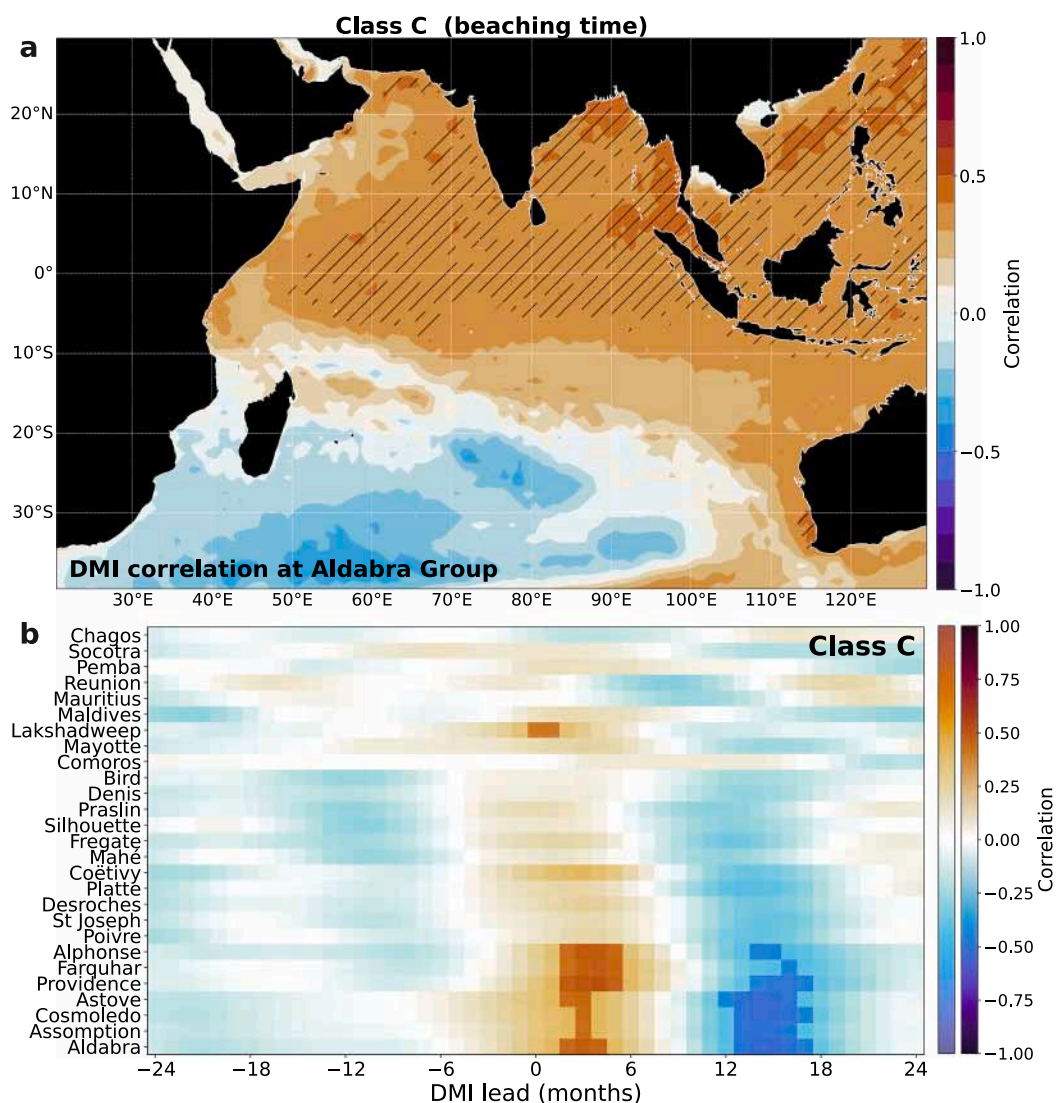
of debris on Aldabra, Burt et al. (2020) identified the origin of 45 PET bottles. In Table 3, we compare the predicted distribution of countries of origin for Class B and Class C debris beaching at Aldabra, to the distribution of countries of manufacture for intact PET bottles found at Aldabra.

For several countries of origin, there is agreement between the two datasets, particularly for Class C debris. Of the 5 largest sources of Class C debris predicted by the model, bottles were found on Aldabra from 3 (Indonesia, India, and South Africa). Indonesia was the largest source of Class C debris in our model, and was the second largest country of manufacture in the sample from Aldabra. However, there are some significant differences. This in itself is not unexpected. The sample size (45) of PET bottles in Burt et al. (2020) is small, and the sample is likely biased against bottles with longer drift times, as only bottles with intact labels could have their country of manufacture identified. Additionally, the country of manufacture of a bottle is not necessarily the same as the country where a bottle entered the ocean. However, the particular countries associated with model-observation disagreement provide interesting insights into the sources of debris for Aldabra.

The most obvious discrepancy between the two datasets is China. In our analysis, China was responsible for a negligible proportion of debris of any class accumulating at Aldabra (<0.1%), but was responsible for the manufacture of almost half of all bottles actually found on Aldabra. Although our debris classes may be an imperfect representation of the physics driving PET bottle transport, no realistic combination of  $\mu_b^*$ ,  $\mu_s$ , or physical scenario results in a significant flux of marine debris from China to Aldabra. More likely is an explanation suggested by Duhec et al.

(2015), that a large proportion of labelled items from Asia accumulating at beaches in Seychelles were thrown overboard or lost from shipping activities (commercial, leisure, and fishing) in the vicinity of Seychelles. Indeed, this is strongly supported by Fig. 5, which shows that Aldabra is directly downstream of the extremely busy shipping lanes linking SE Asia to the Atlantic. This same explanation could account for the number of bottles found on Aldabra from Thailand and Singapore, both of which were more than an order of magnitude more abundant in the cleanup than our predictions based on trajectory analysis. There is also significant fishing activity by vessels which may be avoiding tracking systems (Welch et al., 2022), so it is possible that these could account for some of this debris. Shipping aside, another possibility is that some waste entering the ocean from countries such as Indonesia was manufactured abroad. This could be due to the export of goods for sale and/or the export of waste. Indonesia is a major waste importer, but the main export partners are in Europe and the Americas (Greenpeace East Asia, 2019), so this cannot account for the discrepancies in Table 3. We do not have data on the proportion of bottled drinks sold in Indonesia (or other identified source countries) which are foreign imports, but imports would have to account for almost all PET bottles sold in these countries to explain the discrepancies in Table 3. We therefore suggest that disposal at sea is the most likely explanation for the discrepancies we have found.

Disposal at sea cannot, however, explain the absence of bottles from the Philippines amongst PET bottles found at Aldabra relative to our predictions. We suggest the most likely explanation is that the value for  $\mu_b^*$  diagnosed from drifters based on a global dataset is inappropriate for



**Fig. 9.** (a) Correlation between (log-transformed) time-series of debris beaching at the Aldabra Group from each cell, and the IOD Dipole Mode Index (DMI). Hatching indicates that the time-series in a cell correlates with DMI ( $p < 0.05$ ), 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).

the complex archipelagic coastline and bathymetry around the Philippines, resulting in our analyses underestimating the local beaching rate for debris of Philippine origin, and therefore overestimating the quantity of debris entering the Indian Ocean. Alternatively, the very long drift time (Supplementary Fig. 8) may have resulted in most labels degrading.

**3.3.3.2. Alphonse (Seychelles).** Duhec et al. (2015) carried out a six week marine debris monitoring program at Alphonse during the southeast monsoon in 2013, and identified the country of origin for plastic and glass bottles (and caps) with intact labels. Duhec et al. (2015) found that 75 % of labelled items originated from southeast Asia (primarily Indonesia and Thailand, although two glass bottles were found from the Philippines), with 13 % originating from east Asia (mainly China). In stark contrast to Aldabra, beach clean-ups are carried out regularly on Alphonse, and the survey described by Duhec et al. (2015) only included debris that had beached within the six week period (rather than debris that had been accumulating for many years, as at Aldabra). Due to the prediction of a seasonal cycle in beaching rate, it is important to compare these observations at Alphonse to our predictions for the

southeast monsoon season, rather than an annual mean. During June–July (the main months during which the survey at Alphonse took place), our analysis suggests that 76.4 % of Class C debris beaching at Alphonse originated from southeast Asia, primarily from Indonesia and the Philippines. As with Aldabra, our results do not predict that Thailand or China are significant sources of debris, suggesting that most bottles of Thai or Chinese origin beaching at Alphonse likely entered the ocean at sea. Although south Asia is likely a significant source of Class C debris over the course of a year (Fig. 4), only 6.4 % of Class C debris expected to beach at Alphonse in June and July originated from India or Sri Lanka. Instead, we expect most of this debris to accumulate at Alphonse during the northwest monsoon and subsequent intermonsoon, explaining the absence amongst debris analysed by Duhec et al. (2015), who also predicted this seasonality using backtracking simulations.

In any case, 6 weeks is an insufficient accumulation time to accurately represent debris sources for remote islands in the western Indian Ocean. On average, at least half a year of accumulation is required to reasonably represent the source distribution of Class C debris at Alphonse but, even after a year of accumulation, the relative importance of major debris sources remain statistically indistinguishable (Supplementary Figs. 34–36). This suggests that future observational studies

**Table 3**

Distribution of countries of origin or manufacture from this study (based on Class B and 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.

Origin	Class B (%)	Class C (%)	Burt et al. (2020) (%)
China	<0.1	<0.1	46.7
Indonesia	54.5	50.6	13.3
Thailand	<0.1	0.5	8.9
India	0.5	5.3	6.7
Malaysia	0.5	1.6	6.7
Singapore	<0.1	<0.1	4.4
South Africa	<0.1	5.0	2.2
Comoros	20.7	4.4	–
Tanzania	13.1	3.7	–
Philippines	3.8	21.2	–
Madagascar	2.0	0.7	–
Sri Lanka	0.2	2.4	–
Timor-Leste	1.8	1.7	–
Kenya	1.7	0.3	–

intending to assess sources of debris for remote islands should either target beaches that are not regularly cleaned, or should monitor debris accumulation over many years.

**3.3.3.3. Outer Islands of Seychelles.** Based on a sample of 189 labels found on four islands in Seychelles (Alphonse, Coëtivy, Astove, and Platte), The Ocean Project Seychelles (2019) found that 49 % of labels originated from southeast Asia and specifically noted that the most common countries of origin were Indonesia (26.5 %), Mauritius (12 %), and Malaysia (10.2 %). This is broadly in line with the findings of the other debris monitoring programmes in Seychelles, although one exception is the large proportion of debris originating from Mauritius. Given that no other studies assessing sources of debris in Seychelles noted a large proportion of debris from Mauritius (<4% at Alphonse (Duhec et al., 2015), and there has been no mention at Aldabra (Burt et al., 2020)), it is possible that these items from Mauritius instead originated from nearby ships.

#### 4. Conclusions and implications for conservation

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

We estimate that a large proportion of debris beaching at Seychelles has drifted from southeast Asia (principally Indonesia) and, in the case of the Inner Islands, south Asia (primarily India and Sri Lanka). Since debris drifting from sources such as Indonesia will have been at sea for at least six months, this also increases the risk of invasive species and pathogen introductions through rafting from the eastern and northern Indian Ocean. These results emphasise the scale of the challenge facing small island developing states such as Seychelles, and underlines the need for multilateral discussions around waste management. Smaller and/or less buoyant debris fragments may originate from East Africa (mainly Tanzania) and from within Seychelles itself, although these are unlikely to account for most beaching debris by mass, particularly for the Outer Islands of Seychelles. Our results suggest that Seychelles as a whole is at very high risk from debris that has been lost from shipping and fisheries in the Indian Ocean, and that most debris accumulating at Seychelles from Malaysia, Thailand and, in particular, China, is likely

associated with these activities. This prediction could be used to initiate discussions with shipping companies and cruise operators to reduce these sources of marine pollution. We have also found that abandoned, lost or otherwise discarded fishing gear has a high probability of beaching within Seychelles, directly polluting island ecosystems. Beaching purse-seine fragments are likely associated with fishing activity around Seychelles, but longline fragments could feasibly drift from fisheries across the southern Indian Ocean. Greater enforcement by regional governments of MARPOL Annex V (Marine Environment Protection Committee, 2017), forbidding the discharge of fishing gear and other plastics at sea, would reduce these sources of pollution, particularly for Aldabra.

We have also found that there is likely to be significant predictability in marine debris accumulation rates across Seychelles, primarily from a strong seasonal cycle dominated by the monsoons. For classes of debris experiencing a significant push from the winds, our analysis suggests debris from terrestrial sources and fisheries are most likely to beach at Seychelles (but most significantly the Aldabra, Farquhar, and Alphonse Groups) during the northwest monsoon and subsequent intermonsoon. Beach clean-ups should ideally take place after peak beaching (i.e. May to June for much of Seychelles) to reduce the likelihood of beached plastics breaking down into smaller unmanageable fragments and impacting ecosystems. Due to considerable seasonal and stochastic variability in beaching rates, future observational studies aiming to assess the provenance of debris beaching at remote islands should either target beaches that are not subject to regular clean-ups, or monitor debris accumulation over multiple years. We have also proposed a mechanism by which ENSO and the IOD may modulate this seasonal cycle, and have presented some evidence to suggest that beaching rates of high-windage debris at the more southerly island groups within Seychelles may be greater during and following positive IOD phases. These predictions may be helpful for practitioners deciding when to carry out beach cleanup operations.

There is reasonable agreement between our predictions and the limited quantitative observations of marine debris that are available from across Seychelles. Key discrepancies with observations have highlighted the importance of shipping lanes as a source of marine debris for remote western Indian Ocean islands. Nevertheless, it is important to remember that our trajectory analysis relies on a large number of poorly constrained parameters. There is an urgent need for further studies on the rate of marine macrodebris fragmentation, biofouling, and sinking. Despite the number of marine debris modelling studies incorporating windage into simulations and acknowledging the important role it plays in determining drift trajectories, there are limited publicly available estimates of appropriate windage coefficients for common classes of marine debris. Additionally, the windage coefficient will likely change with time, as debris loses buoyancy and/or fragments. We remind the reader that our debris classes are just examples, rather than validated representatives of bulk debris properties. Our analyses also lack some processes such as tides and defouling of debris at depth, with unclear consequences for long-distance debris transport (Cózar et al., 2014; Suanda et al., 2018; Deschepper et al., 2019). Finally, considerable uncertainty remains in the input function of marine debris into the ocean. Nevertheless, a strength of this study is that our results can be easily recomputed for different combinations of sinking rate, beaching rate, and windage so, if improved constraints in the future demonstrate that our classification of debris (into our four classes A-D) is inappropriate, it will be straightforward to recompute results with the dataset and scripts provided in the Supplementary Datasets.

Supplementary data to this article can be found online at <https://doi.org/10.1016/j.marpolbul.2022.114497>.

#### CRedit authorship contribution statement

**Noam Vogt-Vincent:** Conceptualisation, Methodology, Software, Validation, Formal analysis, Investigation, Data curation, Writing –

Original Draft, Writing – Review & Editing, Visualisation. **April Burt:** Conceptualisation, Writing – Review & Editing. **David Kaplan:** Resources, Formal analysis, Writing – Review & Editing. **Satoshi Mitarai:** Writing – Review & Editing, Supervision. **Lindsay Turnbull:** Conceptualisation, Writing – Review & Editing, Supervision. **Helen Johnson:** Conceptualisation, Methodology, Writing – Review & Editing, Supervision.

## Declaration of competing interest

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

## Data availability statement

All data required to recompute results for arbitrary parameters are archived at the British Oceanographic Data Centre as [Supplementary Dataset 1](#). The matrices described in this manuscript corresponding to the four debris classes are archived as [Supplementary Dataset 2](#). Code required to reproduce figures in this manuscript, and more figures for further parameter combinations, are archived as [Supplementary Dataset 3](#), with documentation available [here](#). The dFAD deployment data cannot be published due to a confidentiality agreement. Requests for access to dFAD deployment and tracking data should be addressed directly to the [Ob7 pelagic ecosystem observatory](#) using the following email address: [adm-dblp@ird.fr](mailto:adm-dblp@ird.fr).

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