

Towards underwater plastic monitoring using echo sounding

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This manuscript is a non-peer reviewed preprint and has been submitted for publication at Frontiers in Earth Science. The DOI of the peer reviewed publication will be provided if accepted.

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14 **Keywords: Macroplastic, hydrology, sonar, marine litter, microplastic**

15 **Abstract**

16 Plastics originating from land are mainly transported to the oceans by rivers. The total plastic transport
17 from land to seas remains uncertain because of difficulties in measuring and the lack of standard
18 observation techniques. A large focus in observations is on plastics floating on the water surface.
19 However, an increasing number of observations suggest that large quantities of plastics are transported
20 in suspension, below the water surface. Available underwater plastic monitoring methods use nets or
21 fish traps that need to be deployed below the surface and are labour-intensive. In this research, we
22 explore the use of echo sounding as an innovative low-cost method to quantify and identify suspended
23 macroplastics.

24 Experiments under controlled and natural conditions using a low-cost off-the-shelf echo sounding
25 device show that plastic items can be detected and identified up to 7 m below the river surface. Eight
26 different debris items (metal can, cup, bottles, food wrappers, food container) were characterized based
27 on their reflection signature. Reflectance from plastic items diverged significantly from organic
28 material and non-plastic anthropogenic debris. During a multi-day trial field expedition in the
29 Guadalete river, Spain, half of the observed plastics items were found below the surface. As most
30 plastic monitoring and removal strategies focus on the upper layer, a substantial share of the total
31 plastic transport may be neglected. With this paper we (1) demonstrate that echo sounding is a
32 promising tool for underwater plastic monitoring, and (2) emphasize the importance of an improved
33 understanding of the existing plastic loads below the surface.

34

35 1 Introduction

36 Plastic pollution in aquatic ecosystems is of increased global concern due to its negative impact on
37 ecosystem health and human livelihood (Cózar et al., 2014; Lau et al., 2020; van Emmerik & Schwarz,
38 2020). Much of the plastic daily discarded on land is leaked into rivers, and transported into the world's
39 oceans (Schmidt et al., 2017; van Emmerik & Schwarz, 2020). However, estimates of plastic emissions
40 from rivers into the oceans are associated with great uncertainties due to methodological difficulties to
41 accurately quantify land-based plastic fluxes into the aquatic environment. To improve the
42 understanding of plastic transport dynamics from source to sink, reliable observations are crucial.

43
44 Plastics are abundant in all components of river systems: floating at the surface, accumulated on
45 riverbanks and floodplains, deposited in the sediment, and suspended in the water column (Schwarz et
46 al., 2019; van Emmerik & Schwarz, 2020). Currently available measurement methods primarily focus
47 on floating plastics (González-Fernández & Hanke, 2017; van Emmerik et al., 2018) or plastic on
48 riverbanks (Vriend et al., 2020), partially because measurements of plastics below the surface are more
49 difficult due to practical constraints. Previous efforts to quantify subsurface plastics depended on
50 heavy-duty cranes or ships to deploy subsurface nets (Liedermann et al., 2018; Morrith et al., 2014;
51 Schöneich-Argent et al., 2020), which often comes with high labour intensity and equipment costs.
52 Observations of subsurface plastics cannot be neglected, as recent work shows that underwater plastics
53 make up the largest portion of the plastic mass balance in the Atlantic Ocean (Pabortsava & Lampitt,
54 2020). To date, it is unknown to what extent this also holds for river systems. As rivers are assumed to
55 be one of the main plastic input sources into the oceans, there is a need to monitor the suspended
56 plastics in rivers.

57
58 To overcome the challenges with current underwater monitoring methods, we explore the use of sonar
59 technology as a potential solution. Sonar (Sound Navigation Ranging), or echo sounding, is based on
60 transmitting soundwaves into the water, which reflects on objects like fish, vegetation and bed. The
61 return time and the strength of the returning signal indicate object distance from the transducer and
62 material robustness, respectively. Sonar is currently used for purposes such as fish detection and seabed
63 mapping. Recent research tested the use of sonar for detecting litter objects in marine environments
64 (Valdenegro-Toro, 2019). In their research, they proposed the use of Deep Neural Networks to survey
65 and detect marine debris in the bottom of water bodies from forward-looking sonar images. A set of
66 objects was placed at the bottom of a small water tank and forward-looking sonar images were
67 generated using an ARIS Explorer 3000 sensor. Investigating the reflections of specific items and
68 opportunities to detect plastic items in more dynamic water bodies, such as rivers, has not been done
69 to date.

70
71 The main goal of this research is to explore the potential of echo sounding for riverine macroplastic
72 (>0.5 cm) monitoring below the water surface using an off-the-shelf low-cost sensor. We systematically
73 investigated the use of sonar for plastic monitoring through (1) indoor controlled tests, (2) semi-
74 controlled outdoor tests, and (3) uncontrolled application under natural conditions. The controlled tests,
75 to get an insight into the scanning technique and detection abilities of the echo sounder, were performed
76 in a swimming pool. During these tests, several influencing factors on the sonar signal were examined.
77 The semi-controlled tests were carried out in the Rio de San Pedro, Spain. This test aimed to investigate
78 the plastic detection of sonar for different plastic items. Lastly, the sonar was applied for macroplastic
79 monitoring under natural conditions in the Guadalete river, Spain. In this paper, we demonstrate that
80 (1) plastics can be detected below the surface using sonar, (2) specific macroplastic items have unique
81 reflections, and (3) results from the Guadalete river suggest plastic items below the surface accounts
82 for a substantial share of the total transport.

83 2 Methods

84 2.1 Principles of echo sounding

85 Sonar technology is based on the transmission of a sound signal and receiving the reflection. The
86 transmitted sound waves travel from the sonar transducer in the shape of a cone with increasing
87 footprint. The beam angle of the cone depends on the frequency with which the signal is emitted and
88 the transducer technology. In general, the higher the frequency, the smaller the cone angle. For this
89 research, a single beam sonar with Compressed High Intensity Radiated Pulse (CHIRP) technology is
90 used. The CHIRP technology differs from traditional sonars in the way frequencies are emitted. A
91 CHIRP sonar emits a continuous flow of a range of frequencies, while a traditional sonar sends out a
92 single frequency pulse at a time. By emitting pulses with different frequencies, ranging from low to
93 high, clearer sonar readings of higher resolution can be obtained, which enables improved target
94 separation compared to traditional sonars (Christ & Wernli, 2014).

95
96 Many echo sounding appliances translate sonar scans into sonar backscatter images. The displayed
97 signals obtained with, for instance, fish finders are a result of a 2D horizontal scan over the depth.
98 Since the 2D spherical plane is transformed to one point on the backscatter image, no indication of
99 where the fish is present in the scanned horizontal plane can be obtained. Emitting a burst of pluses
100 results in a vertical profile of single points at a certain moment in time. When displaying continuously,
101 the horizontal axis on the backscatter imagery indicates time, the depth below the sensor is presented
102 on the vertical axis. In this way, information about the position of the fish over the depth of the water
103 column can be collected.

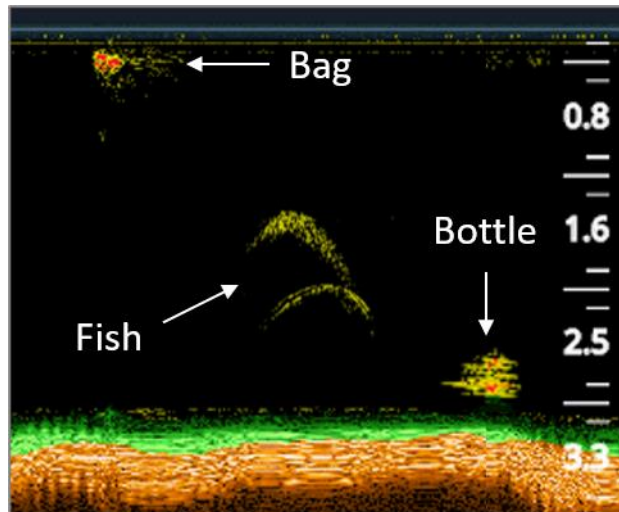
105 2.2 Sensor

106 The experiments performed were executed using the Deeper Smart Sonar CHIRP+ (Deeper CHIRP+),
107 which is a low-cost commercial fish finder. The sensor is a floating, GPS and Wi-Fi enabled fish finder,
108 using CHIRP technology. It has a diameter of 6.35 cm and a weight of 90 grams. The Deeper CHIRP+
109 enables scanning aquatic areas with three different beam widths (7, 16, 47 degrees) with corresponding
110 frequency domains (675, 290, 100 kHz), allowing for accurate target determination and separation (up
111 to 1 cm).

112
113 The Deeper CHIRP+ operates with the Deeper Smart Sonar mobile application, which can be installed
114 on a phone or tablet. In the app, the different settings, such as the scanning beam width and sensitivity
115 can be selected. Besides the sonar readings, information about the water depth and temperature are
116 provided in the app. The sonar scan data can be saved and uploaded to Lakebook, an online platform
117 where data of the scanning activities can be stored and viewed. From Lakebook, only raw bathymetry
118 data can be exported as CSV format. Exporting raw data on signal strength and intensity is not possible.
119 This sensor was chosen because of the ratio between scanning resolution/target separation and price.
120 Besides, the ability to save and store scanning data was advantageous. The downside of this sensor is
121 the limitation of raw sonar data export.

122
123 Since raw sonar data could not be exported, screenshots of the sonar signal reflections were taken and
124 processed using MATLAB. The obtained screenshots were segmented, using K-Means clustering, to
125 exclude the background pixels (Shan, 2018). Binary images were obtained from which the dimensions
126 of the sonar signal reflection could be calculated in pixels. The 'width' of the sonar reflection in the
127 backscatter imagery depicts the time the object is underneath the transducer, and is influenced by

128 velocity of the flow (and object) with respect to the sonar transducer. To correct for this, the width and
129 depth dimensions of the sonar reflection were calculated separately. The signal width was scaled for
130 the flow velocity measured by recording the time of movement over a known distance. An example
131 sonar recording including a plastic bag, bottle and fish is shown in Figure 1.
132

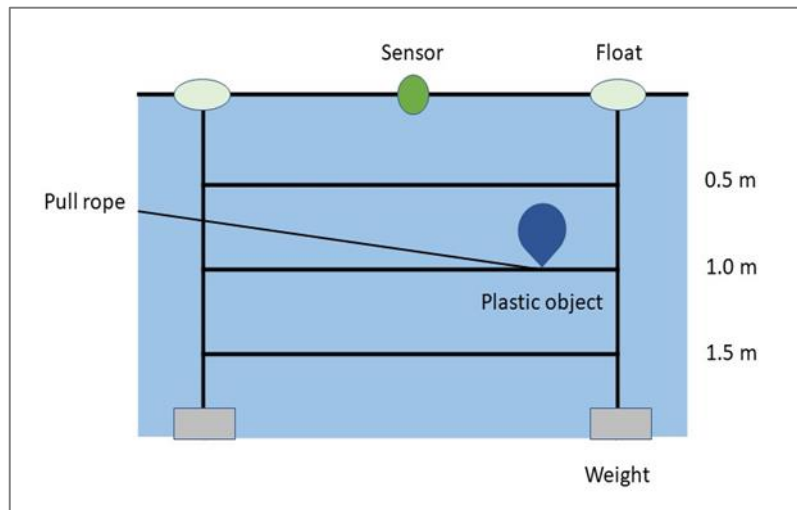


133
134 **Figure 1:** Sonar image example including a plastic bag, bottle and fish. Obtained using the Deeper
135 CHIRP+ fish finder. The numbers on the vertical axis present the depth below the water surface [m].

136 2.3 Controlled tests in the pool

137 The controlled tests aimed to investigate influencing factors on sonar reflection, such as the orientation
138 of objects, flow velocity and object depth. We conducted three experiments to isolate the effects of (1)
139 object size, (2) object depth, (3) flow velocity. Additionally, we tested the influence of object
140 orientation on the sonar signal reflection.

141
142 The controlled tests were done in the Kerkpolder swimming pool in Delft (51°59'25.9"N 4°19'53.3"E).
143 A framework of ropes was constructed, allowing passing items underneath the sensor at different
144 depths, velocity, and orientation, see Figure 2. We minimized the influence of object orientation during
145 the first experiments by using spherical balloons filled with water as test objects. The reflected signal
146 was therefore mainly influenced by actual object size, depth and flow velocity.



147

148 **Figure 2:** The experimental set-up used during the controlled tests in the swimming pool.

149 To investigate if a larger object returns a larger sonar signal reflection, a small (8 cm diameter) and
 150 large balloon (15 cm diameter), filled with water (same as ambient water), were passed underneath the
 151 sensor for fixed depth (0.5 m) and speed (0.15 m/s). The flow velocity was defined by recording the
 152 time of movement (pulling the objects with a rope) over a fixed distance. Secondly, to see if the depth
 153 of an object in the water column does influence the sonar signal return, the balloon of 15 cm was passed
 154 by the sensor at a depth of 0.5 and 1 m below the water surface, at a fixed velocity of 0.15 m/s. Thirdly,
 155 the influence of flow velocity on the sonar signal reflection was examined by pulling the 15 cm balloon
 156 underneath the sensor for fixed depth (0.5 m) at two different flow velocities, 0.15 and 0.25 m/s,
 157 respectively. These different experiments were repeated ten times. We tested the influence of object
 158 orientation in a separate experiment. For this, we used a filled 1.5 L plastic water bottle. The bottle was
 159 fixed to a depth of 1 m and held horizontally orientated for a duration of 30 seconds. This was thereafter
 160 repeated for the bottle being vertical orientated.

161

162 The used echo sounder has several options for beam width. We used a beam angle (total angle) of 7
 163 degrees, which provides the highest scanning resolution (target separation of 1 cm) and lowest spatial
 164 resolution (smallest scanning area). These beam settings result in a blind zone of 15 cm at the water
 165 surface, for which the sensor is not able to detect objects due to surface clutter. In the end, the
 166 significance of the results was determined using an independent t-test with 0.05 as significance level.

167

168 2.4 Semi-controlled tests in the Rio de San Pedro

169 Semi-controlled test were carried out in in the Rio de San Pedro, a tidal river close to the city of Puerto
 170 Real (36°31'53.9"N 6°12'56.5"W). The goal was to obtain data on plastic detection with sonar for
 171 different plastic items. The sensor was deployed in the Rio de San Pedro (Figure 3 (1)) to collect
 172 reflection signals for specific plastic objects, and test the performance under natural river conditions.
 173 The experiment was conducted by releasing a set of objects, attached to thin fishing lines, repeatedly
 174 into the river, passing the scanning beam of the sensor between 0.5 and 2.5 m below the surface. As
 175 the objects were released into the river, they passed the sensor driven by the river flow velocity, as
 176 illustrated in Figure 4 (1). This was repeated ten times per item. To obtain a robust dataset, and apply
 177 the sensor for varying conditions (turbidity and salinity), this experiment was repeated on five days (3,
 178 10, 14, 25 and 29 October, 2019). The set of items used for this experiment was based on the most
 179 abundant plastic items in river systems according to literature (González Fernández et al., 2018; van

180 Emmerik et al., 2020). To obtain a broad overview of the detection abilities of sonar, items of different
 181 dimensions and material properties were used. The set included a cup, bag, can, small plastic bottle,
 182 large plastic bottle, small food packaging item, large food packaging item, and a food container (photos
 183 of the objects and the object dimensions are presented in the supplementary materials). During these
 184 tests, the scanning beam width of the sensor was set to the narrow beam of 7 degrees. To analyse if the
 185 sonar signal footprint was significantly different for the tested items, an independent t-test with a
 186 significance level of 0.05 was used.
 187



188
 189 **Figure 3:** The measuring locations used during the fieldtrip in Andalusia, Southern Spain. Location
 190 1, Rio de San Pedro, used for the semi-controlled tests with plastic targets. Location 2, Rio de
 191 Guadalete, were multiday monitoring is performed for varying tide

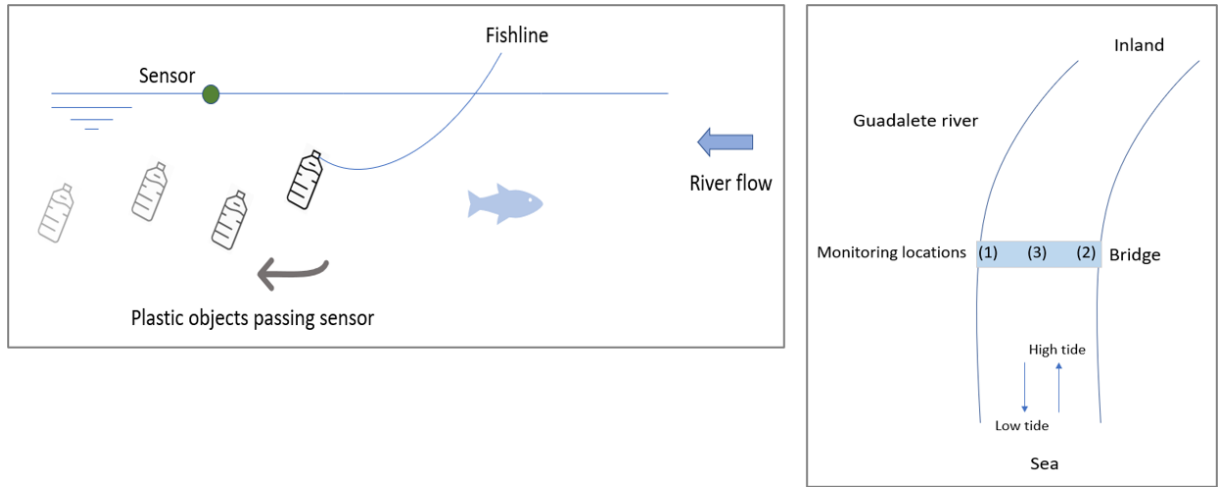
192 **2.5 Field tests in the Rio de Guadalete**

193 The objective of the third experimental campaign was to apply the sensor for monitoring macrolitter
 194 in a natural river system. To test the sensor in a natural river system, the sensor was operated during
 195 18 hours of monitoring in the Guadalete river in El Puerto de Santa Maria (36°35'58.6"N 6°13'17.5"W)
 196 (Spain). The sensor was deployed from a pedestrian bridge (of 100 m wide) over the river. The river
 197 monitoring took place on 8, 11, 17, 22, 23, 24, 26 and 28 October 2019 for varying tidal conditions.
 198 Monitoring was done for one hour per testing day and tidal condition. Additionally, to investigate the
 199 cross-sectional litter distribution, we monitored at three locations across the river width. The river flow
 200 at the measurement location was bidirectional because of tidal influence in the Gulf of Cádiz (Atlantic
 201 Ocean). We therefore investigated the difference in vertical and cross-sectional litter distribution for
 202 ingoing and outgoing tide. The monitoring location and setup is shown in Figure 3 (2) and Figure 4
 203 (2).
 204

205 Plastic litter objects were identified based on the backscatter images obtained during the semi-
 206 controlled tests, using both the signal signature as the signal indicated strength (colour). Fish were
 207 discarded from the sonar readings by their specific arc-shaped reflection. To correct for the shape of
 208 the angled scanning beam (cone), the monitored items over the river depth were scaled to 1 m river
 209 width. The depth was divided into four zones. For each zone, the total number of items per hour is

210 presented. Besides, a division is made between the two tidal flow conditions (incoming tide and
 211 outgoing tide).

212
 213 The sensor was deployed using the wide beam (47 degrees) which enables scanning with the highest
 214 spatial resolution (largest scanning area) but the lowest scanning resolution (least detailed scanning).
 215 These beam settings result in a blind zone of 80 cm depth for which objects cannot be detected by the
 216 sensor. The significance of the results is determined using an independent t-test with 0.05 as
 217 significance level.
 218



219
 220 **Figure 4:** Experimental set-ups for the semi-and uncontrolled tests. Set-up 1, applied in the Rio de
 221 San Pedro, passing plastic items underneath the sensor. Set-up 2, the three monitoring locations over
 222 the cross-section of the Rio de Guadalete.

223
 224 **3 Results and discussion**

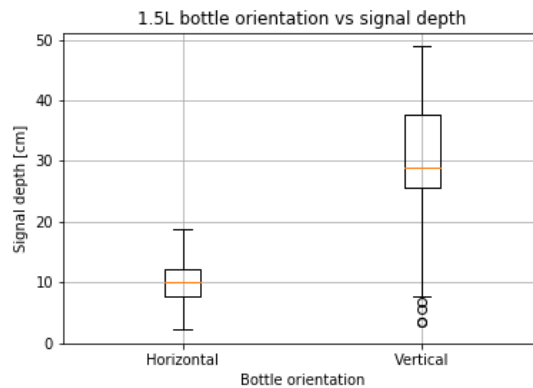
225 **3.1 Controlled tests in the pool**

226 A significant relation was found between the sonar signal reflection and the actual item size. A larger
 227 item (15 cm balloon) resulted in a larger displayed sonar signal compared to a smaller item (8 cm
 228 balloon). No significant relation was observed between the depth at which an item is present in the
 229 water column and the sonar signal reflection. A significant relation was found between the flow
 230 velocity and the signal reflection. For items passing with a larger flow velocity (0.25 m/s), the signal
 231 reflection was significantly smaller compared to the signal reflection for a lower flow velocity (0.15
 232 m/s).

233
 234 Based on these results, we identified some potential sources of uncertainty. We found several outliers
 235 in the observations, that may be explained by the method for pulling the items through the water. These
 236 outliers can be caused by pulling the objects with a rope instead of letting them naturally flow in the
 237 water when passing the sensor. Pulling could induce water displacement in front of the objects and
 238 possible disturbance in the sonar signal. Moreover, the filled balloons were not as spherical as
 239 envisioned and deformed while pulling them through the water. This deformation (changing object
 240 dimensions) could lead to a spread in the observed sonar reflections.
 241

242 Moreover, there was no clear influence of object depth evident from the sonar signal. However, only
243 two different depths (0.5 m and 1 m) were examined. Possibly the influence of depth can be present
244 when testing for a larger range in depth. Lastly, tests were performed at two different velocities (0.15
245 and 0.25 m/s). It was found that the velocity with which items pass the sensor does influence the sonar
246 signal reflection. A higher flow velocity results in a smaller reflection, compared to a lower flow
247 velocity. It is, however, not tested to what extent objects can still be identified with increasing flow
248 velocity.

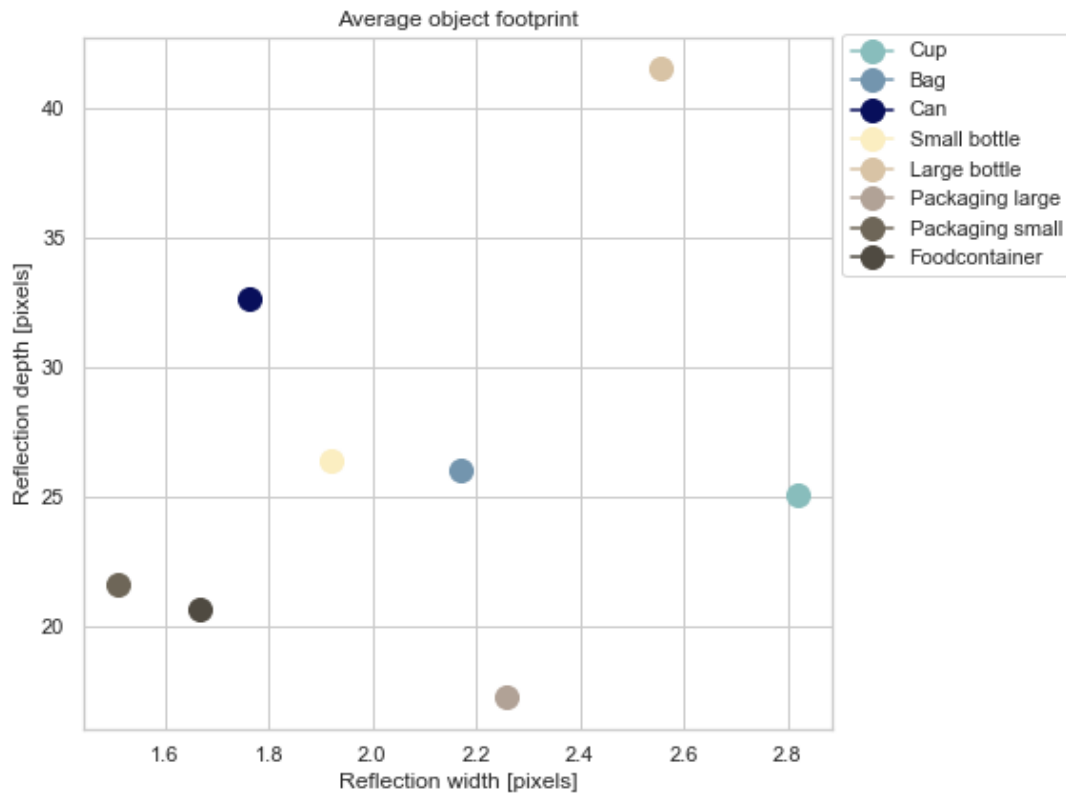
249
250 The results obtained from the bottle orientation test are displayed in Figure 5. The signal reflection
251 differed significantly for the horizontal and vertical orientated bottle. The sonar reflection for the
252 horizontal and vertical placement of the bottle was 10 cm and 28 cm, respectively. Compared to the
253 actual dimensions of the water bottle, which is 8.5 cm diameter and 27 cm height, the depth of the
254 sonar signal reflection corresponds approximately to the order of magnitude of the actual dimensions
255 of the bottle.



256
257 **Figure 5:** Results obtained during object (bottle) orientation experiment of the controlled tests in the
258 swimming pool.

259 3.2 Semi-controlled tests in the Rio de San Pedro

260 From the semi-controlled experiments, in the Rio de San Pedro, we found that the average reflection
261 footprints of specific items varied substantially (Figure 6). It seems the detected items can be
262 characterized by specific sonar reflections. When looking at the actual item size and the reflection
263 footprint, one would expect, according to the results in section 3.1 that a larger item results in a larger
264 sonar reflection footprint. This is however not the case for all items tested.



265

266 **Figure 6:** The average sonar reflection footprint (depth and width) of the different items tested.

267 Besides, a variation in the data is observed, Figure 7. The reflection depth, width and area data for the
 268 different items are not consistent but spread. When comparing the reflectance depth, width and area of
 269 the different items, Table 1, we see that at least one dimension is significantly different for 18 out of
 270 the 28 combinations. This supports the potential for litter qualification using sonar.

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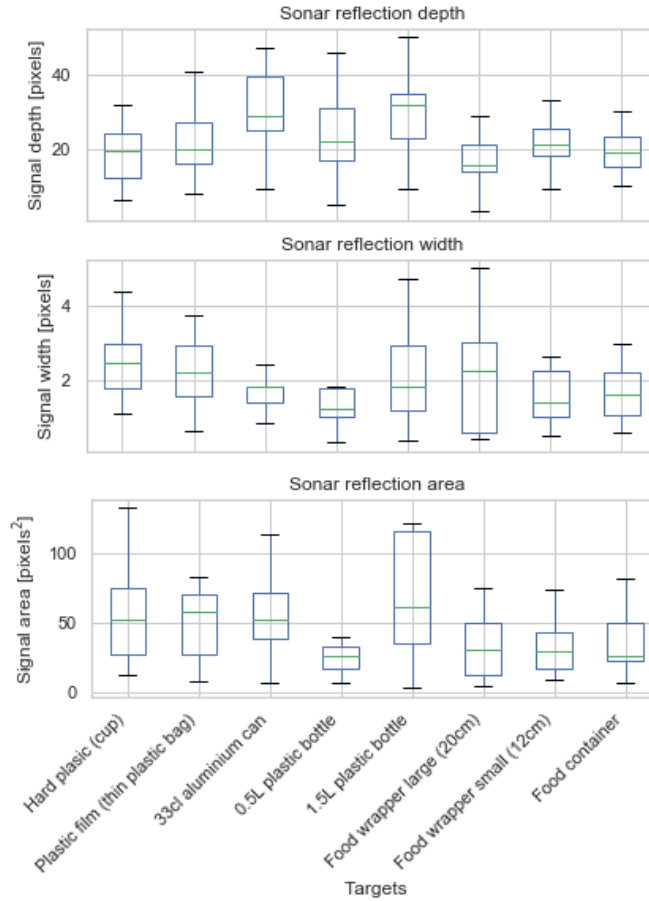


Figure 7: The total data of the sonar reflection depth, width and area, for the different items (targets).

271

272 **Table 1:** The significance in depth (d), width (w), and area (a) of the sonar footprints for the tested
 273 items compared to each other. When there is a significant difference, the depth, width or area (d-w-a)
 274 is indicated.

	Cup	Bag	Can	Bottle S	Bottle L	Food wrapper L	Food wrapper S	Food container
Cup	X							
Bag	-	X						
Can	w	-	X					
Bottle S	-	a	a	X				
Bottle L	-	-	-	a	X			
Food wrapper L	a	d-a	d-a	-	d-a	X		
Food wrapper S	w	w	d-a	-	d-w-a	d	X	
Food container	w-a	a	d-a	-	d-a	-	a	X

275

276

277 Possible reasons for the inconsistency (spreading and no direct link with the actual item size) in the
 278 data is the influence of the orientation and deformation of the objects. For example, a water bottle, as

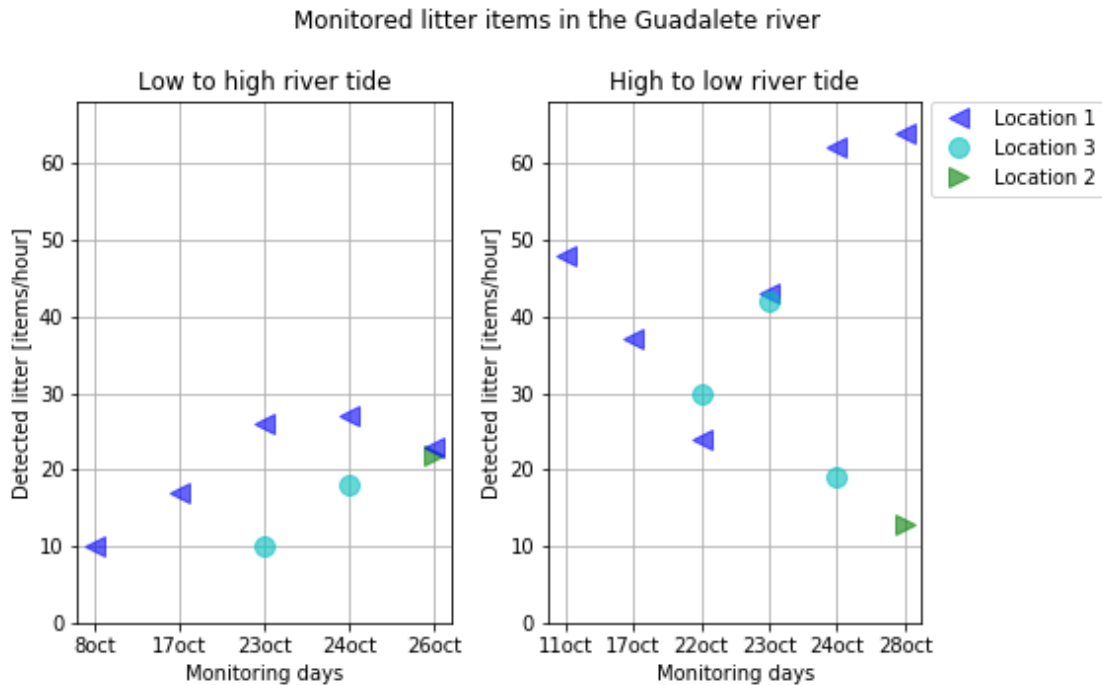
279 shown in Figure 5, can result in a very different footprint when orientated differently. Moreover, items
 280 such as plastic bags and packaging are likely to deform, which can lead to potentially very different
 281 sonar reflections. This makes the identification of items according to their sonar footprint complex.

282
 283 Besides the dimensions of the sonar signal reflection, the sonar signal intensities are also examined.
 284 The metal can corresponds to the highest signal intensity and the food wrapper to the lowest signal
 285 intensity. When comparing this to the material properties of the items it can be recognised that for some
 286 objects the measurements fit the expectations (higher material density results in higher sonar signal
 287 intensity). However, no direct link between the sonar signal intensity and the material properties of the
 288 total of tested objects was observed in this study. The potential of classifying items based on their
 289 material properties and sonar reflections seems although interesting to investigate further, using for
 290 example Artificial Intelligence.

291 **3.3 Field application in the Rio de Guadalete**

292 Lastly, the sensor was applied during a multiday trial monitoring campaign in the Rio de Guadalete.
 293 The number of monitored items per hour are shown in Figure 8. In total, the river was monitored for
 294 18 hours over eight different days and varying location over the cross-section of the river. The results
 295 showed that significantly more items are transported during river ebb tide (water flows from inland to
 296 the sea), compared to the river flood tide (water flows from the sea inland).

297



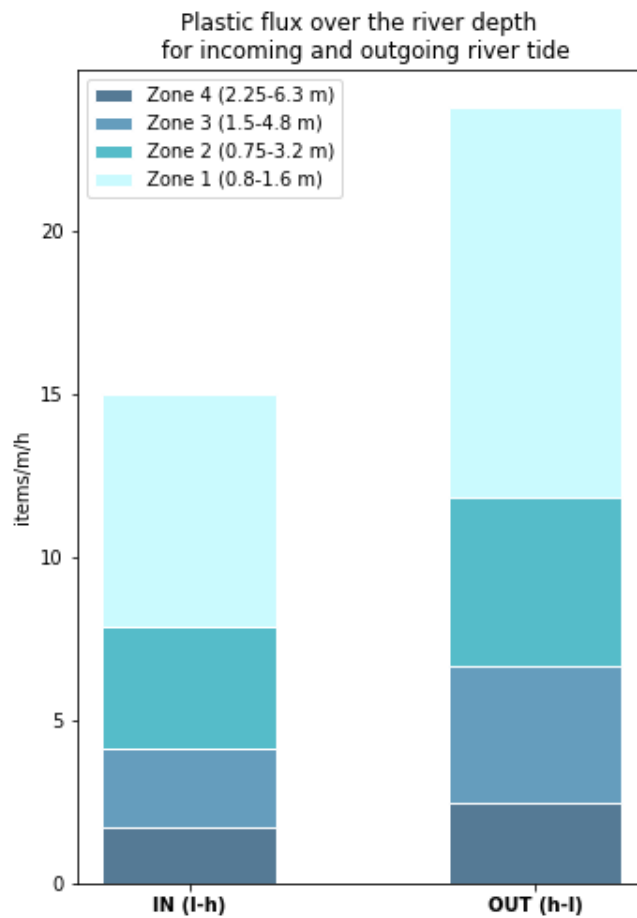
298

299 **Figure 8:** Total monitored items during the field campaign using the wide scanning beam (47 degrees),
 300 for the three different locations over the river’s cross-section. Left: monitored items for river water
 301 level going from low to high (water flows from the sea inland). Right: monitored items for river water
 302 level going from high to low (water flows from inland towards sea).

303 On average, during ebb tide (high to low river tide), 38 items/hour were detected by the sensor. For
 304 flood tide (low to high river tide), 19 items/hour were detected. Furthermore, we found a difference in
 305 litter items over the river cross-section. It appears that more litter is transported at location 1 compared

306 to locations 2 and 3. In order to find an explanation, the river’s cross-section was mapped using the
 307 sensor, showing that the river bottom is not uniformly shaped over the width of the river. We observed
 308 erosion on the outer bend, which coincides with the monitored litter transport peak. Generally, flow
 309 velocities are higher in the outer bend and potentially more items could pass the sensor compared to
 310 the inner bend.

311
 312 Besides counting litter items, the depth at which the litter particles were present is indicated, leading
 313 to the particle distribution as illustrated in Figure 9. For each zone, the total number of items per hour
 314 is presented. No clear difference is observed for the two tidal flow conditions (IN-OUT). According to
 315 the results presented in Figure 9, most litter items are present in Zone 1. An important remark is that
 316 due to surface clutter a blind zone, for which the sensor is not able to detect objects, of 80 cm was
 317 present at the water surface. In other words, items present in the top 80 cm of the water column are not
 318 taken into account. Based on our findings, 50 percent of the monitored litter is present in deeper layers
 319 (Zone 2, 3, and 4) of the water column.



320
 321 **Figure 9:** Monitored litter (items/hour/m river width) distribution over the river depth (divided in 4
 322 zones) for incoming and outgoing river tide.

323 Note that the counted litter items were identified as plastics according to the footprint data obtained
 324 during the semi-controlled test. However, the dataset collected during the semi-controlled experiments
 325 does not cover the total range of possible litter items. Therefore there is the possibility that other litter
 326 items are wrongly identified as plastics, leading to a higher plastic load than actually present. To ensure
 327 litter items are correctly identified as plastics, more research is needed to determine footprints of
 328 different types of items such as other anthropogenic debris and organic litter. Fish resulted in a very

329 distinct signal reflection, illustrated in Figure 1, and are accordingly assumed to be filtered correctly
330 from the data.
331

332 **3.4 Synthesis**

333 **Using echo sounding to detect plastic**

334 Our findings show that echo sounding has potential for monitoring subsurface macroplastics. Plastic
335 items can be detected and possibly be classified based on their size and material properties. Being able
336 to monitor suspended plastics in rivers takes us a step closer to estimate global plastics transport rates.
337

338 The dimensions of objects in the sonar reflection imagery are related to the actual size of the passing
339 object (a larger item results in a larger reflection). However, sonar reflections are found to be sensitive
340 to object orientation and deformation. Another factor that influences the sonar reflection is flow
341 velocity. Items passing with high velocity are displayed significantly smaller than items passing with
342 low velocities. The flow velocity upper limit for the detection of objects using echo sounding was not
343 considered in this study (it was tested up to 0.25 m/s). Depending on the actual object size, flow velocity
344 could probably be a limiting factor for plastic detection using echo sounding.
345

346 For a widespread application of the echo sounding technique in riverine plastic monitoring some
347 challenges remain. More fundamental testing is needed to discard other litter types (vegetation etc.)
348 from the sensor readings, to be certain on monitoring only anthropogenic litter and plastics.
349 Furthermore, the classification of the different plastic litter objects would be beneficial for source
350 identification and targeted cleaning strategies. We did not find a direct link between object size,
351 material properties and reflected signal. However, our results showed that the potential is there. Very
352 specific and consistent testing of objects ranging in either size or material property could contribute to
353 more robust monitoring using echo sounding.
354

355 **The Deeper CHIRP+ and potential of other sensors**

356 For this research we used the Deeper CHIRP+ fish finder. We chose this sensor because of its
357 accessible price, size, and user-friendliness. For a proof-of-concept this sensor suited his purpose well.
358 The main disadvantage of this sensor is the limitation in raw data export. No raw sonar data could be
359 exported, therefore screenshots of the sonar signals were processed. In general, the accuracy of the
360 results could be affected due to sonar image processing, instead of using raw sonar data.
361

362 The sensor was deployed using its different scanning beam settings. For the different settings, blind
363 zones occur near the water surface at which no objects can be detected. For the narrow and wide
364 scanning beam, a blind zone of 15 cm and 80 cm, respectively, is present. During the executed tests, it
365 was assured that the items passed the sensor below the blind zone. However, for the monitoring activity
366 in the Guadalete river, it needs to be considered that the collected data does not include the full river
367 depth, due to the blind zone at the water surface. For most echo sounding devices, blind zones or
368 blanking distances are present. This leads to limited employability in shallow waters and the use for
369 near-surface objects. The impact of this is however limited since most research efforts and cleaning
370 strategies focus, due to sampling difficulties, on (near) surface plastics (approximately up to 1.5 m
371 depth), and therefore the potential of monitoring with echo sounding devices beyond this 1.5 m proves
372 its complementarity.
373

374 Different, more advanced sensors, such as an ADCP or Multibeam echo sounder could potentially lead
375 to more detailed sonar readings and allowing particle size/properties indication. ADCPs are designed

376 for velocity measurements but are currently applied for various purposes. The study of Sassi et al.
377 (2012) shows the applicability of ADCP for monitoring suspended particulate matter in rivers and
378 marine environments. Additionally, using horizontally mounted ADCPs at riverbanks, which enables
379 monitoring during high discharges (Hoitink et al., 2009), indicates also the potential for litter
380 monitoring in rivers. However, the costs of these devices are large (>20.000 Euro) compared to
381 conventional fish finders, which makes them less broadly applicable.

382

383 **Monitoring in natural rivers**

384 When applying the obtained knowledge from the controlled and semi-controlled tests to the field, the
385 following aspects should be considered when using echo sounding as a monitoring technique. As
386 previously stated, the actual litter size is hard to estimate from the sonar readings because of object
387 orientation, deformation, and flow velocity, implying an uncertainty when using the sensor for
388 monitoring purposes. In addition, obtained data on litter transport depends on the chosen beam width,
389 leading to the presence of a blind zone at the water surface.

390

391 From the monitoring data obtained in the Guadalete river, a distinct difference between fish and
392 anthropogenic litter could be observed. When comparing the sonar signal data to fish finding theory,
393 fish can be discarded from other objects by the specific shaped signal. However, this assumption is
394 only based on fish finding theories and has not been validated in practice.

395

396 **Plastics in suspension**

397 According to our results, 50 percent of the plastics are present below 1.6 m from the water surface
398 (measured from 0.8 m depth due to blind zone). This has a large impact on current monitoring projects,
399 which focus mostly on the plastics in the top layer. Taking into account the material properties of
400 (suspended) plastics, it is likely that litter items are present at different depths based on their density.
401 Moreover, turbulence, litter shape and vegetation may also influence the vertical location of the
402 particles.

403

404 The fact that, in the Guadalete river, 50 percent of the transported litter was present in deep layers of
405 the water column stresses the importance of monitoring subsurface plastics, as they likely account for
406 a large share of the total plastic transport. Recent work shows that underwater plastics make up the
407 largest portion of the plastic mass balance in the Atlantic Ocean (Pabortsava & Lampitt, 2020), this
408 might be the same in rivers. If we want to solve the plastic crisis, more effort is needed to develop
409 monitoring methodologies for underwater plastics. The river surface cannot be the carpet of the future
410 (everything beneath we don't see).

411

412 **4 Conclusions**

413 Echo sounding can be used for detecting suspended riverine macroplastics. Litter items can be counted,
414 while fish can be discarded from the specific signal reflections. Moreover, mean item reflection signals
415 yield unique combinations of width, depth and surface, which can potentially be used to identify
416 different litter types. Litter size was related to the sonar signature, although factors such as flow
417 velocity, object orientation and deformation need to be also considered when estimating size. This
418 remains challenging and further experiments are needed to collect more robust reflection statistics on
419 litter items. In the Guadalete river, significantly more suspended litter is transported when water flows
420 towards the sea compared to water flowing inland. Fifty percent of the counted litter items were present
421 in the deep layers (> 80 cm depth) of the water column.

422

423 Echo sounding is potentially useful to gain a better understanding of the suspended litter transport,
424 from which prevention and mitigation strategies could be optimised. For further research, it is
425 recommended to use an echo sounder for which the raw sonar data can be exported as a standard digital
426 file. Moreover, the set of test objects should be extended, including a wider range of sizes and shapes.
427 Objects of different size made of the same material and objects of the same size and different material
428 properties should be combined for testing. Side-scan or multibeam sonars might also lead to more
429 accurate characterization of litter sizes and materials.

430 **5 Conflict of Interest**

431 The authors declare that the research was conducted in the absence of any commercial or financial
432 relationships that could be construed as a potential conflict of interest.

433 **6 Author Contributions**

434 Conceptualization: SB, TvE, WL, Methodology: SB, TvE, WL, DGF, Formal Analysis: SB, Data
435 collection: SB, DGF, Visualization: SB, TvE, Writing – original draft: SB, TvE, Writing – reviewing
436 and editing: all authors.

437 **7 Funding**

438 SB was received funding from the Lamminga Fund and the department of Water Resources
439 Management (TU Delft).

440 **8 Acknowledgments**

441 This paper is based on the MSc thesis: The sound of plastic: A proof-of-concept for detecting
442 suspended riverine macroplastics with echo sounding (Broere, 2020). The data collection was done in
443 collaboration with the University of Cádiz, for which we are thankful.

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503 **10 Supplementary Material**

504 The Supplementary Material for this article can be found online at: (frontiers link)

505 **11 Data Availability Statement**

506 The datasets generated for this study can be found in the 4TU.ResearchData repository [LINK].