¹ Supplementary Note for ² The streaming of plastic in the Mediterranean Sea

Alberto Baudena^{1*}, Enrico Ser-Giacomi^{2†}, Isabel Jalón-Rojas^{3†}, François Galgani⁴, and Maria Luiza Pedrotti¹

 ¹Sorbonne Université, CNRS, Laboratoire d'Océanographie de Villefranche, UMR 7093 LOV, Villefranche-sur-Mer, France
 ²Department of Earth, Atmospheric and Planetary Sciences, Massachusetts Institute of Technology, 54-1514 MIT, Cambridge, MA 02139, USA.
 ³CNRS, UMR5805 EPOC, University of Bordeaux, 33615 Pessac, France
 ⁴French Research Institute for Exploitation of the Sea (IFREMER), Bastia, France

* Correspondence to: alberto.baudena@gmail.com.

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Supplementary Note

10 1 In situ vs model comparison: sensitivity analysis

11 **1.1** Plastic categories, horizontal diffusion K_h , and half-life on the beach 12 T_W

Virtual particle concentrations in each Mediterranean sub-basin were compared to the eight
categories of *in situ* concentrations of plastic debris (Material and Methods, Section 1.1). In
Scenario M, significant correlations were found between the *in situ* plastic-debris concentrations
and the model values in the following categories (Supplementary Fig. 4, first panel):

• **Category 2:** (particles of size 5–20 mm, g/km^2), $R^2=0.91$, p<0.05 (Pearson test);

- Categories 3,4: (particles of size greater than 20 mm and of all sizes, respectively, g/km²), R²=0.92 and R²=0.96 respectively, p<0.01;
- Categories 7,8: (surface covered by plastic debris and by plastic fragments, respectively, m²/km²), R²=0.93, R²=0.96, respectively, p<0.01.

²² When *in situ* plastic-debris concentration was expressed as the number of items per km², no ²³ significant correlation was found (p>0.05). This could be due to the fact that the number of ²⁴ plastic items depends on fragmentation. This dynamic was not included in the present version ²⁵ of the model due to the lack of quantitative information on this process.

The correlations with the *in situ* concentration were calculated for each of the 16 scenarios separately. Significant correlations were recurrently found for the same plastic concentration categories (Supplementary Fig. 4, lower panel): Category 2 (except for the scenario with T_W =50 days K_h =15 m²/s), and Categories 3, 4, 7, and 8. The persistence of the correlations across the different scenarios strengthens the decision to average together the outputs of the 16 scenarios.

1.2 Proportions of particles emitted from cities p_C , rivers p_R , and vessels p_V

The correlations with the *in situ* concentration were calculated using four different proportions 33 of particles emitted from cities p_C , rivers p_R , and vessels p_V (upper table of Supplementary 34 Fig. 5). Case 1 (40-40-20%) is usually assumed in the literature (1), Case 2 (62-32-6%) was 35 recently estimated by (2) for the Mediterranean Sea, while Cases 3 and 4 represent intermediate 36 proportions. Significantly correlations were recurrently found for the same plastic concentration 37 categories (Supplementary Fig. 5A–D, Categories 2, 3, 4, 7, and 8), but the largest correlations 38 were found with the proportions 50-30-20% (Supplementary Fig. 4, upper panel). These results 39 indicate a robustness of the model with respect to these parameters, and corroborate the p_C , p_R , 40 and p_V proportion used. 41

42 **2** In situ vs model comparison: distance from shore

To assess the variability in particle concentration as a function of distance from shore, in 43 situ concentrations at the different stations were regrouped according into four shore-distance 44 classes: stations (i) less than 20 km (inner continental shelf); (ii) between 20 and 40 km (middle 45 continental shelf); (iii) between 40 and 60 km (far continental shelf); and (iv) more than 60 km 46 from the coastline. For each distance class, the mean concentration and standard error were 47 obtained. In the model, the distances from the shore were calculated for each trajectory point, 48 at each daily time step. The distances obtained in this way were then regrouped according to the 49 distance classes specified above. Finally, the *in situ* and virtual concentrations were normalized 50 to permit comparison. 51

For Scenario M, there was agreement between the simulated and observed concentrations for Categories 1 and 2 (Supplementary Fig. 6A–B; R^2 =0.91 and R^2 =0.94, respectively, p<0.05, Pearson test). Large particle concentrations were found close to the shore, with few particles in

a strip between 20 and 60 km, followed by an increased concentration in the open sea (> 60km). 55 This result is consistent with previous measurements of plastic distribution as a distance from 56 shore in the Mediterranean Sea (3). However, a larger observed concentration close to the coast 57 could also be explained by a greater number of large debris (heavier) occurring close to land 58 sources. To verify this hypothesis, we calculated the size distribution for each class and we 59 normalised it, so that the sum of the abundance of a given class was equal to 1 (Supplementary 60 Fig. 6C). The size distribution did not change considerably when changing the shore distance. 61 A slight difference is present in class 40–60 km, likely due to the relatively number of in situ 62 stations (7 out of 122). The results show that the distribution of plastic concentration as a dis-63 tance from the shore is not explained by a greater occurrence of larger items close to the shore. 64 Furthermore, they indicate the coastal area as a region of debris retention, in accordance with 65 our results (e.g. Fig. 5, Supplementary Fig. 12). 66

3 Simulated plastic budget

The total number of particles that beached, sank, or that were on the surface (expressed as a percentage of the total number N of particles released between January 2013 and December 2016) was calculated for each month of 2013–2016 for Scenario M (Supplementary Fig. 7, upper panel). The result refers to the ensemble average obtained from 4 biofouling times (50, 100, 150, and 200 days; M 1.2.5 and 1.2.6), shown separately in the second panel of Supplementary Fig. 7.

The number of sinking and beached particles increased linearly with time, representing, at the end of December 2016, \sim 86.5 and \sim 11.5% of particles, respectively. Floating particles constituted the remaining \sim 2.2%; this percentage oscillated, increasing during winter and decreasing during summer. This could be due to the fact that the number of particles released from the rivers was larger during the winter months or that storm frequency was maximal during winter; this fostered both particle beaching and washing-off. Washing-off effect was dominant over
beaching in winter: the number of beached particles increased more slowly then (red line).
This increased the number of particles in the water during winter. However, we stress that these
results are still uncertain, and should be considered carefully, as both the beaching and washingoff modeling had limitations.

Remarkably, the number of particles in the water became steady after just 4-5 months from the 84 beginning of the release. It also did not change significantly when the biofouling time (sec-85 ond panel of Supplementary Fig. 7), the horizontal diffusion coefficient (K_h) and half-life of 86 beached plastics (T_W) were varied (Supplementary Note S.4 and Fig. 8). The third panel of 87 Supplementary Fig. 7 shows the mean time spent in the water by the particles according to 88 their origin and to the biofouling time. Particles spent on average between 20 and 30 days in 89 the water. This value was slightly lower for particles originating from terrestrial sources and 90 consistently higher for plastics originating from vessels (40 to 90 days), in accordance with (4), 91 but the values obtained here were slightly higher. This difference was likely due to the differ-92 ent description of washing-off and a greater T_W (one order of magnitude larger) in the present 93 study, which favored a longer time spent by the particles in the water (M 1.2.4). 94

⁹⁵ Despite the fact that 80% of particles were released from land sources, almost the half the ⁹⁶ sinking particles were originated from vessels (fourth panel of Supplementary Fig. 7). As the ⁹⁷ biofouling time increased, the number of sinking particles declined. The percentage of sinking ⁹⁸ particles with a biofouling time of 50 days (8% and 10% for vessel and land sources, respec-⁹⁹ tively) was different from the values obtained by (*4*) (6% and 3.2%, respectively). Again, this ¹⁰⁰ could be related to the different washing-off dynamics used in the present study.

4 Sensitivity of the plastic budget

Supplementary Figure 8 shows the plastic budget for each of the 16 scenarios, analogous to that reported in Section 3 (percentage of particles that were floating, beached or sunk during the period of particle release, 2013–2016). The trends did not change significantly across the 16 different panels. The number of beached plastics decreased to about 5% when T_W was increased from 25 to 100 days, and when the horizontal diffusion K_h was decreased from 15 to 0 m²/s. This decrease was offset by a corresponding increase in biofouled particles. The percentage of particles in the water, on average, did not seem to be affected by changes in T_W or K_h .

5 Validation of the advective time length

Here we discuss the decision to advect the virtual particles for a period of \sim 380 days (the ad-110 vective time). This represented a trade-off between an acceptable representation of the cycle of 111 plastic particles at sea and the optimization of computational performances. As seen in Section 112 S.3, the percentage of virtual particles yet not beached or sunk, that is particles still floating 113 in the water at the end of the advective time, constituted only $\sim 2.2\%$ of the particles released 114 (upper panel of Supplementary Fig. 7). Such a value was consistent with the findings of (4). In 115 addition, it was consistent with the short mean time spent in the water by the particles (Supple-116 mentary Fig. 7, third panel), which was much shorter than the advective time. 117

Of the particles that were still floating in the water at the end of the simulation (Fig. 9), the majority (80%) were found less than 40 km from land (Fig. 9, first panel). In addition, these particles were beached and washed off a consistent number of times and with a greater frequency than the other particles (Fig. 9, second and third panels). This suggests that the majority of these particles remained close to the shore, where they were continuously beached and resuspended. It is likely that this process continued until they became permanently beached or sank ¹²⁴ due to biofouling. This coastal-retention dynamic has also been suggested by (5). For all these ¹²⁵ reasons, we considered that an advective time of \sim 380 days permitted a realistic simulation of ¹²⁶ the plastic-debris cycle in the Mediterranean basin. Finally, we note that the advective times ¹²⁷ used in previous Mediterranean Lagrangian studies targeting plastic debris were either shorter ¹²⁸ or similar that adopted here (e.g., (6–8))

6 Sensitivity test of crossroadness and plastic crossroads

¹³⁰ 6.1 Horizontal diffusion K_h and half-life on the beach T_W

The crossroadness fields found for each of the 16 scenarios separately were almost identical 131 (Fig. 10). The intensity slightly increased for higher T_W values, due to the fact that fewer 132 particles remained on the beaches. Remarkably, the positions of the plastic crossroads were 133 very similar across the different scenarios: the crossroad close to Mallorca island and the three 134 crossroads in the Adriatic basin did not change position. Three or four crossroads were always 135 present in the Cilician basin, three or four persisted in the Turkish coastal sector of the Aegean 136 sea, and five or six along the Algerian-Tunisian shore. Finally, two crossroads were always 137 found in proximity to the Bomba Gulf in Lybia, and a crossroad always close to the Suez Canal. 138 The ranking of the crossroads varied slightly, but the percentage of intercepted particles as a 139 function of the number of crossroads considered did not change significantly (not reported). 140 This quantity increased slightly with increasing T_W (fewer particles retained on the shore) and 141 with decreasing K_h (less dispersion). These considerations indicate a robustness of the plastic 142 crossroads with respect to changes in the horizontal diffusion coefficient K_h and the half-life of 143 plastics on the beach T_W . 144

145 6.2 Radius σ of the circular crossroads

The crossroadness sensitivity was tested against changes in the radius σ defining the extent 146 of the crossroads. Two additional σ values were tried: 0.2° and 0.3°. In both cases, a buffer 147 of 0.1° from land sources was imposed. A larger σ allowed the crossroads to intercept more 148 virtual particles, thus increasing the crossroadness (Fig. 11). However, the two crossroadness 149 patterns remained almost identical. Notably, in both cases the disposition of the most important 150 crossroads matched that for $\sigma=0.1^{\circ}$, the value chosen. The only exception was the emergence 151 of two crossroads (for both 0.2° and $0.3^{\circ} \sigma$), the first in the open sea in the Gulf of Sidra, 152 the second in the Ligurian-Provençal Sea, south of the Hyéres Islands. The only significant 153 change occurred to the ranking of the crossroads, mainly because they covered a larger surface. 154 These considerations illustrate that the distribution of the crossroads was robust to changes in 155 the crossroad area. 156

With σ =0.1°, 60 crossroads were necessary to intercept ~20% of the particles (Fig. 5), ~30 crossroads with σ =0.2°, and ~25 crossroads with σ =0.3° (not reported). The total surface area covered by these crossroads was 0.93%, 1.6% and 2.8%, respectively, of the Mediterranean surface. This highlights the capacity of the algorithm to decrease the total surface covered by the crossroads selection as σ decreases. Future higher-resolution studies could benefit from this aspect, improving the performance of the crossroads identification.

6.3 Buffer around land sources

¹⁶⁴ Crossroad could not be located inside the buffers around land sources. Here, we calculated the ¹⁶⁵ percentage of particles that stayed inside these buffer regions while varying their size. On in-¹⁶⁶ creasing the buffer size from 0.1° to 0.3° , the percentage of particles retained inside the buffers ¹⁶⁷ increased only slighlty, from ~67% to ~70% (Fig. 12); at a buffer size of 0.5° , it increased ¹⁶⁸ to ~76%. The percentage of particles intercepted by the crossroads also did not change significantly when the size of the buffer region was increased. This indicates the soundness of the
 crossroadness analysis with respect to this parameter.

171 6.4 Proportions of particles emitted from cities p_C , rivers p_R , and vessels 172 p_V

Plastic crossroadness and 20 most important crossroads were calculated using four different proportions of particles emitted from cities p_C , rivers p_R , and vessels p_V (upper table of Supplementary Fig. 5). The results (Fig. 13) show that the crossroadness slightly changed only in Case 2, due to the lower input from vessels, but the pattern remained identical. Neither did the crossroad disposition vary, but only the ranking. These results show the robustness of the crossroadness and the crossroads with respect to these parameters.

179 6.5 Biofouling time

Here we investigated the importance of biofouling in the crossroads identification by calculating 180 the total time that particles had spent in the water when they were intercepted by a crossroad, 181 and comparing it with the biofouling time. After a certain amount of time in the water, it was 182 possible that the particles were sufficiently biofouled and sank, and would therefore not be 183 intercepted by the crossroads. The upper panel of Fig. 14 shows the cumulative pdfs of the 184 time spent in the water by the particles intercepted by each of the first 10 crossroads. Some 185 crossroads, such as the 2nd or the 4th, mainly intercepted particles that had spent relatively 186 short time in the water ($\sim 95\%$ less than 50 days). For other crossroads, such as the 6th and the 187 8th, this percentage was lower (\sim 50%). Overall, however, the cumulative pdfs were similar. 188 On average, $\sim 78\%$ (96)% of the particles had spent less than 50 (150) days in the water at the 189 time of their interception. We note that the biofouling times adopted here ranged between 50 190 and 200 days. 191



¹⁹³ of particles released $N \simeq 1.472 \times 10^8$) intercepted by the first 10 crossroads as a function of ¹⁹⁴ the time spent in the water at the time of their interception (Fig. 14, lower panel) to determine if ¹⁹⁵ the ranking of the crossroads could be affected by the biofouling time. The number of particles ¹⁹⁶ intercepted did not change significantly for biofouling times larger than ~50–100 days; below ¹⁹⁷ that time period, only the ranking of few crossroads changed. These findings confirmed that the ¹⁹⁸ biofouling effect, even if not considered explicitly, did not affect significantly the crossroads ¹⁹⁹ distribution, as virtual particles were mainly intercepted during their first days in the water.

7 Sensitivity to shore steepness

In order to analyze the relative importance of shore steepness, we compared two scenarios: one 201 included the shore steepness (Scenario SS) and one did not (Scenario NSS). In Scenario NSS, 202 particles that reached land were automatically considered as beached. The standard method-203 ology (M 1.2.3) was applied in Scenario SS to determine beaching. Both scenarios were run 204 with the same K_h and T_W values (0 m²/s and 50 days, respectively). Particle concentrations 205 for the five Mediterranean sub-basins were calculated using the same method as for the other 206 scenarios (M 1.2.8). These quantities were compared with the eight categories of *in situ* particle 207 concentrations (Fig. 15, left panel). The correlation coefficients obtained with Scenario NSS 208 were significant for Category 2 (p<0.05), and 3 and 4 (p<0.01). However, R^2 was always less 209 than the corresponding value calculated with Scenario SS (Fig. 15, right panel). In addition, 210 Scenario NSS displayed no significant correlation in Categories 7 and 8. 211

The simulated particle concentrations as a function of distance from shore (see Section S.2 for calculation details) were compared with the *in situ* observations (Categories 1 and 2). Lower correlation coefficients were obtained for Scenario NSS (Fig. 16, left panels) compared to those for Scenario SS (Fig. 16, right panels: Category 1, NSS R²=0.79, SS R²=0.96; Category 2, NSS R²=0.84, SS R²=0.97). In addition, *in situ* and simulated distributions were not similar (p > $_{217}$ 0.05) for Scenario NSS, whereas they were for Scenario SS (p < 0.05).

These findings indicate that taking into account shore steepness improved the performance of the Track-MPD model, increasing the agreement with observations by improving the beaching description.

²²¹ 8 Sensitivity of surface sinking and net beaching rates on K_h ²²² and T_W

8.1 Horizontal diffusion K_h and half-life on the beach T_W

Supplementary Figure 17 shows the surface sinking and the net beaching rates for each of the 16 scenarios. In general, the two metrics did not change consistently across the different runs. Increasing the half-life of beached plastics T_W decreased slightly the number of beached particles and, in turn, increased the surface sinking rate. When increasing the horizontal diffusion K_h , the number of beached particles did not change significantly, while the surface sinking rate was smoothed.

8.2 Proportions of particles emitted from cities p_C , rivers p_R , and vessels p_V

The surface sinking and the net beaching rates were calculated using four different proportions 232 of particles emitted from cities p_C , rivers p_R , and vessels p_V (upper table of Supplementary Fig. 233 5). The results (Fig. 19) show that the beaching rate did not changed significantly. The surface 234 sinking rate was lower in Case 2 only, due to the lower input from vessels, but the pattern re-235 mained identical. This result is consistent with the fact that particles released from vessels spent 236 more time in water, and contributed to almost 50% of the surface sinking rate (Supplementary 237 Fig. 7, third and fourth panels). Overall, this sensitivity test shows the robustness of the surface 238 sinking and net beaching rates with respect to the proportion of particles emitted from cities p_C , 239

rivers p_R , and vessels p_V .

Qualitative comparison of net beaching rates and observa tions of beached plastic items

Poeta et al., 2016 (9) reported seasonal variations in plastic deposition on four beaches close 243 to Montalto (central Italy) between spring 2014 and winter 2015. Supplementary Figure 18 244 shows: (i) the mean values obtained by the authors (9); and (ii) the seasonal net beaching rate 245 predicted by the TrackMPD model for the same area (note that both quantities were normalized 246 to allow comparison). A good qualitative agreement emerged between the two metrics, except 247 for winter, during which the difference was more pronounced. These results are evidence that 248 the TrackMPD model provides good estimates of beaching rates at local and seasonal scales, 249 although further improvements are necessary. In addition, the net beaching rates calculated 250 here agree qualitatively with the findings of (10) and (11). The two studies observed massive 251 beach pollution in the Cilician basin, which our model predicts to be significanly affected by 252 net beaching rates (\sim 10–14 kg/km/day). 253

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²⁵⁵ 10 Qualitative comparison of surface sinking rates and ob ²⁵⁶ servations of seafloor plastic concentration

Supplementary Table 1 reports the literature on seafloor plastic surveys at several locations across the Mediterranean basin. Before comparing them with the surface sinking rates predicted by our model, we stress that the difficulty of this task was exacerbated by several factors. Different sampling techniques were adopted in the seafloor plastic observations (for instance, bottom trawls, which can be conducted only on muddy bottoms, or scuba-dive samplings). The minimum size of the plastic debris measured varied from study to study. In addition, plastic ²⁶³ concentration was measured either as items/km², as kg/km² or as items per kg of dry sediment.
²⁶⁴ The debris collected were composed of plastic with a density larger than seawater, which were
²⁶⁵ not modelled in the present study. Finally, campaigns were conducted on different years, and
²⁶⁶ this could obviously lead to important differences in the plastic concentrations measured.

Despite these premises, the highest concentrations of benthic plastics in the Mediterranean to 267 date were in Mersin bay (12) and in the Gulf of Antalya (13), both of them in the Cilician 268 basin. Consistent with these observations, our modeling results suggested that these regions 269 experienced the highest surface sinking rates in the Mediterranean (Figs. 4 and S17). One of 270 the lowest seafloor values recorded in the Mediterranean was on the Sardinian shelf (14), which 271 was also one of the regions with the lowest surface sinking rates in our model. The study of (15)272 in the Gulf of Lion describes a rather stable-over-time region of enhanced bottom plastic con-273 centrations around $5^{\circ}E$. This area was characterized by surface sinking rates which were larger 274 than in the rest of the Gulf of Lion. 275



Figure 1: Scenario M. First panel: Each column shows the percentage (y-axis) of virtual 279 particles (relative to the total number of particles released $N \simeq 1.472 \times 10^8$) intercepted by each 280 of the first 10 crossroads. The colors of the 3 rectangles forming each column indicate the 281 origin of the particles: red for particles released from cities, blue for rivers, and green for 282 vessels. The black and white lines inside the red and blue rectangles split the contribution of 283 different cities and rivers, respectively. For instance, the first crossroad intercepted $\sim 0.75\%$ 284 of particles coming from a single city (Antalya, Turkey), and $\sim 0.10\%$ from three other cities: 285 Mersin (0.04%); Alanya (0.04%); and Side (0.02%). The shapes of the symbols inside the 286 green rectangles indicate the basin in which the particles were released: left-pointing triangles, 287 western Mediterranean; dots, Tyrrhenian Sea; open circles, Adriatic Sea; crosses, Aegean Sea; 288 downward triangles, central Mediterranean; right-pointing triangles, eastern basin. To facilitate 289 the reading, the second panel shows the crossroadness and the 20 most important crossroads 290 for Scenario M, as reported in Fig. 3. 291



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Figure 2: Fraction of plastics mass which is biofouled (m_B/m) as a function of the time spent in

water, calculated using Expr. (2) (M 3.2.5). The curves are calculated considering four different biofouling times T_{BF} : 50, 100, 150, and 200 days.



Figure 3: Number of storms (expressed as percentage) in the Mediterranean coasts during the simulation period (2013–2017). The storms were detected from sea wave height time series and used to identify washing-off events (M 3.2.4)



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Figure 4: **First panel:** Sub-basin comparison of virtual particle concentrations (Scenario M) with the 8 *in situ* categories of observed plastic concentration (*x*-axis). Blue and red dots indicate R and R² coefficient values, respectively. Black stars, when present, identify a significant correlation between simulated and *in situ* concentrations (*: p<0.05; **: p<0.01; ***: p<0.001). **Second panel:** The same comparison, but for each of the 16 scenarios separately. The set of K_h and T_W values used is shown in the title of each subplot.





Figure 5: Upper table: list of the proportion of particles released from cities, rivers, and vessels (p_C , p_R , and p_V) used to test the sensitivity of the results with respect to these parameters. Panels A–D: sub-basin comparison of virtual particle concentrations (Scenario M) with the 8 *in situ* categories of observed plastic concentration (*x*-axis). The R and R² cales (*y*-axis) were calculated as in the first panel of Supplementary Fig. 4, for different values of p_C , p_R , and p_V (case 1–4, upper table).



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Figure 6: Distribution of in situ (blue columns) and virtual (orange columns) particle concen-324 trations at the location of the Tara Expedition stations as functions of the distance from the 325 shore. Virtual particle concentrations were obtained for Scenario M. In situ concentrations refer 326 to Categories 1 (panel A) and 2 (panel B) and are showed with relative uncertainties (standard 327 deviation: black error bars). Both concentrations (simulated and in situ) were normalized to 328 allow comparison. R^2 values for the correlation between the two distributions (Pearson test) 329 were 0.91 and 0.94, respectively (p<0.05 in both cases). Panel C: normalised particle size 330 distribution according to the distance from the shore. 331



Figure 7: (on previous page) First panel: Percentage of the total number N of virtual parti-337 cles released that, during the release period (January 1, 2013 to December 31, 2016; x-axis), 338 beached (red line), sank (black line), or remained floating (blue line). The three quantities 339 were obtained from the ensemble average of simulations with four different biofouling times 340 T_{BF} (50, 100, 150, and 200 days). These are shown separately in the second panel. Lines of 341 the same thickness refer to the values computed with the same biofouling time: thinnest lines 342 T_{BF} =50 days, thickest line T_{BF} =200 days; intermediate thicknesses T_{BF} =100 and T_{BF} =150 343 days. Third panel: Mean time (in days, y-axis) that particles spent in the water as a function of 344 the biofouling time T_{BF} (in days, x-axis). Black line: all particles; red line: particles released 345 from land sources (cities and rivers); blue line: particles released from vessels. Fourth panel: 346 Percentage of the N virtual particles (y-axis) that sank due to biofouling as a function of the 347 biofouling time T_{BF} (x-axis). Red line: particles released from land sources; blue line: particles 348 released from vessels. 349



Figure 8: Plastic budget, calculated as in the first panel of Supplementary Fig. 7, for each of the scenarios separately. The K_h and T_W values used are shown in the title of each subplot.

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Figure 9: First panel: Cumulative pdf of distances from shore of particles that were still floating 361 in the water at the end of the advective time. Second panel: pdf of number of washing-off 362 events for particles (i) still in the water (blue columns) and (ii) not in the water (in that they were 363 either beached or biofouled; orange columns) at the end of the advective time. The color of the 364 columns where the two distributions superpose is brown. Third panel: pdf of the frequency of 365 washing-off events per day of particles (i) still in the water (blue columns) and (ii) not in the 366 water (orange columns). The frequency was obtained by dividing the number of washing-off 367 events of each particle by the time it spent in the water. 368



Figure 10: Plastic crossroadness and the 20 most important crossroads computed for each of the 16 scenarios separately, as in Fig. 3. $\sigma = 0.1^{\circ}$. The K_h and T_W values used are shown in the title of each subplot.



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Figure 11: Plastic crossroadness and 20 most important crossroads calculated with two different σ values (first panel: $\sigma = 0.2^{\circ}$; second panel: $\sigma = 0.3^{\circ}$) for Scenario M. Method as in Fig. 3.



Figure 12: Scenario M. Blue line: percentage of the N virtual particles that remained inside the buffer region around each land source as a function of the buffer size (*x*-axis). Red line: corresponding percentage of the total number of virtual particles released from land sources.



Figure 13: Plastic crossroadness and 20 most important crossroads calculated using different proportion of particles released from cities, rivers, and vessels (p_C , p_R , and p_V ; upper table in Supplementary Fig. 5).



Figure 14: Scenario M. First panel: Cumulative pdf of the time spent in the water by the 398 virtual particles from their release until they were intercepted by a crossroad ($\sigma = 0.1^{\circ}$). Solid 399 lines: cumulative pdfs calculated by considering all the particles intercepted by one of the first 400 ten plastic crossroads; black dotted line: by considering all the virtual particles intercepted by 401 all the crossroads. For example $\sim 80\%$ of the virtual particles, at the moment at which they 402 were intercepted, had spent less than 50 days in the water since their release. Second panel: 403 As for the first panel but with the percentage relative to the total number of particles released 404 $N \simeq 1.472 \times 10^8$. 405



Figure 15: Comparison of simulated and *in situ* sub-basin particle concentrations calculated as in Fig. 4. Left panel: correlations for Scenario NSS, in which shore steepness was not taken into account. **Right panel:** correlations for the same scenario but with shore steepness taken into account (Scenario SS).



Figure 16: Comparison of *in situ* (blue columns) and virtual (orange columns) normalized 415 particle concentrations as a function of distance from the shore (calculated as in Fig. 6). In 416 situ particle concentrations refer to Category 1 (first row) and 2 (second row) and are showed 417 with relative uncertainties (standard deviation: black error bars). Virtual particle concentrations 418 were calculated without taking into account shore steepness (left panels, Scenario NSS) and 419 with taking it into account (right panels, Scenario SS). R²=0.79 (upper left panel) and 0.84 420 (lower left) for Scenario NSS (both p-values >0.05); 0.96 (upper right panel) and 0.97 (lower 421 right) for Scenario SS (both p-values < 0.05). 422



Figure 17: Surface sinking and net beaching rates, calculated as in Fig. 4, for each of the 16 scenarios separately. The K_h and T_W values used are shown in the title of each subplot.



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432 Figure 18: Blue columns: seasonal deposition of plastic litter on Montalto beaches (west coast

of central Italy), measured between spring 2014 and winter 2015 (9). Orange columns: simu-

⁴³⁴ lated seasonal net beaching rate on the same beaches. Both quantities were normalized so that

the sum over the four seasons was equal to 1.



Figure 19: Surface sinking and net beaching rates, calculated as in Fig. 4, but using different proportion of particles released from cities, rivers, and vessels (p_C , p_R , and p_V ; upper table in Supplementary Fig. 5).

444 Supplementary Table 1

ence	ltems/km2	Kg/km2	Items/(kg of dry sediment)	Sampling year	Sampling location	Sampling depth (m)	Mesh size	Notes
	92	/	/	1994-1997	Gulf of Lion		> 1 cm	
t al., 2000	105	1	/	1998	East Corsica	50-1000	> 2 cm	
	263	/	/	1998	Adriatic Sea		> 2 cm	
1000	199	/	/	Nov 1997	Patras Gulf	80-120	- 1 E	
et al., 1999	70	/	/	May 1998	Echinades Gulf	247-360	> 1.5 CII	
endris et	142	1	1	2000-2003	Patras and Lakonikos Gulfs	15-320	> 1.5 cm	Small gulfs, complex topography
	1211	/	/		Saronikos Gulf			
lis et al.,	611	-	/		Patras Gulf	60 460		Small gulfs,
	416		/	Jall-Mar∠UIS	Echinades Gulf	064-06	> 3 CIII	topography
	24	/	/		South of Cyprus			
t al., 2015	/	/	2±2	Unspecified	Malta Grand Harbour	4-22	less than 5mm	Sampling inside the port
l., 2017	11715	/	/	May-Jun 2014 and Jul 2015	North of Po delta	21-23	Unspecified (video assessment)	
lu et al.,	2670	86.3	/	Unspecified	Mersin Bay, Cilician basin	Unspecified	Unspecified	
et al., 2018	10-470	137.25	/	Oct 2014-Feb 2015	Antalya bay	10-300	> 2.4 cm	
ivera et al.,	/	0-11	/	Feb-Aug 2014	0 to 35 km from Alicante	50-700	> 4/5 cm	
' et al.,	/	4.5-52.4	/	2016-2017	West of Barcelona	5-70	> 4 cm	
ət al., 2018	803	/	/	Jun-Jul 2012	South of Sicily	5-30	Unspecified (video assessment)	
al., 2018	35 ± 4	7.35 ± 2.4	/	2013-2015	Sardinia	0-800	>2 cm	
al., 2019 -	/	25 ± 60	/	2012-2015	Morocco	10-600	Unspecified	
k et al.,	30319	/	/	1700	-	< 3	Unspecified	Observ. very close
	6745	~	/	6102-2102	Israel	0-5	(observ. through scuba-dives	to the shore

concentration.
plastic
seafloor
Mediterranean
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Table

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