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# 1 A field guide for monitoring riverine macroplastic entrapment in water hyacinths

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20 **Keywords:** plastic, marine debris, hydrology, microplastics, vegetation, Vietnam

## 21 Abstract

22 River plastic pollution is an environmental challenge of growing concern. However, there are still many  
23 unknowns related to the principal drivers of river plastic transport. Floating aquatic vegetation, such as water  
24 hyacinths, have been found to aggregate and carry large amounts of plastic debris in tropical river systems.  
25 Monitoring the entrapment of plastics in hyacinths is therefore crucial to answer the relevant scientific and  
26 societal questions. Long-term monitoring efforts are yet to be designed and implemented at large scale and  
27 various field measuring techniques can be applied. Here, we present a field guide on available methods that  
28 can be upscaled in space and time, to characterize macroplastic entrapment within floating vegetation. Five  
29 measurement techniques commonly used in plastic and vegetation monitoring were applied to the Saigon river,  
30 Vietnam. These included physical sampling, UAV imagery, bridge imagery, visual counting, and satellite  
31 imagery. We compare these techniques based on their suitability to derive metrics of interest, their relevancy at  
32 different spatiotemporal scales and their benefits and drawbacks (SWOT analysis). This field guide can be used  
33 by practitioners and researchers to design future monitoring campaigns and to assess the suitability of each  
34 method to investigate specific aspects of macroplastic and floating vegetation interactions.

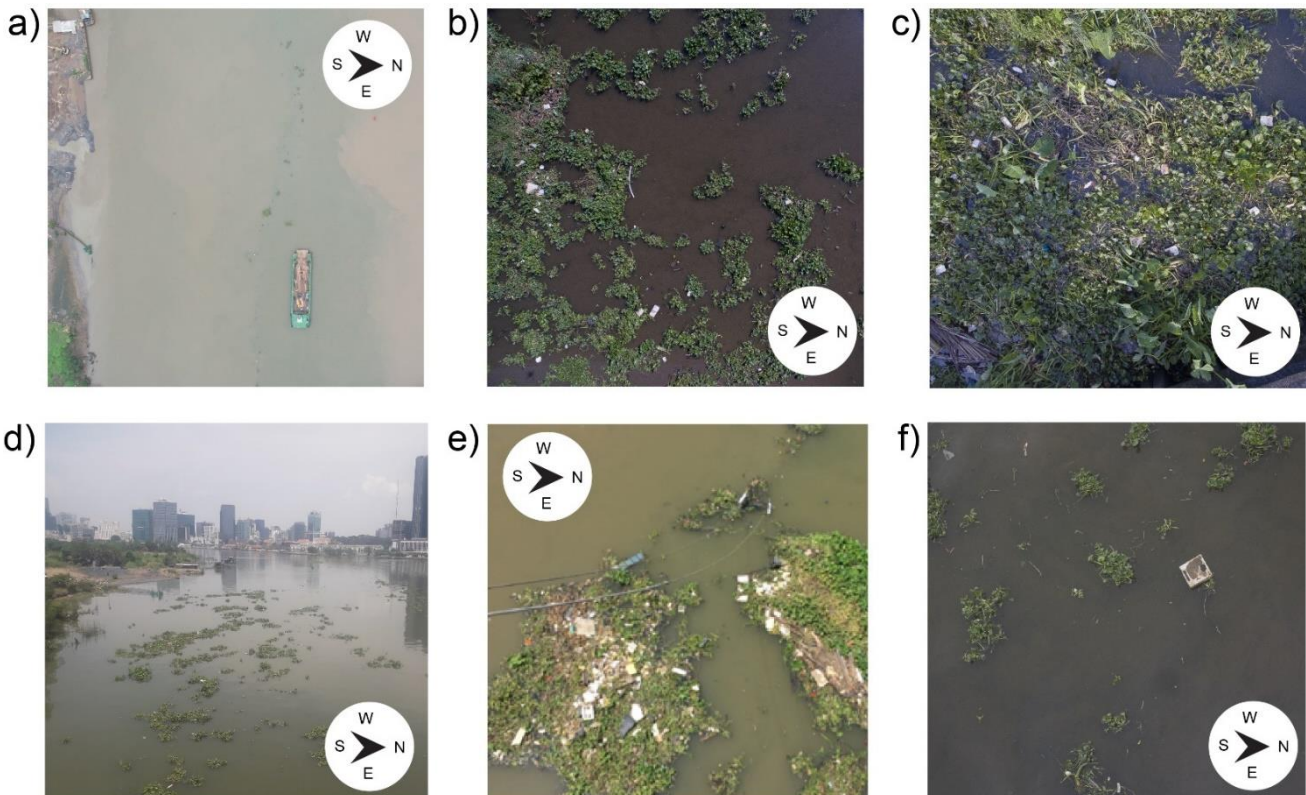
## 35 1. Introduction

36 Plastic pollution threatens terrestrial, freshwater and marine life, and causes significant economic losses.  
37 Quantifying the amount of plastic entering the marine environment is crucial for implementing and assessing  
38 plastic reduction strategies. Rivers are assumed to be a major pathway for land-based plastic pollution  
39 transported towards the ocean, and river plastic emissions are estimated between 0.8–2.7 million tons of plastic  
40 per year (Meijer et al., 2021). Estimating the quantities of plastic emitted into the oceans is a challenging task  
41 and so far model-based estimates have large uncertainties (Lebreton et al., 2017). Several studies pointed out  
42 the need to improve the parametrization and calibration of large-scale models with in-situ measurements  
43 (González-Fernández & Hanke, 2017; Meijer et al., 2021; van Emmerik et al., 2018). Additional field

44 measurement efforts could further improve our understanding of the drivers of riverine plastic transport and  
 45 accumulation processes.

46 Research on riverine plastic highlighted that the quantities and composition of transported plastics vary  
 47 geographically, temporally and along the water column (Liedermann et al., 2018; van Calcar & van Emmerik,  
 48 2019). However, the drivers of floating macroplastic transport in rivers remain unknown for most rivers in the  
 49 world. This prompts for further investigation of the processes governing macroplastic propagation through rivers.  
 50 In some rivers, hydrometeorological processes, such as river flow dynamics, change between seasons and  
 51 extreme events are important drivers of plastic transport fluctuations (Hurley et al., 2018; Kurniawan & Imron,  
 52 2019; van Emmerik, Strady, et al., 2019; van Emmerik, Tramoy, et al., 2019). Yet, hydrometeorological factors  
 53 are not solely responsible for riverine plastic transport (Roebroek et al., 2021). Studies focused on coastal  
 54 ecosystems suggest that vegetation may act as a sink for plastic litter, but focus on coastal ecosystems, once  
 55 the plastic has exited the river mouth (Brennan et al., 2018; Cozzolino et al., 2020; Ivar do Sul et al., 2014;  
 56 Martin et al., 2019; Olivelli et al., 2020). The role of vegetation – such as floating aquatic weeds and riparian  
 57 vegetation – along the river is understudied, despite recent evidence these accumulate and transport  
 58 macroplastic (Schreyers al., 2021; van Emmerik, Strady, et al., 2019). Riverine vegetation could act as static  
 59 accumulation sites and/or carriers of plastics into the ocean. Ultimately determining vegetation’s role requires  
 60 testing (and possibly adapting) the current measurement techniques used to monitor riverine macroplastic  
 61 accumulation and transport.

62 Recent findings indicate that, on average, close to 80% of all observed floating macroplastics in the Saigon river  
 63 accumulate in water hyacinth beds (Schreyers et al., 2021a). Hyacinths are a floating aquatic weed that tend to  
 64 form large patches (several meters of width and length) and in which important quantities of debris can get  
 65 entrained (Figure 1). Water hyacinth is an invasive species present in many tropical freshwater systems (Hailu  
 66 & Emanu Getu, 2018), and it is very likely that floating macroplastic accumulates at large rates within water  
 67 hyacinth patches in other rivers as well. For instance, the Chao Phraya river in Thailand has a high abundance  
 68 in both water hyacinth coverage and plastic pollution concentrations (Kleinschroth et al., 2021; Ta & Babel,  
 69 2020). In addition, other plant species present in rivers may entrap and/or transport plastic too. This calls for  
 70 further monitoring efforts on the role of water hyacinths as macroplastic carriers in rivers. Hence, it is necessary  
 71 to develop practical and consistent methods for monitoring plastic entrapment in vegetation, in order to further  
 72 study the role of water hyacinths in plastic transport dynamics.



73

74 **Figure 1.** Photographs taken at Ho Chi Minh City, Vietnam. The upper figures show the UAV images taken at  
 75 different altitudes. The lower figures show photographs taken from bridges, facing the river downstream, at  
 76 approximately 15 m of altitude. (a) UAV image taken at approximately 60 m, showing the southern cross-section  
 77 of the river. Patches of hyacinths are visible a few meters downstream (West) from the boat (b) Patches of  
 78 different sizes clearly visible from the UAV image at an altitude of approximately 10 m. (c) A patch of water  
 79 hyacinths visible from an altitude of less than 10 m. Plastic items are clearly visible. (d) Overview of the Saigon  
 80 river. Large water hyacinths patches are visible on the forefront. (e) Large water hyacinths patches, many  
 81 entangled plastic items are visible. (f) Individual water hyacinths, not aggregated in large patches and a free-  
 82 floating debris.

83 Riverine macroplastic monitoring is a rapidly evolving field, utilizing a variety of measurement techniques, from  
 84 low to high-tech and from in-situ to satellite imagery. Some of these methods have also recently been adapted  
 85 to monitor floating hyacinth patches and the entrapment of plastic debris in hyacinths. For instance, visual  
 86 counting and UAV imagery were previously used to characterize water hyacinths entrapment and transport over  
 87 the Saigon river (Schreyers et al., 2021). Physical sampling of plastic and hyacinths enabled to estimate the  
 88 composition of the debris found and the mass of both debris and hyacinths (van Emmerik, Strady, et al., 2019).  
 89 The detection of floating debris patches in coastal waters and the quantification of water hyacinths coverage  
 90 over entire river systems is made possible by optical satellite imagery (Biermann et al., 2020; Kleinschroth et  
 91 al., 2021). With growing understanding of the role of vegetation in plastic debris entrapment in rivers, various  
 92 methods can be mobilized at different spatial and temporal scales, often complementing each other. Despite  
 93 these initial monitoring efforts on macroplastic entrapment in hyacinths, long-term and large-scale monitoring  
 94 plans are yet to be designed. These would require a better understanding of the suitability and relevancy of each  
 95 monitoring technique, as well as their benefits and drawbacks.

96 Here, we present a field guide with measurement techniques to quantify and characterize macroplastic  
 97 entrapment in floating water hyacinths. This field guide can be used by practitioners, local authorities or scientists  
 98 for designing long-term monitoring campaigns on floating macroplastic in rivers. This is needed since water  
 99 hyacinths were found to be an important sink and carrier for macroplastic in the Saigon river, and most likely  
 100 play a similar key role in the transport and entrapment of debris in other tropical rivers. We adapted five  
 101 measuring methods suitable to monitor the entrapment of plastics within hyacinth patches (visual counting,  
 102 Unmanned Aerial Vehicle imagery, bridge imagery, physical sampling and satellite imagery). These methods  
 103 were tested nearly simultaneously during a campaign conducted at the Saigon river and we show which insights  
 104 they provide on quantifying macroplastic entrapment in hyacinths. We discuss the benefits, limitations and  
 105 practical considerations of these various measuring techniques, from low to high-tech and from in-situ to satellite  
 106 imagery. This is useful for designing a suitable monitoring strategy as the field guide provides an overview of  
 107 the possible options available. To determine what (combination of) measuring methods to choose, we present  
 108 (1) the main metrics that can be derived from each method (2) the relevancy of each method at different spatial  
 109 scales, and (3) the benefits and drawbacks of each technique, through a Strengths, Weaknesses, Opportunities,  
 110 and Threats (SWOT) analysis. The goal of this practical field guide is to help designing long-term monitoring  
 111 campaigns on plastic entrapment in floating aquatic vegetation.

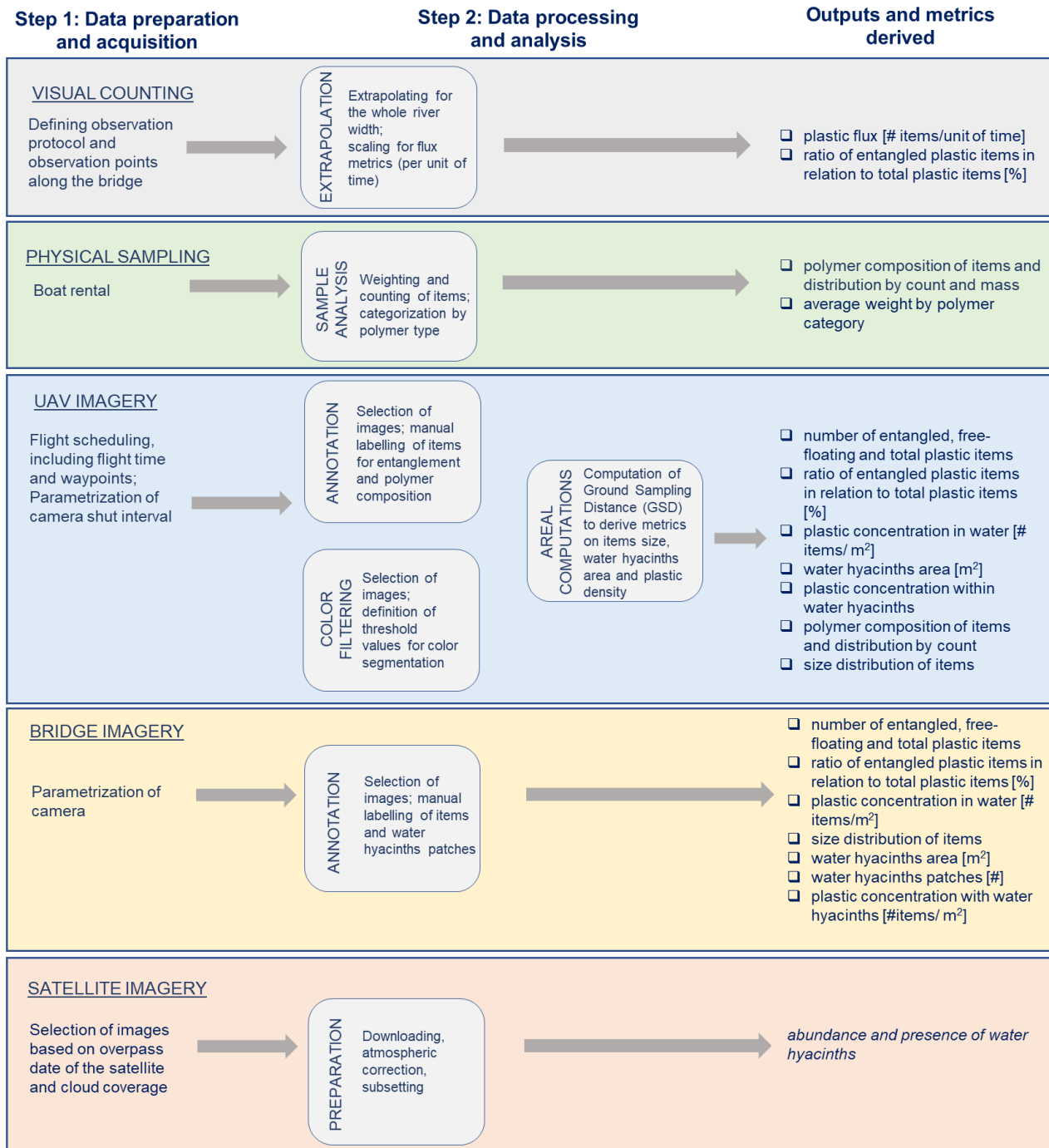
## 112 **2 Materials and Methods**

113 In this section, we present the five measuring methods that were adapted to monitor the role of water hyacinths  
 114 in macroplastic entrapment and transport. We first provide an overview of the measuring techniques used (figure  
 115 2). We summarize the data preparation and processing steps necessary to derive the outputs presented, for  
 116 each method. This overview can be beneficial for a first understanding of the metrics that can be derived by  
 117 each measuring technique and get a sense of the time, effort and resources required for the data collection and  
 118 analysis. Practical considerations on how to use the various measuring methods are further explained in the  
 119 text, with a dedicated sub-section for each method. There, we first provide a general description of the technique  
 120 – including different options in the measurement set-up – and then detail how we adapted these methods for  
 121 the field measurements at the Saigon river.

122 The field campaign was conducted in Ho Chi Minh City (HCMC), Vietnam's most populated city. The Saigon  
 123 river runs through the city and joins the Dong Nai river a few kilometers downstream from HCMC. All field  
 124 measurements were done in an area close to the Thu Thiem bridge. Several factors determined the choice of  
 125 this site, such as the accessibility to the riverbank in that area (required for UAV surveys and for the physical  
 126 sampling) and the presence of a bridge on which it is safe for surveyors to stand (used for visual observations)

127 and not too height to allow plastic item detection. This site has been used for river plastic monitoring extensively  
 128 since 2018, including studies focused on the development of new methods to monitor macroplastic transport  
 129 (van Emmerik et al., 2018; van Emmerik, Strady, et al., 2019) and to monitor the role of water hyacinths in  
 130 macroplastic transport (Schreyers et al., 2021a).

131 Given that flow velocity and tidal dynamics could also influence macroplastic flux and plastic cross-sectional  
 132 concentrations (Haberstroh et al., 2021; van Calcar & van Emmerik, 2019), these were also characterized during  
 133 the field campaign. Ultimately, flow velocity is a key metric to understand transport dynamics in rivers, including  
 134 plastic debris movements. Each visual counting measure was accompanied by an estimate of the flow velocity  
 135 using the 'Pooh Sticks' method (Bull & Lawler, 1991). The tidal regime was also noted during the visual  
 136 observations.



137

138 **Figure 2.** Flowchart illustrating and summarizing steps required for deriving metrics on water hyacinths  
 139 distribution, macroplastic transport, macroplastic entrapment in water hyacinths and plastic items  
 140 characteristics.

## 141 **2.1 Visual observations**

142 Macroplastic flux measurements can be conducted using the visual counting method (van Emmerik, Strady, et  
 143 al., 2019). This method consists of counting the number of macroplastic items visible at the water surface and  
 144 flowing downstream for a specified duration from a bridge. For large rivers, several observations points are  
 145 usually defined along a bridge due to the influence of local hydrodynamic conditions across the river (van  
 146 Emmerik, Strady, et al., 2019). In order to monitor the plastic flux of items entrapped in hyacinth mats and free-  
 147 floating items, this method can follow the protocol elaborated in Schreyers et al. (Schreyers et al., 2021). For a  
 148 determined duration, the surveyor can count at one observation point all plastic items visible within floating water  
 149 hyacinths. Immediately after, plastic litter outside the hyacinths can be counted for same duration. The surveyor  
 150 then moves to the next observation point and repeats the counting. Alternatively, simultaneous counting of  
 151 plastic items within and outside hyacinths can be done at one observation point by two surveyors. In this case,  
 152 one surveyor counts all plastic litter visible inside hyacinth patches, while at the same time a second surveyor  
 153 counts the number of items found outside the hyacinth patches at the same observation point. This second  
 154 option in the visual option protocol could help in reducing the variation in results due to a time difference. The  
 155 measurement duration can be adapted depending on the flow velocity. In low discharge conditions,  
 156 measurements of 5 to 20 min might be necessary (Vriend, van Calcar, et al., 2020), whereas for higher plastic  
 157 fluxes, measurements typically last 2 min. The number of observation points along a bridge can be adapted,  
 158 depending on the river width. Ideally, the observation points are equally distant from each other and enable to  
 159 cover most of the river width. It is estimated that at each observation point, the surveyor is able to see 15 m of  
 160 the river width.

161 For the Saigon river in-situ survey, the visual counting was done by one surveyor, with two subsequent measures  
 162 of plastic items for each observation point. The observations were conducted from the Thu Thiem bridge on the  
 163 23 May 2020, between 6:45 and 16:15. The bridge was on average 14.6 m above the water level. The measures  
 164 were made at 11 observation points along the bridge. Covering all the bridge observation points took 50 minutes  
 165 to 1h30, depending on the time dedicated to taking photographs, as well as breaks. Over the whole day, five  
 166 sets of observations were made for the 11 observations points. Subsequent data analysis included extrapolation  
 167 of plastic flux for the entire river width and conversion to obtain flux values (Figure 2). More information can be  
 168 found on Supplementary Materials (text S1).

## 169 **2.2 Physical sampling**

170 The plastic composition and mass of items can be measured via active sampling. Using a boat, surveyors  
 171 retrieve samples of floating hyacinths at the water surface. Sampling of free-floating items can also be taken,  
 172 using for instance trawls or mantra nets. The samples are then taken to a laboratory for further analysis. There,  
 173 the wet mass of each hyacinth and/or plastic sample is measured. For the hyacinth samples, the patches area  
 174 can be computed by dividing the total hyacinth mass by the average plant biomass per m<sup>2</sup>.

175 In the laboratory, anthropogenic debris are taken apart from the vegetation. The debris are then dried either by  
 176 air or using an oven. Air-drying is preferable when the samples are large. The dried items can be sorted and  
 177 counted by categories. The categorization method depends on the research objectives and resources available.  
 178 A simple categorization method would separate plastic items from non-plastic debris (such as glass, textile,  
 179 metal or rubber). More detailed categorization include the grouping of items per polymer category or their  
 180 classification by item identification, for instance following the River-OSPAR litter categorization (Vriend,  
 181 Roebroek, et al., 2020). Following their classification, items are weighted in order to retrieve dry mass statistics.  
 182 The weighting can be done separately for each item or for an entire category group. Additional measures can  
 183 include the measurement of the size of plastic items, or a qualitative indication on the level of degradation and  
 184 fragmentation of each item.

185 At the Saigon river, a team of two surveyors collected 16 samples of water hyacinths of approximately 1 m<sup>2</sup>,  
 186 using a boat close to the Thu Thiem bridge. The samples were taken on the 23 May 2020, between 7:30 and  
 187 15:00. No samples of free-floating items were taken during this measurement campaign. The subsequent mass  
 188 and count measurements were conducted at the Asian Water Research Center (CARE) laboratory in Ho Chi  
 189 Minh City. For each sample, the wet mass of hyacinths was weighted. Hyacinth patches areas were then

190 calculated based on the wet mass of hyacinths and the plant biomass. Existing literature indicated an average  
191 wet water hyacinth biomass of 15-30 kg for 1 m<sup>2</sup> (Reddy & Sutton, 1984). The plastic items were air-dried, then  
192 sorted and counted by plastic categories (Vriend, van Calcar, et al., 2020). Six plastic categories were retained:  
193 polystyrene (PS), expanded polystyrene (EPS), polyethylene terephthalate (PET), soft polyolefins (PO soft),  
194 hard polyolefins (PO hard) and other plastics (Rest). The weighting of the plastic items was determined by  
195 plastic group. These measures enabled to derive metrics on the plastic concentration within hyacinth patches  
196 (figure 2).

### 197 **2.3 UAV imagery**

198 UAVs (Unmanned Aerial Vehicles) can be used for riverine monitoring of both plastics and hyacinths distribution.  
199 UAV surveys can be conducted along riverbanks or transecting the river. The latter enables to detail the  
200 distribution and variability of plastic items and hyacinths at a river cross-section. To minimize human errors, it is  
201 preferred to program the flights automatically, using an app such as Litchi or DJI GO 4, rather than flying the  
202 UAV manually. These applications enable to define flight parameters such as the elevation, speed, path, number  
203 and location of hovering waypoints along the path, as well as set-up the camera, including its orientation angle  
204 and mode of acquisition of images. Recent applications of UAV surveys show that a flying elevation comprised  
205 between 4 and 6 m above the water level enables to distinguish macroplastic items (Geraeds et al., 2019). Other  
206 considerations such as the presence of trees, buildings or fences along the riverbanks are also important for  
207 defining the flying elevation. To avoid orthorectification of the images in post-processing, it is recommended to  
208 set-up the gimbal angle of the camera at 90°. Several flights can be conducted at the same river cross-sections,  
209 due to the possibility to set-up flights in automatic mode. The flights can be programmed with a 'stop and go'  
210 modality, in order to take several pictures at each predefined waypoint and to allow time for the device  
211 stabilization.

212 We conducted ten UAV flights upstream and downstream to Thu Thiem bridge, on 23 May 2020. The flights  
213 conducted during the ebb tide crossed the river 100 m downstream of the bridge (n = 4), those conducted during  
214 the flood tide at 80 m upstream (n = 6). A DJI Phantom 4 Pro UAV (DJI, Shenzhen, China; <http://www.dji.com>)  
215 and its 20 megapixels standard integrated sensor was used for RGB image acquisition. Each UAV flight crossed  
216 the river perpendicularly to the water flow, at an elevation of approximately 5 m above the water level. This flying  
217 elevation was chosen because it optimizes the number of pixels in the images and still allows to identify plastic  
218 categories. Field measurements that only seek to count the number of items, without categorizing them into  
219 plastic types, might prefer a higher elevation of approximately 7-10 m above the water level. A total of 22  
220 waypoints were determined for the entire river width (approximately 320 m), and the UAV hovered for 14  
221 seconds at each waypoint.

222 A total of 3,936 images were taken during these ten UAV surveys. Some images were discarded because they  
223 were blurry (n = 261). All duplicates were removed (n = 3,547) and we ultimately used only the best images  
224 taken at each waypoint (n = 128). An online annotation tool, the Visual Geometry Group Image Annotator (VIA),  
225 was used to manually label items by plastic category and to indicate whether items were entangled in hyacinths  
226 or not. The plastic categories were the same used for the physical sampling. Rectangular-shaped bounding  
227 boxes were drawn around each identified item, which allowed to estimate the size of plastic debris. In addition  
228 to plastic identification, the UAV image collection also enabled to inform water hyacinths coverage and  
229 distribution across the river cross-section. To estimate the surface covered by the floating hyacinths, we used  
230 the Open CV library in Python. Only images with visible water hyacinths were retained for analysis (n = 75). The  
231 presence of hyacinth patches was assessed through visual examination of the UAV image collection. The Open  
232 CV library was used to discriminate pixels with water hyacinths using color segmentation functions. The color  
233 filtering can discriminate pixels with a specific color, using the RGB values of each image. A mask to select  
234 green areas of floating vegetation was created and defined by thresholds in the RGB space. The upper and  
235 lower threshold values were adjusted by trials and errors for small subsequent group of images, due to  
236 differences in background reflectance among the entire UAV image collection. More information can be found  
237 on Supplementary Materials (texts S2 and S2, figure S1 and S2 and Table S1).

### 238 **2.4 Camera imagery from bridges**

239 Photographs taken from bridges can also provide information on the plastic entrapment profile across the river  
240 and on water hyacinths distribution. Using camera photographs from bridges to take snapshots of the river  
241 surfaces makes it possible to detect floating macroplastic items and water hyacinth patches. The identification  
242 of plastic debris and vegetation can then be done either automatically or manually. The automatic detection

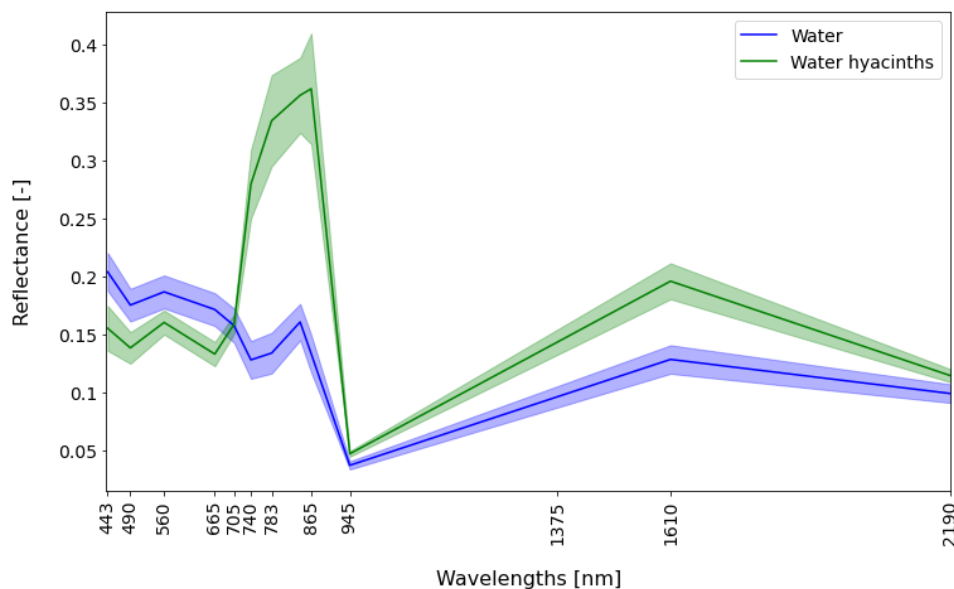
243 presents certain challenges, especially considering the presence of water hyacinths which adds complexity for  
 244 object detection and image processing.

245 The photographs were taken from the Thu Thiem bridge on the 23 May 2020, at an altitude comprised between  
 246 12 and 15.9 m, depending on the varying water level throughout the day. The photographs were taken nearly  
 247 simultaneously with the visual counting measurements, at the same 11 observation points. A Samsung camera  
 248 (SM-J330G, samsung.com, South Korea) was used to take the pictures. The camera was held by hand without  
 249 a stabilization device, which resulted in some blurry images. The photographs were captured just after the visual  
 250 counting measurements were completed at each observation point. Typically, one to four photographs of the  
 251 river surface were taken after each measurement. The surveyor faced the river downstream with the camera  
 252 oriented perpendicular to the water flow during the ebb tide. During the flood tide, the surveyor took photographs  
 253 facing upstream. In total, 139 photographs were taken.

254 From the total image collection, 83 photographs were used for the detection of plastic debris and hyacinth  
 255 patches. Images discarded ( $n = 57$ ) were either too blurry for identifying items or did not have hyacinths nor  
 256 plastic items on them. The images were uploaded into the VIA online annotation tool. Plastic items were  
 257 identified and rectangular shaped bounding boxes drawn around these elements. The annotation tool enabled  
 258 to specify if plastic items were entangled in patches or freely floating in the open water. The categorization of  
 259 items by polymer type was not possible due to poor visibility. The estimates of hyacinth patches areas were also  
 260 done using bounding boxes. This method was preferred over the color segmentation approach used for the UAV  
 261 imagery, because the lower quality (low contrast, presence of blurry images) of the bridge imagery dataset did  
 262 not enable easy filtration of vegetation pixels. In addition, the bounding box method has the advantage of  
 263 providing information about the average hyacinth patch sizes and the number of patches observed.

## 264 2.5 Satellite imagery

265 Field measurement techniques enable to precisely map the water hyacinths area at a river cross-section; but  
 266 cannot give a synoptic view of the water hyacinths distribution and abundance for a river system. Satellite  
 267 imagery, on the other hand, can cover part or even an entire river system. Mapping the distribution of water  
 268 hyacinths is possible due to the distinct spectral signals of the aquatic vegetation compared to the surrounding  
 269 water (figure 3). The distinction between water hyacinth and terrestrial riverbank vegetation can prove  
 270 challenging due to similarities in their spectral information and thus requires an exact delineation of the river  
 271 edge, for instance by masking the water area on a satellite scene with low water hyacinth presence.



272

273 **Figure 3.** Mean spectral signatures of water hyacinths ( $n = 26$ ) and water ( $n = 12$ ) from the Sentinel-2 images  
 274 taken on 22 May 2020. The x-axis shows the Sentinel-2 MSI spectral range from visible blue light at 490 nm, to  
 275 short-wave infrared light at 2190 nm. The y-axis shows remote sensing reflectance (unitless) from Sentinel-2 for  
 276 water hyacinths and water for selected pins.



277 Water hyacinths, as other types of vegetation, show a high reflectance peak in the NIR wavelengths of light  
 278 (~850 nm), and absorption in the red (~660 nm). Water, on the other hand, reflects highly in the green spectrum  
 279 of light (~560) and absorbs in the vegetation-red edge bands (~705 and 740). Water pixels also show a peak in  
 280 the NIR, which could be due to high loads of sediment in the water and/or eutrophication. Satellite remote  
 281 sensing can be an effective technique to collect high quality and standardized optical scenes for the detection  
 282 of floating aquatic vegetation (Dogliotti et al., 2018; Schreyers et al., 2021b)

283 We used a Sentinel-2 imagery scene to qualify the spatial distribution of water hyacinths in the Saigon river. A  
 284 satellite image was captured over the Saigon river on the 22 May 2020 by the Sentinel-2A multispectral sensor  
 285 at 3:05:51 UTC time, corresponding to 10:05:51 local time. This was the closest available Sentinel-2 scene by  
 286 date from the field measurements, all conducted the day after, on 23 May 2020. The Sentinel-2A/MSI (S2A)  
 287 Level 1C files were retrieved from the European Space Agency (ESA) Copernicus Open Access Hub  
 288 (<https://scihub.copernicus.eu>). and processed for atmospheric correction on the ESA software Sentinel  
 289 Application Platform (SNAP), version 8.0. Atmospheric correction was applied to the Level-1C image to generate  
 290 a Level-2 scene, using the Sen2Cor 280 processor. Further, the image was resampled to 10 m spatial resolution  
 291 and cropped for a region of interest focusing on the Saigon river area, to reduce the scene size and the  
 292 processing time. A false color composite was generated to visually examine areas of water hyacinths  
 293 accumulation. The imagery covers approximately 69 km of the total 225 km of river length of the Saigon river –  
 294 31% of its total length.

## 295 **2.6 SWOT analysis**

296 In order to orient the choice of the best (combination) of measuring methods to choose for future monitoring  
 297 campaign, we conducted a Strengths, Weaknesses, Opportunities, and Threats (SWOT) analysis (Menon et al.,  
 298 1999). This type of analysis comes from an interdisciplinary approach and is useful to identify challenges and  
 299 factors that can influence the set-up of a monitoring strategy. The goal of this type of analysis is to identify the  
 300 main strengths, weaknesses, opportunities, and threats of the system under consideration. Overall, a SWOT  
 301 analysis helps the identification of strengths and weaknesses of future riverine plastic and vegetation monitoring  
 302 strategy to achieve its goals, pinpointing to specific benefits and drawbacks of each technique.

## 303 **3 Results and Discussion**

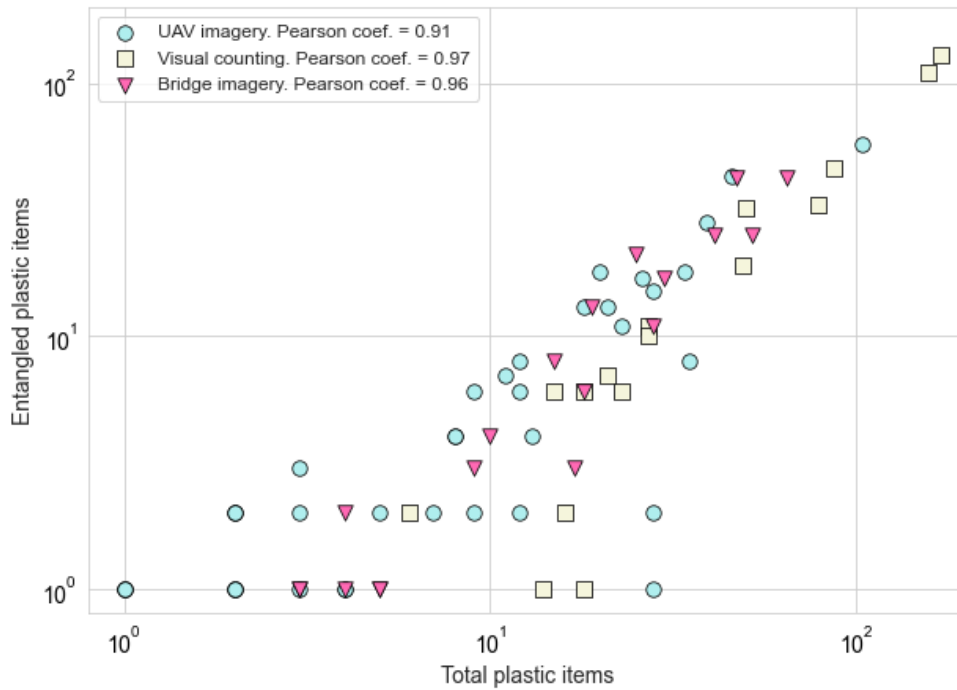
304 The methods described provide unique insights on four key aspects of plastic entrapment in water hyacinths:  
 305 (1) plastic transport and entrapment within water hyacinths, (2) plastic items characteristics (3) water hyacinths  
 306 spatial distribution (4) hydrological influences on spatiotemporal plastic entrapment in hyacinths. The results of  
 307 the field monitoring campaign at the Saigon river on these four aspects and their respective sub-elements are  
 308 hereby presented and discussed. Our scope is to present the level of information that each measuring technique  
 309 can bring on the abovementioned four aspects.

310 Secondly, we provide a broader overview for designing a monitoring strategy, to ultimately determine what  
 311 (combination of) measuring methods to choose. The combination of metrics of interest and measuring technique  
 312 is first discussed. Secondly, we examine the relevancy of each technique at different spatiotemporal scales.  
 313 Lastly, the overall benefits and drawbacks of each technique are detailed with a SWOT analysis.

### 314 **3.1 Plastic transport and entrapment in water hyacinths**

#### 315 **3.1.1 Entrapment of floating macroplastics in water hyacinths**

316 The UAV imagery, visual counting and bridge imagery techniques register similar proportions of floating  
 317 macroplastics entrapped in water hyacinths: on average, 39%, 45% and 51% of items was found entangled,  
 318 respectively. These three field measurement techniques indicate a statistically significant ( $p < 0.05$ ) and very  
 319 high positive correlation between entangled items and total floating macroplastics (figure 4). This demonstrates  
 320 that water hyacinths entrap a significant portion of the total floating macroplastic fluxes in the Saigon river. The  
 321 good agreement between the three measuring techniques is also an indication of their validity in characterizing  
 322 the entanglement of floating macroplastic in hyacinths.



323

324 **Figure 4.** Entangled plastic items in relation to total floating macroplastics as observed in the Saigon river near  
 325 Thu Thiem bridge. Each point corresponds to one observation (i.e.: one visual counting measurement, one  
 326 camera photograph and one UAV image). Note the logarithm scale for both axes.

327 **3.1.2 Plastic item concentration in water hyacinths**

328 The plastic concentration in hyacinths derived from the UAV imagery is double than the estimate from the bridge  
 329 imagery dataset (table 1). The physical sampling found a plastic concentration comprised between 17 and 34  
 330 items per m<sup>2</sup> of vegetation, one order of magnitude higher than the imagery based values. The higher  
 331 concentration value found by the UAV imagery when compared with the bridge imagery is a likely result of UAV  
 332 higher resolution due to shorter distance between the sensor and the river surface. This higher concentration  
 333 range found with the samples is the likely result of the specificity of the physical sampling method, where entire  
 334 hyacinths patches were retrieved and analyzed. As a result, submerged and small plastic particles were also  
 335 counted and weighted, contrary to the imagery datasets. Considering that the hyacinth roots can be over 1 m  
 336 long, plastic debris may get entangled in the rhizosphere. Thus, the higher plastic concentration found by the  
 337 physical sampling technique might be the result of entrapment mechanisms of plastic in the submerged part of  
 338 hyacinths. Further studies are needed to better understand the role of hyacinth roots in entrapping plastic debris.

339 In addition, the physical sampling enabled to derive metrics on the mass concentration of plastic items within  
 340 hyacinths. A total of 380 kg of wet hyacinths biomass were sampled, and 2.1 kg of plastic were found, which  
 341 indicates a mean mass concentration of 6.6 g of dry plastic per kg of wet hyacinth biomass.

342

343

344 **Table 1.** Plastic concentration in hyacinth patches. The concentrations are all mean values. The physical  
 345 sampling area concentration is expressed as a range, as it considers the range of 15-30 kg of wet water  
 346 biomass for 1 m<sup>2</sup> of hyacinth area (see Methods section).

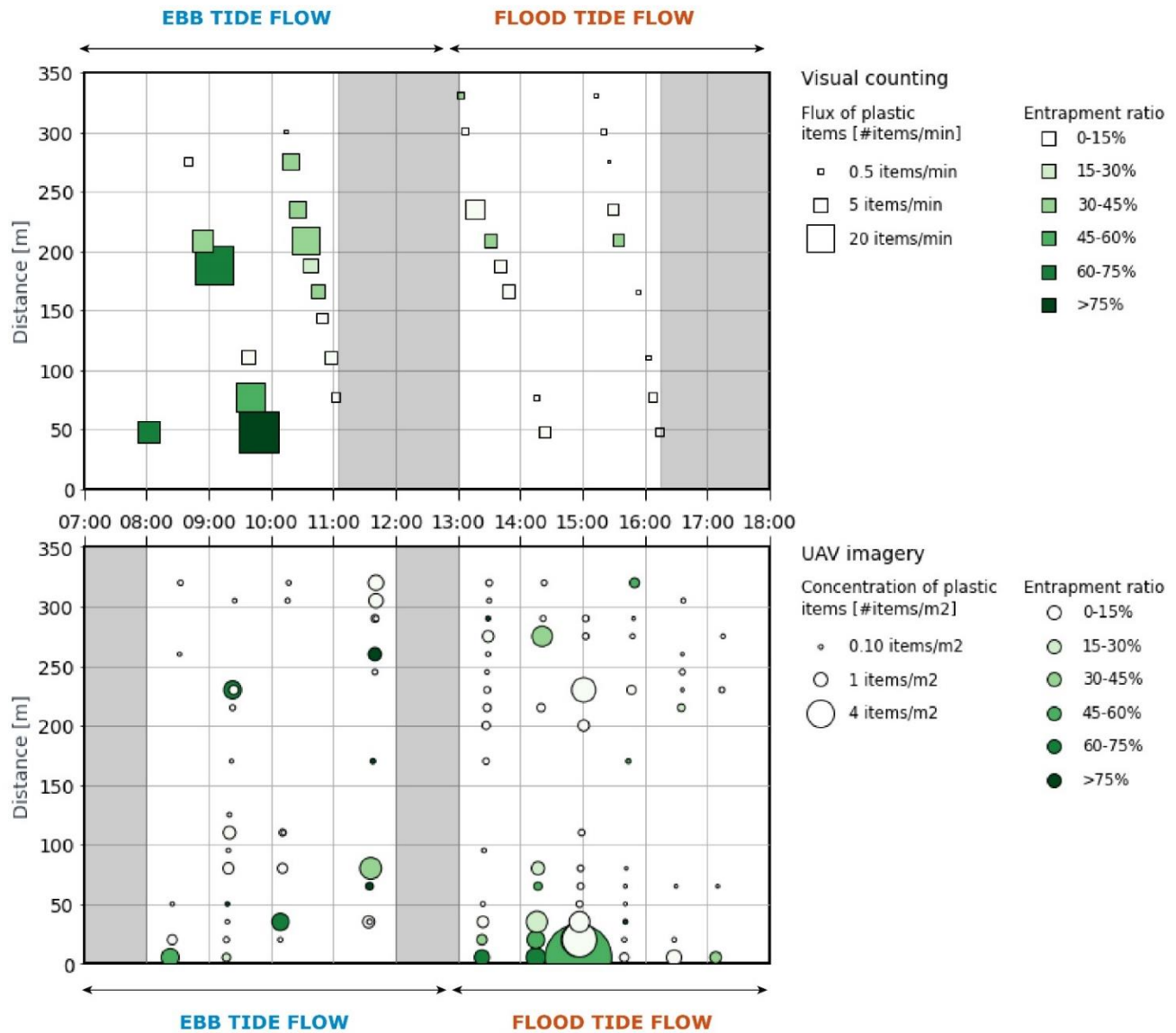
Measuring technique	Plastic concentration in hyacinths	
---------------------	------------------------------------	--

	Area concentration [# items/m <sup>2</sup> ]	Mass concentration [g/kg]	Total river sampled surface [m <sup>2</sup> ]
<b>UAV imagery</b> ( <i>n</i> = 75)	2.14	N/A	2,575
<b>Bridge imagery</b> ( <i>n</i> = 82)	0.75	N/A	4,227
<b>Physical sampling</b> ( <i>n</i> = 16)	17-34	6.6	N/A

347

### 348 **3.1.3. Spatiotemporal variability of macroplastic and water hyacinths**

349 We compared the spatiotemporal variability throughout the measurement day of floating macroplastic  
350 concentrations and flux, derived respectively from the UAV imagery and the visual counting techniques (figure  
351 5). Floating macroplastic concentrations and fluxes vary along the river width, ranging between 0.05 and 23.7  
352 items/m<sup>2</sup> and between 0.00 and 42.8 items/min, respectively. Some similarities are noticeable between the two  
353 datasets analyzed. The results of visual counting and the UAV imagery indicate that high fractions (above 45%)  
354 of entrapped plastics were registered in the first 50 m from the northern riverbank and at 240 m. Both  
355 measurement techniques also registered low ratios of entrapment, plastic concentrations and flux in the river  
356 sections comprised between 100-150 m and 300-350 m. However, many differences are also noticeable, with  
357 high macroplastic fluxes not necessarily corresponding to high plastic concentrations, and vice-versa. This was  
358 observed for instance at 10h., at 50 m from the northern riverbank, where a flux of macroplastic as high as 42.8  
359 items/min was measured, but concentrations were only of approximately 1 item/m<sup>2</sup> of surface. Several factors  
360 may account for this discrepancy between the plastic flux and the plastic concentration in water.



361

362 **Figure 5. Spatiotemporal distribution of macroplastic concentration and entrapment ratios on 23 May**  
 363 **2020 at Thu Thiem.** Each point/square corresponds to one observation. The greyed areas represent period  
 364 where no measurements were conducted due to breaks. The distance is calculated using the northern riverbank  
 365 as the origin. It should be noted that the flux of plastic items during ebb tide is seaward and landward during  
 366 flood tide.

367 Firstly, the observations from the visual counting and the UAV imagery do not strictly coincide in location and  
 368 time. Hence, the noted discrepancies could indicate rapid changes in plastic transport and entrapment between  
 369 the observations, even when relatively close in time and space. For instance, the first 50 m of the river were not  
 370 monitored through visual counting, whereas three waypoints with the UAV track enabled to detect plastic close  
 371 to the northern riverbank of the Saigon river. Conversely, the UAV imagery could have missed for instance a  
 372 water hyacinth patch that was captured by the visual counting technique. Secondly, the UAV flights were  
 373 conducted at 5 m of elevation approximately, three times lower than the elevation of the bridge. This resulted in  
 374 a much higher visibility of the plastic items, which was also enhanced by the possibility to zoom in on the images  
 375 when manually labelling the items. The UAV imagery allows to spot debris that might be otherwise missed when  
 376 counting macroplastic items from the Thu Thiem bridge. Following this explanation, the visual counting  
 377 technique might thus underestimate the overall macroplastic flux. Thirdly, another possible factor pertains to the  
 378 nature of the measurements undertaken. The UAV imagery technique takes 'snapshots' of the macroplastic  
 379 concentrations, thus not capturing the propagation of such plastic concentrations depending on their flux,

380 contrary to the visual counting technique. A segment of the river with high macroplastic flux could register low  
 381 concentration of items per m<sup>2</sup> at the water surface, and vice-versa in the case of a slow streamflow. It is likely  
 382 that this phenomenon is registered close to the riverbanks, where the flow velocity is lower. In the first 50 m of  
 383 the river, high plastic concentrations were found, and visual examination of the images confirms the presence  
 384 of accumulation zones of hyacinths and debris very close to the northern riverbank. These considerations are  
 385 an argument in favor of combining several monitoring methods rather than favoring one. In section 3.4. we will  
 386 further discuss the role of stream flow velocity in plastic transport and entrapment across the river section.

### 387 **3.2 Plastic items characteristics**

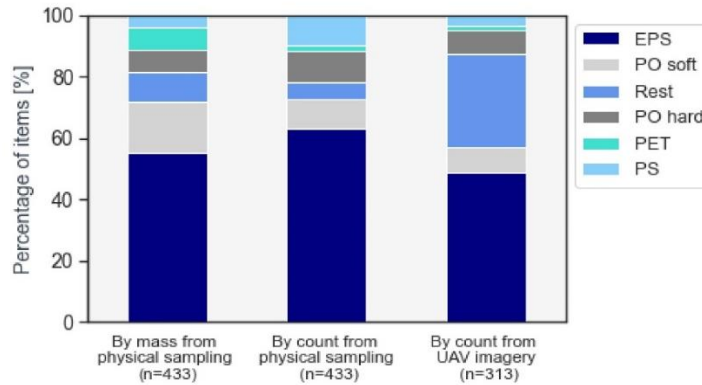
#### 388 **3.2.1 Plastic categories and items characteristics**

389 Figure 6 presents the composition of plastic items entrapped in hyacinths, based on the physical sampling and  
 390 the UAV imagery. Expanded polystyrene (EPS) was the most abundant type found overall: 48% based on the  
 391 UAV imagery, 63% for the physical sampling. Items made of EPS easily fragment, possibly explaining the higher  
 392 proportion of these items found in the collected sample. PET (polyethylene terephthalate) items were the least  
 393 found polymer type, in the same proportion for both methods (1.6%). PET items are easy to identify, due to the  
 394 relative large size of these items (usually above 15 cm) and are often distinct in shape (plastic bottles), even  
 395 with an aerial view and entrapped in floating vegetation. PO soft (soft polyolefins) debris - bags, foils - and PO  
 396 hard (hard polyolefins) – such as lids and toys – were also found in similar proportions among both methods  
 397 (PO soft: 9 and 8%, PO hard: 10 and 8% for physical sampling and UAV imagery).

398 The category 'Rest' is found in considerably higher proportions for the UAV imagery based analysis than the  
 399 physical sampling (30% and 6% respectively). The item classification from the physical sampling yields more  
 400 accurate and reliable results, since it is done through close visualization and physical handling. The high  
 401 proportion of 'Rest' items found by the UAV imagery analysis could be the result of this higher level of inaccuracy  
 402 for characterizing plastic categories with this method. Items that were unclear (blurry, no shape or dominant  
 403 material could be identified, partially hidden by vegetation, glare reflection affecting their visibility, etc.) were  
 404 automatically attributed this label. Also, the physical sampling dataset only reflects the composition of items at  
 405 the sampling sites, whereas the UAV imagery enables to cover the entire river width. This time and space  
 406 mismatch in datasets could also explain the discrepancy noted in the proportion of 'Rest' items. Lastly, PS  
 407 (polystyrene) items are found in considerably higher proportion by the physical sampling (10%) than UAV  
 408 imagery (3%). Several PS items were probably categorized as 'Rest' through manual labelling.

409 The physical sampling method is the only measurement technique that enables to directly derive statistics on  
 410 the mass of plastic items (figure 6b). PET items are the heaviest found, at 22.3 g on average per item. This is  
 411 the result of the larger size of these items and the fact that they were often found unfragmented. EPS and PS  
 412 items were, on the contrary, more often found fragmented.

a)



b)

Plastic category	EPS	PO soft	Rest	PO hard	PET	PS	Total
Count of items	273	41	25	44	7	43	433
Mean mass (g)	4.3	8.3	8.2	3.6	22.3	1.8	4.85

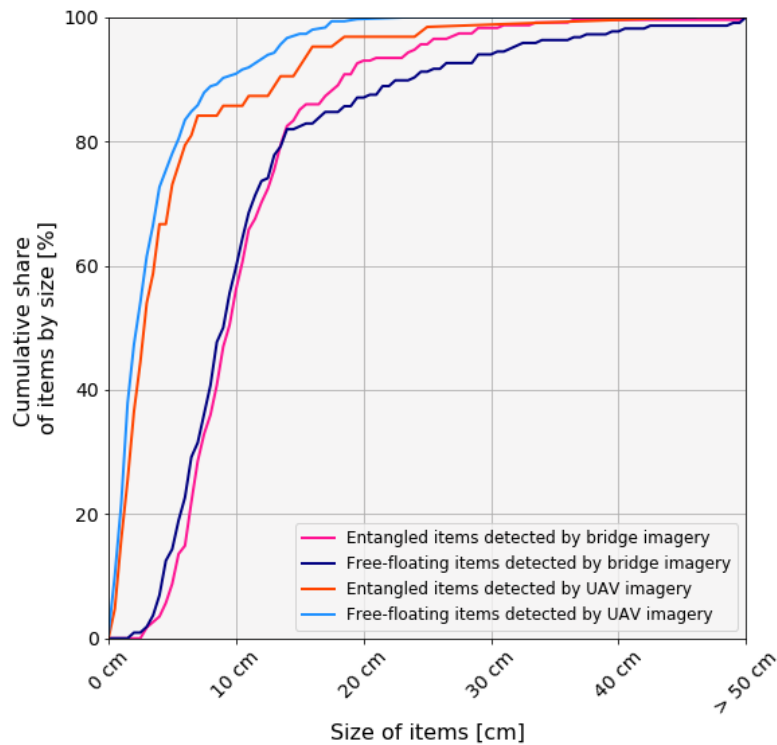
413

414 **Figure 6.** a) Plastic categories of observed items from physical sampling and UAV imagery analysis. Expanded  
 415 polystyrene (EPS), soft polyolefins (PO soft), other plastics (Rest), hard polyolefins (PO hard), polyethylene  
 416 terephthalate (PET), and polystyrene (PS). b) Mean masses of plastic items derived from the physical sampling.

### 417 3.2.2 Plastic items size

418 The UAV and bridge imagery datasets enabled to estimate the size of each identified plastic item (figure 7).  
 419 Almost all items identified on images taken from the bridge are above 5 cm of size, both for the entangled and  
 420 free-floating items (respectively, 94 and 86% of the total count of items). Very few items were detected below  
 421 2.5 cm (n = 2), none of which were smaller than 1 cm of size. For the UAV imagery, only 33% of entangled items  
 422 are above 5 cm, and 25% of free-floating. Both methods show that hyacinths aggregate larger items than  
 423 otherwise freely observed in the river. The UAV surveys found that items larger than 2.5 cm were more frequently  
 424 entrapped than not. The bridge imagery shows a similar pattern for items larger than 5 cm, except for the largest  
 425 objects (above 50 cm of size). However, very few items above 50 cm of size (n = 3 for the bridge imagery, n =  
 426 0 for the UAV imagery) were found.

427 The UAV images were taken at a lower altitude (approximately 5 m) compared to the images taken from the  
 428 bridge (between 12 and 15.9 m), allowing to better detect small items. As a result, the spatial resolution for the  
 429 UAV images is three times higher (0.1 cm/pixel on average) than for the bridge imagery (0.3 cm/pixel on  
 430 average). The lower quality of the camera used from the bridge and the absence of stabilization devise also  
 431 explain that items below 5 cm of size were not found abundantly with this method. Plastic items below 2 cm of  
 432 size were not spotted at all by the bridge imagery dataset, whereas the UAV images detected items as small as  
 433 0.23 cm.



434

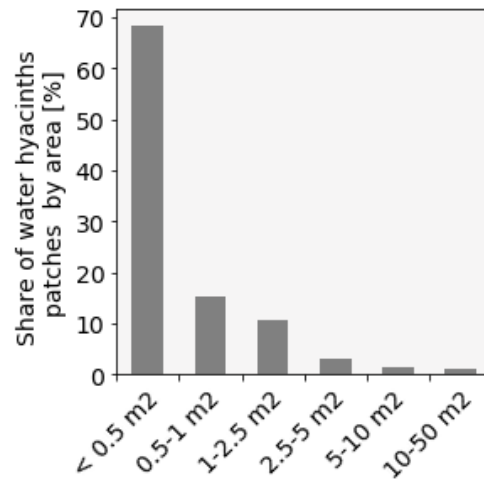
435 **Figure 7.** Size distribution of plastic items from the bridge (n = 444) and UAV imagery (n = 359).

### 436 3.3 Water hyacinths spatial distribution

#### 437 3.3.1 Water hyacinths coverage area at Thu Thiem

438 The bridge imagery found that hyacinth patches had a median area of 0.29 m<sup>2</sup>. Almost all (97%) hyacinth  
 439 patches found were smaller than 5 m<sup>2</sup>, accounting for 63% of the total hyacinth area found (figure 8). The visual  
 440 examination of the bridge photographs shows that the hyacinth patches were often adjacent to each other.

441 Analyzing both the UAV and bridge images enabled to estimate the water hyacinths area (table 2). The total  
 442 area of water hyacinth patches based on UAV images (111 m<sup>2</sup> during entire observing period) was a bit over  
 443 one-third of that assessed based on the bridge images (305 m<sup>2</sup>). This discrepancy can be explained by the fact  
 444 that 2-3 images were taken at the same observation point on the bridge (one after the other), whereas  
 445 overlapping images with presence of water hyacinths were less frequently found in the UAV imagery dataset.  
 446 Another factor to account for this difference relates to the method used in quantifying water hyacinths area. The  
 447 estimate of water hyacinths area was done through a color segmentation process for the UAV images (pixel-  
 448 based), whereas for the bridge images, rectangularly shaped bounding boxes were manually drawn around the  
 449 visually identified water hyacinths. This probably results in an overestimation of water hyacinths area for the  
 450 bridge images.



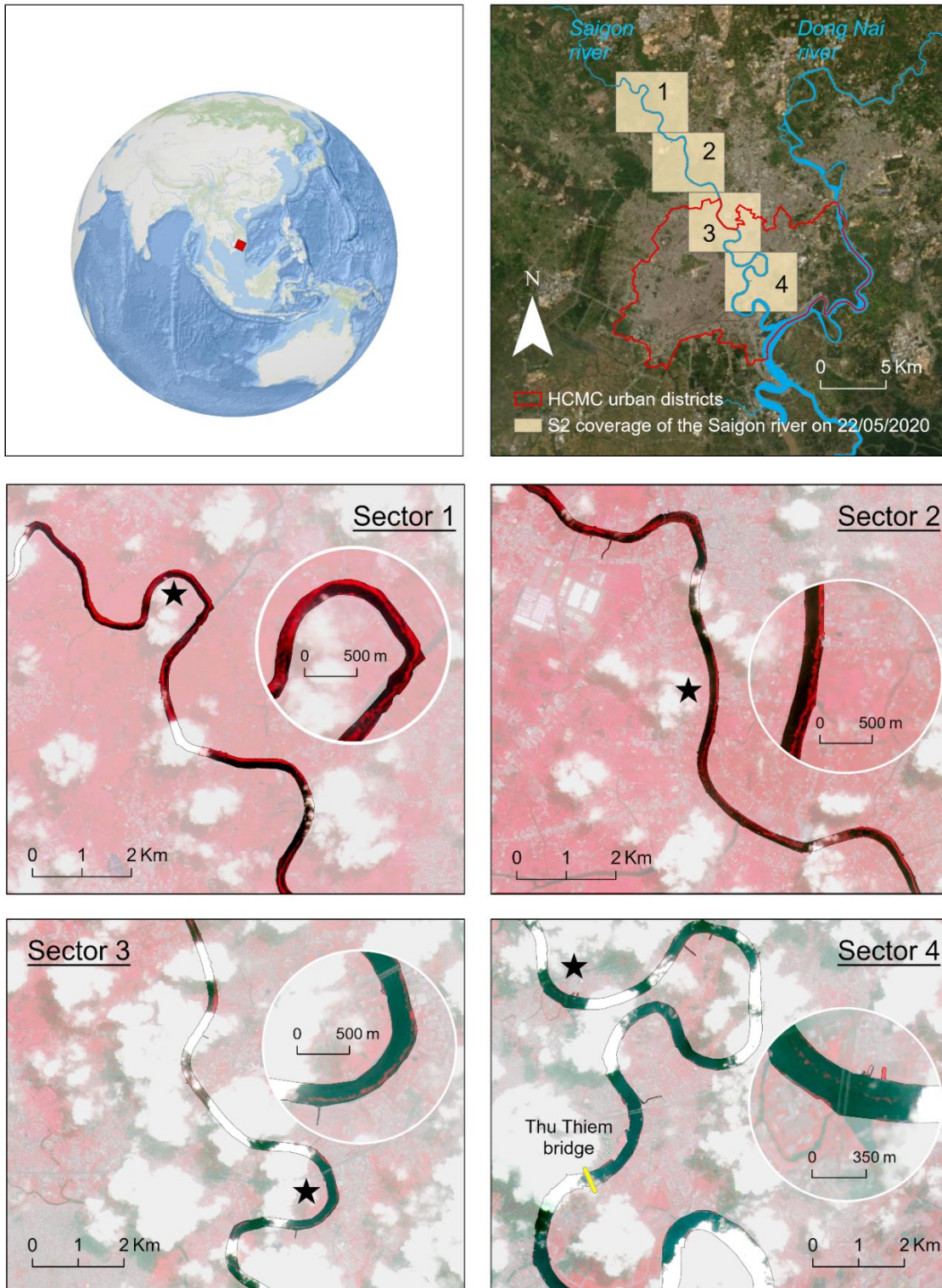
451 **Figure 8.** Area distribution of water hyacinths patches from the imagery (n = 369) taken at the Thu Thiem  
 452 bridge.

### 453 3.3.2 Water hyacinths spatial distribution in the Saigon river

454 Figure 9 displays Sentinel-2 imagery taken on 22 May 2020 (sector 4). The two upstream sectors show important  
 455 accumulations of water hyacinths, sometimes taking up almost entirely the river width. The hyacinth patches in  
 456 these upstream sectors form long trails, which tend to aggregate mostly on the eastern river bank. In the two  
 457 downstream sectors, the Saigon river progressively becomes wider and enters the most densely populated  
 458 areas of Ho Chi Minh City. The satellite image shows that pixels suggesting water hyacinths abundance are  
 459 more concentrated in the upstream sections. Further downstream (sectors 3 and 4) the hyacinths mats seem  
 460 less dense, continuous and large than in the upstream sector, but pixels indicating vegetation content are still  
 461 visible at some locations. In those sections, the cloud coverage does not enable a continuous mapping of the  
 462 hyacinths presence. The relatively small size of patches that was estimated at Thu Thiem (on average 0.8 m<sup>2</sup>)  
 463 is also an indication of a possible change in water hyacinths characteristics along the river. At Thu Thiem, the  
 464 cloud coverage hampers visibility and thus only a few pixels that could indicate presence of vegetation were  
 465 visually detected.

466 We hypothesize that these large patches detected by satellite imagery in the upstream sections break down due  
 467 to boat traffic in Ho Chi Minh City which typically occurs in the mid-channel of the Saigon river. Another  
 468 explanation for the reduced amount and size of the drifting hyacinth patches in the downstream sections might  
 469 be increased flow velocities. Above a threshold of 0.4 m/s, the hyacinth patches become unstable and break  
 470 apart more easily (Petrell & Bagnall, 1991). The average flow velocity measured on 23 May 2020 at Thu Thiem  
 471 bridge exceeded 67% of the time this threshold (on average: 0.54 m/s). Further investigation on how navigation  
 472 and flow velocity might disrupt and cause the disintegration of water hyacinths and possibly the release of  
 473 plastics in the open water are needed to better understand water hyacinths distribution in riverine ecosystems  
 474 and plastic transport dynamics.





475

476 **Figure 9.** Water hyacinths distribution maps over part of the Saigon river. Panel 1 indicates the localization of  
 477 Ho Chi Minh City. Panel 2 shows the section of the Saigon river covered by Sentinel-2 imagery taken on the 22  
 478 May 2020. Panels 3 to 6 display Sentinel-2 imagery with the False Color band combination (near-infrared, red  
 479 and green bands). The stars indicate the approximate location of the detailed inset in each sector. Pixels with  
 480 vegetation appear in red, pixels appearing in dark blue and black indicate water. Thu Thiem bridge is also  
 481 indicated in Panel 6.

482

483 **3.4 Exploring hydrological influences on spatiotemporal plastic entrapment variability**

484 **3.4.1 Change in tidal regime**

485 The diurnal variability observed in the plastic flux (0.00 and 42.8 items/min) and concentrations (between 0.05  
 486 and 23.7 items/m<sup>2</sup> can be partly explained by a shift in the tidal regime. At 13h00 on the measurement day, the  
 487 flow transitioned from ebb (seaward) to flood (inland) current. All measurement techniques register higher plastic  
 488 concentrations and entrapment ratios in ebb than flood tide regimes (table 2), with the exception of the average  
 489 plastic concentration measured by UAV imagery. Such variations are probably driven by the considerable  
 490 decrease registered in the macroplastic flux (almost four times lower) during the flood tide. Only the plastic  
 491 concentration in the water estimated by the UAV imagery was found to be higher during the flood tide than  
 492 during ebb regime. This can be attributed to the large hyacinth/plastic accumulation hotspot observed close to  
 493 the northern riverbank, which was picked up by UAV flights, but not with the other methods (figure 4). The  
 494 curvature of the river and a lower streamflow close to the river shore could explain this observation. Overall, the  
 495 tidal regime was found to be the driving factor in determining macroplastic flux. Water hyacinths, which carry  
 496 approximately half of the floating macroplastic items observed, were found to be moving mostly during the ebb  
 497 tide.

498 **Table 2.** Plastic concentrations, entrapment ratios and macroplastic flux in ebb and flow tide flows on 23 May  
 499 2020. The macroplastic flux averages were integrated for the entire river width.

	Plastic items transport and entrapment				Plastic items in relation to water hyacinths				Water hyacinths coverage			
	Average plastic concentrations [# items/m <sup>2</sup> ]		Average total macroplastic flux [#items/min]		Average entrapment ratio [%]		Average plastic concentrations in hyacinths [#items/m <sup>2</sup> ]		Water hyacinths total area [m <sup>2</sup> ]		Water hyacinths patches [#]	
<i>Tidal regime</i>	<i>Ebb</i>	<i>Flood</i>	<i>Ebb</i>	<i>Flood</i>	<i>Ebb</i>	<i>Flood</i>	<i>Ebb</i>	<i>Flood</i>	<i>Ebb</i>	<i>Flood</i>	<i>Ebb</i>	<i>Flood</i>
<b>Visual counting</b>	N/A		138	48.6	53.2	10.7	N/A		N/A		N/A	
<b>UAV imagery</b>	0.467	0.621	N/A	N/A	51.9	27.1	2.16	1.97	84.3	26.3	N/A	
<b>Bridge imagery</b>	0.124	0.018	N/A	N/A	54.6	3.57	0.753	0.312	301	3.27	357	12.0

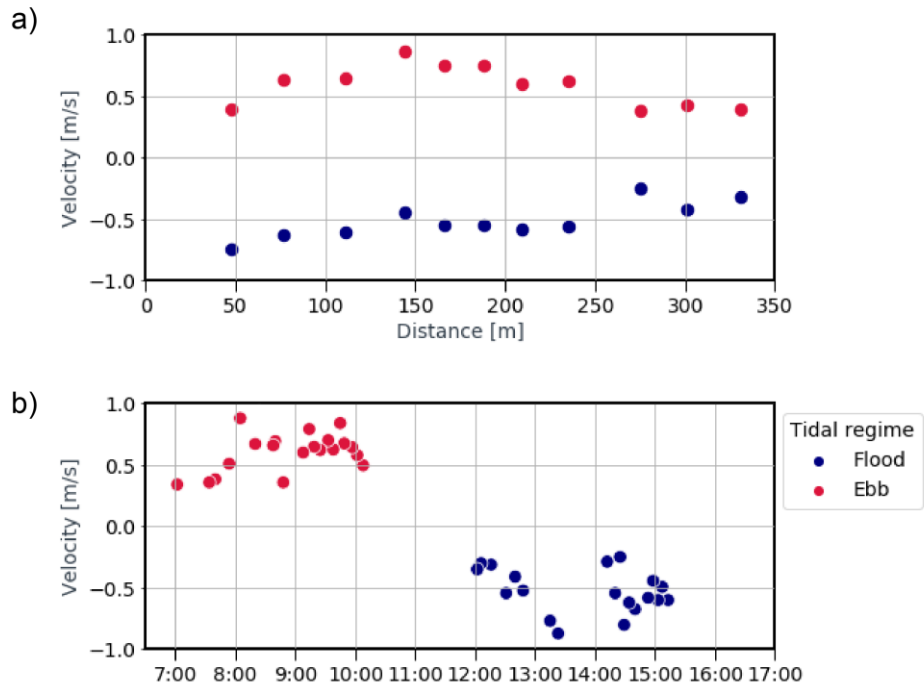
500

501 **3.4.2 River flow velocity**

502 The flow velocity varies between 1.05 m/s and 0.11 m/s for ebb tide and -0.87 m/s and -0.25 m/s for flood tide.  
 503 The variations are noticeable both across the cross-section and throughout the day (figure 10). The mean  
 504 streamflow velocity does not, however, show important variations between the ebb and flood tides (mean  
 505 velocity for the ebb tide = 0.57 m/s, for the flood tide = - 0.53 m/s). Macroplastic fluxes for both entangled and  
 506 free-floating items were weakly correlated with stream flow velocity (for both entangled and free-floating flux:

507 Pearson's  $r = 0.30$ ,  $p = 0.051$ ). The water hyacinth area estimated from the bridge imagery was also not related  
 508 to stream flow velocity (Pearson's  $r = -0.22$ ,  $p > 0.05$ ). The distribution of macroplastic flux along the river width  
 509 does not show any significant relationship with the average stream flow (both ebb and flood tides) at the  
 510 observation points nor with the presence of water hyacinth patches (all  $p$ - values  $> 0.05$ ). Our observations  
 511 indicate that flow velocity is of minor importance for the plastic flux and water hyacinths abundance. These  
 512 findings support those of van Emmerik, Strady et al. (2019), conducted over 10-month, which also did not find  
 513 a clear relation between flow velocity and macroplastic fluxes.

514



515 **Figure 10.** Streamflow velocity of the Saigon river close to Thu Thiem bridge. a) Average flow velocity along  
 516 the river width [distance from the northern river bank] b) Flow velocity throughout the day. Negative values  
 517 indicate stream flow in flood tidal regime, positive in ebb tide.  
 518

519 **4 Which method to choose?**




































520 The results from the field campaign showed that several measuring methods can be mobilized to provide a  
 521 comprehensive overview on different aspects regarding the role of water hyacinths in macroplastic transport  
 522 and entrapment. To determine what (combination of) measuring methods to choose to design future monitoring  
 523 campaign, we provide a comprehensive overview of aspects to consider for such strategies. We emphasize that  
 524 there is no single best method, and that the choice in measuring methods ultimately depends on the monitoring  
 525 goals, the local context and the available resources. To guide practitioners in this choice, we compared the  
 526 measuring techniques based on three set of criteria: (1) the suitability for each technique to derive metrics of  
 527 interest, (2) their relevancy at different spatiotemporal scales, (3) and their overall strengths, weaknesses,  
 528 opportunities and threats (SWOT).

529 **4.1 Suitability of each technique based on the retrieval of metrics of interest for riverine**  
 530 **macroplastics**

531 Figure 11 presents the metrics (green marks) that can be derived from various monitoring techniques to  
 532 characterize the contribution of water hyacinths in transporting floating macroplastic. Orange marks indicate  
 533 method-metric combinations where either it is uncertain if direct measurement is feasible, or when the metric  
 534 may be retrieved from a combination of direct measurement and secondary sources. For instance, in the case  
 535 of the use of camera, the plastic categorization depends on the height and distance at which the photographs  
 536 are taken.

537 For some specific metrics, multiple methods are available, allowing for direct comparison of the results. That is  
 538 the case for the ratio of entangled items in relation to the total plastic items found. However, a few metrics can  
 539 only be retrieved with a specific monitoring technique. The visual counting is to date the only technique that  
 540 enables to measure plastic transport flux. This metric is crucial to ultimately estimate plastic emissions into the  
 541 ocean and thus cannot be neglected. Similarly, physical sampling is the only technique that can estimate the  
 542 mass of items, a metric necessary for deriving plastic mass balance. All metrics have their benefits in providing  
 543 insights on water hyacinths abundance, plastic transport and/or entrapment within hyacinths or the  
 544 characteristics of plastics.

545 The matrix also indicates a tentative level of accuracy, based on the results from the field measurements  
 546 previously presented. Future uses of these measuring technique might alter this first assessment of accuracy  
 547 levels. Overall, the physical sampling appears as the most accurate method. However, the retrieval of water  
 548 hyacinths mass holds some uncertainties depending on whether wet or dry mass is of interest.

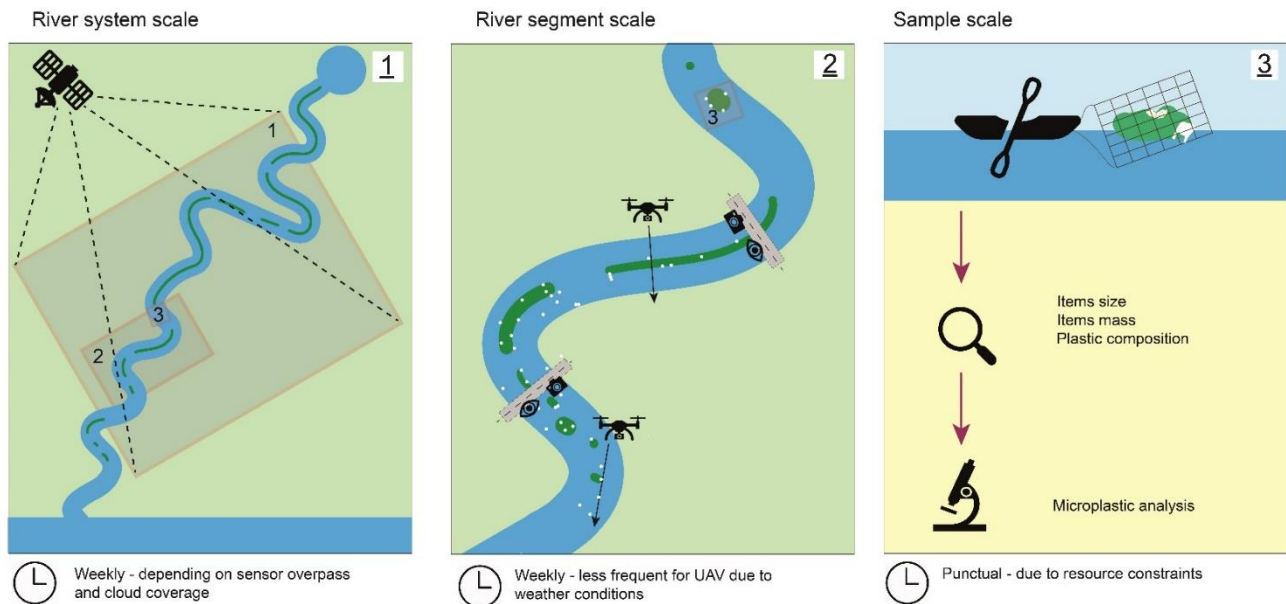
	Plastic transport and entrapment in hyacinths				Plastic items characteristics			Water hyacinths spatial distribution			Accuracy level	
	Macro-plastic flux	Entangle-ment ratio	Items concentration in water	Items concentration in hyacinths	Plastic categories	Size distribution	Items mass	Number of patches	Water hyacinths area	Water hyacinths mass		
Satellite imagery 												 High
Visual counting 												 Medium
Bridge imagery 												 Moderate
UAV imagery 												 Moderate
Physical sampling 												 High

549 **Figure 11.** Matrix comparing measuring techniques based on key metrics of interest for detecting riverine  
 550 macroplastic.  
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552 **4.2 Relevancy of each technique at different spatiotemporal scales**

553 In addition, we compare the measuring techniques based on their spatial coverage and resolution, as well as  
 554 temporal frequency (figure 12). The assessed techniques are able to characterize the contribution of water  
 555 hyacinths in floating macroplastic transport and entrapment at three different spatial scales. Firstly, satellite  
 556 imagery can characterize the spatial distribution and seasonality trends in water hyacinths coverage for a river  
 557 system. Certain sensors, such as Sentinel-1 and 2, enable to cover large areas simultaneously (the wide swaths  
 558 of Sentinel-2 is of 290 km, and 250 km for Sentinel-1). Mosaicking several satellite imagery scenes taken the  
 559 same day might allow to monitor an entire river system. The exact spatial coverage and resolution depend on  
 560 the sensor characteristics but are typically within the order of 1-10 m/ per pixel. Secondly, UAV images, visual  
 561 counting and bridge imagery can characterize floating macroplastic flux and entrapment at a few river segments.  
 562 The imagery techniques can be used for finer characterization of the presence of hyacinths, since smaller  
 563 individual hyacinths can be seen, in complement of the satellite imagery. The spatial resolution in UAV imagery  
 564 is in the order 0.1 cm/pixel with a flight elevation of approximately 5 m, which enables to detect the plastic  
 565 category of items and to estimate items size. For imagery taken from a bridge at approximately 12-16 m, the  
 566 resolution is approximately of 0.3 cm/pixel, allowing to detect the items, but not their category.

567 With these methods, a few river segments can be covered, depending on the presence of infrastructure present  
 568 (road, bridge) that guarantees access to the sites of interest. Additionally, UAVs could be used to survey areas  
 569 that are not accessible via these infrastructures. Thirdly, the physical sampling can provide insights on the  
 570 specific composition of a sample: it provides high accuracy in determining the plastic composition and can be  
 571 used to derive the mass of items as well as further microplastic analysis.



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**Figure 12.** Schematic overview of the various spatial scales and temporal frequency at which the measurements can be done.

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Due to the complex nature of the plastic-water hyacinth interactions, multiple perspectives are needed to understand the transport and entrapment processes, and variability of plastic and water hyacinths variables at different spatiotemporal scales. A nested monitoring framework is useful for large-scale river monitoring, while maintaining local relevance.

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The temporal frequency and coverage can also play a role in evaluating the relevancy of each method. The satellite imagery can combine both high frequency and long temporal coverage. Due to the frequent revisit time (1 or 2 images per week) and global coverage since their launch dates in 2015 and 2017, the Sentinel-2 constellation holds the potential for mapping water hyacinth distribution over several years. Ultimately, these satellites can help characterizing interannual and seasonality trends in hyacinths coverage. For all other monitoring techniques, the frequency in measurements is solely dependent on resources. However, a weekly frequency in measurements for the visual counting and the imagery techniques seems appropriate due to the fact that both macroplastic transport and hyacinths presence are highly dynamics in time. Lastly, physical samples are typically punctual measurements, due to the resource constraints in both the sample retrieval and then the subsequent analysis.

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### 4.3 Benefits and drawbacks of each technique (SWOT analysis)











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Other factors than the metrics of interest and spatiotemporal dimensions can enter into consideration when defining a monitoring strategy. The shortcomings of each technique, for instance, as well as new opportunities that arise in their use thanks to technological developments. We present a SWOT analysis in figure 13 that summarizes these elements for each measuring technique. Considerations pertaining to the cost-efficiency, processing requirements, labor and time, technical expertise and environmental conditions were included. For instance, the visual counting technique is relatively low-cost and involves few processing steps in data analysis. The satellite imagery also comes at low cost for the users and can be used without direct access to the river of interest. The physical sampling and UAV imagery are more time-consuming and demanding in terms of resources and equipment, but enable a more detailed and accurate characterization of plastic items.

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In terms of opportunities, two very promising future development in machine learning and multispectral/hyperspectral imagery could have major operational implications for monitoring the contribution of vegetation to floating macroplastic transport. Although current machine learning applications to automatically detect floating plastic objects are on the rise, their overall use remains exploratory, because they typically require large training datasets (Lieshout et al., 2020). In addition, the presence of water hyacinths patches adds complexity for object detection and image processing, because the plastic items are less easily detectable when entrapped. Future improvements in machine learning and object detection algorithms could imply faster processing of imagery datasets, thus greatly facilitating the extensive use of UAV and bridge imagery. Similarly,

607 the development of multispectral or hyperspectral sensors for satellite imagery or UAV technology that detect  
 608 plastic components would also justify the increased use of these techniques (Tasseron et al., 2021). In particular,  
 609 if the detection of plastic/hyacinths mixture by satellite imagery would be ascertained, the role of satellite imagery  
 610 would become even more central, given its large geographical coverage, high temporal resolution and cost-  
 611 effectiveness. The successful detection of floating debris in coastal waters using satellite imagery is promising  
 612 for future applications in riverine systems (Biermann et al., 2020).

		Internal factors	
		STRENGTHS	WEAKNESSES
Satellite imagery		S1 Covers large geographical areas S2 High temporal resolution (observations every 5 days) S3 Does not require surveyors on the ground S4 Low cost technology, possible use of open source imagery and software	W1 Spatial resolution of 10 m x 10 m, smaller patches might be missed W2 Cannot distinguish plastic items W2 Requires ground truth data for classification validation / accuracy assessment
Visual counting		S1 Only measurement technique that enables to estimate floating macroplastic fluxes S2 Few processing steps involved, rapid analysis S3 Can be coupled with the collection of other metrics of interest (water elevation, flow velocity, tidal characteristics)	W1 Requires availability of trained surveyors W2 Requires presence of a bridge safe enough for surveyors to stand and conduct the measurements W3 Observer bias
Bridge imagery		S1 Enables to characterize floating macroplastic accumulation and hyacinth coverage across the river width S2 Can be conducted at the same time than visual counting, direct complement of the flux measurements	W1 Manual labelling of items is time-consuming W2 Bridge elevation generally too high to distinguish plastic categories W3 Requires presence of a bridge safe enough for surveyors to stand and conduct the measurements W4 Risk of blurry images if no stabilization device is used
UAV imagery		S1 Enables to characterize floating macroplastic accumulation and hyacinth coverage across the river width S2 Can be used to characterize plastic categories S3 Can enable to survey areas difficult to reach	W1 Can only cover a few river segments W2 Manual labelling of items is time-consuming W3 Dependency on batteries
Physical sampling		S1 High accuracy for plastic categorization S2 Enable to establish plastic mass balance S3 Can be used to integrate analysis on microplastic S4 Provide insights on submerged macroplastics entangled in hyacinth roots	W1 Requires the availability of a boat W2 Analysis of sampled items is time-consuming and requires access to lab facilities W3 Provides only 'snapshot' views
		External factors	
		OPPORTUNITIES	THREATS
Satellite imagery		O1 Fast development in unmixing pixels techniques and data fusion between medium-resolution and high-resolution sensors O2 Fast development in new sensors that might capture plastic accumulations	T1 Hampered by cloud conditions for optical imagery T2 End of satellite mission programs
Visual counting		O1 Enable comparisons among different river systems if other similar measurements are conducted elsewhere	T1 Constrained by weather conditions T2 Flow and tidal conditions can affect in-situ measurements
Bridge imagery		O1 Fast technological process in object detection with machine learning O2 Development of camera sensors that can automatically detect vegetation and floating debris O3 Improved image processing techniques available	T1 Constrained by weather conditions T2 Relatively high cost of new sensors, lack of funding T3 Risk of theft for camera/video sensors mounted on a bridge
UAV imagery		O1 Fast technological process in object detection with machine learning O2 Development of camera sensors that can automatically detect vegetation and floating debris O3 Improved image processing techniques available	T1 Heavily constrained by weather conditions T2 Flying over certain areas might be prohibited T3 Relatively high cost of new sensors, lack of funding
Physical sampling		O1 Can be used for further microplastic studies O2 Well accepted method among plastic studies	T1 Flow and tidal conditions can affect in-situ measurements

613

614

Figure 13. SWOT analysis for each measuring technique tested.

615

## 5 Concluding remarks

616 In this paper, we present five monitoring techniques to assess the role of water hyacinths in transporting and  
 617 accumulating macroplastic. We provided a field guide on how to use monitoring methods to characterize water  
 618 hyacinths distribution, plastic transport, the relation between water hyacinths and plastic, as well as plastic items  
 619 characteristics. The field guide was built around in-situ surveys conducted at the Saigon river, Vietnam as well  
 620 as satellite imagery analysis. There, we tested five different methods commonly used in plastic and vegetation  
 621 monitoring in riverine ecosystems. The practical set-up of these monitoring techniques was adapted to our scope  
 622 of investigating the role of floating vegetation in plastic propagation and provided details facilitate future use.  
 623 Further, the field guide presented and compared the results derived from each monitoring techniques, which  
 624 illustrate the various aspects that can be considered when investigating plastic entrapment in vegetation.

625 Our framework for observing vegetation entangled riverine macroplastic enables to compare the suitability of  
 626 measuring techniques and can help to inform and optimize future measuring initiatives. All monitoring techniques  
 627 can characterize different aspects of the role of floating vegetation in plastic transport and entrapment. Although  
 628 there is also overlap in the aspects that the different techniques cover, certain essential metrics can only be  
 629 derived by a specific method (flux measurements by visual counting and plastic mass balance by physical  
 630 sampling). We also present the different spatiotemporal scales and resolutions that each measuring technique  
 631 can contribute to, an aspect particularly relevant for long-term monitoring efforts at a large scale. We emphasize  
 632 the importance of combining several techniques which are complementary. Satellite imagery analysis can  
 633 provide a thorough overview at the river system scale of the vegetation abundance and its spatial and seasonal  
 634 dynamics, thanks to the large geographical and temporal coverage it offers. UAV and bridge imagery can  
 635 complement the visual counting observations, by providing additional insights on water hyacinth patches and  
 636 their characteristics (number, area, accumulation zones), as well as the size, the typology and concentrations  
 637 of items found. Further, the SWOT analysis summarized the main benefits and drawbacks pertaining to each  
 638 measuring techniques and practical considerations on the cost, labor and equipment requirements might be  
 639 useful to practitioners in designing monitoring strategies. The SWOT analysis also highlighted that future  
 640 technological advancements in sensor characteristics and machine learning could substantially foster the use  
 641 of satellite imagery and in-situ imagery techniques.

642 With this field guide we provide a step towards additional monitoring efforts on the role of vegetation in plastic  
 643 transport at a river system scale. The majority of tropical and subtropical rivers report an increasing presence of  
 644 water hyacinths that disrupts navigation activities and negatively impact aquatic ecosystems. Large-scale and  
 645 extended monitoring of this floating aquatic weed, spread and distribution as well as their linkages with plastic  
 646 retention mechanisms is thus of global interest. Better characterizing the nexus between plastic and vegetation  
 647 is crucial to improve our understanding of plastic transport processes and dynamics. Hyacinths and other types  
 648 of floating vegetation are easily detectable by satellite imagery and it can hypothesized that they trap and carry  
 649 large shares of floating debris for other river systems. If this is ascertained, quantifying the coverage of floating  
 650 vegetation could be used as a proxy for estimating macroplastic quantities in rivers. Lastly, such monitoring  
 651 efforts could inform the operational clean-up strategies, for instance by testing the co-removal of plastic and  
 652 hyacinths.

### 653 **Author Contributions**

654 TvE, LS and MvdP: design and conceptualization of the study. TLN, NP and TK: data collection. LS and EV:  
 655 data analysis. LS: writing. MvdP, TvE, ES, LB, TK, TLB, SK, SvdB: editing and reviewing.

### 656 **Funding**

657 The work of LS was supported by NWO Open Mind grant 18127. The work of MvdP and TvE was supported by  
 658 the 4TU.Federation Plantenna project. TvE is supported by the Veni research program The River Plastic  
 659 Monitoring Project with project number 18211, which is (partly) funded by the Dutch Research Council (NWO).  
 660 This study is part of the Plastic Plants project, supported by the Discovery Element of the European Space  
 661 Agency's Basic Activities (ESA contract no. 4000132682/20/NL/GLC).

### 662 **Conflict of Interest**

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

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