

1 **Rivers of waste: anthropogenic litter in intermittent Sardinian rivers, Italy**
2 **(Central Mediterranean)**

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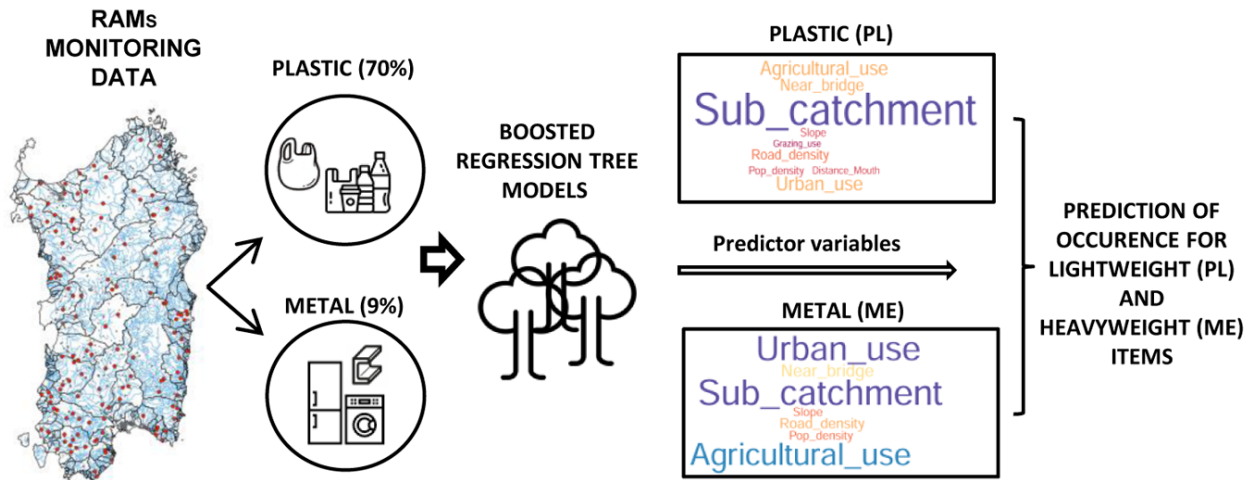
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14 **Graphical abstract**

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19 **Abstract**

20 While the increasing accumulation of anthropogenic litter in the marine environment has
21 received considerable attention over the last decade, litter occurrence and distribution in
22 rivers, the main source of marine litter, have been comparatively less investigated.
23 Moreover, little information is available about the amount and typology of Riverine
24 Anthropogenic Macro-litter (RAM) entering marine environments from intermittent rivers
25 in low populated areas of the Mediterranean basin. To provide insights on this issue, we
26 investigated density and composition of RAM accumulated over a total of 133 riverbanks,
27 belonging to 37 river basins in the Sardinia Island (Mediterranean Sea). We report here that
28 plastics, especially single-use items, represent the most frequent and abundant RAM
29 category in all investigated basins. Statistical modelling revealed that occurrence of
30 lightweight RAM (especially plastic) is mostly explained by levels of urban (12.3% of the
31 relative contribution) and agricultural (12%) land use of the territory, whereas the proximity
32 of bridges to the sampling point (21%) and the local population density (19.8%) are best
33 predictors of heavy weighted RAM items (i.e., large metal items, appliances) occurrence.
34 Our results confirm that plastics represent an important component of RAM and pinpoint
35 that, beside plastic reduction policies and better waste management, actions aimed at
36 abating and monitoring litter contamination should be localized on the proximity of
37 bridges, whatever the local population density. Finally, to fill existing knowledge gaps in
38 understanding the severity of litter discharge and accumulation in the Mediterranean Sea,
39 land-to-sea systematic monitoring campaigns at appropriate spatial and temporal scales
40 should be put in place.

41

42 **Keywords:** intermittent rivers; macro-litter, plastic, Non-linear Boosted Regression Tree
43 Models

44

45 **Introduction**

46 Anthropogenic litter in aquatic environments is an emerging issue of global concern due to
47 its negative impacts over the different hierarchical levels of ecological organization which,
48 ultimately, have also socio-economic consequences (Conchubhair et al., 2019; Kühn et al.,
49 2015; Newman et al., 2015; Rochman et al., 2016). In the last few decades, the focus of
50 investigation on the accumulation of anthropogenic litter (mainly plastic) has been heavily
51 skewed towards the marine environment (Galgani et al., 2015), despite the fact that rivers
52 represent the most important conduits for the transportation of anthropogenic litter to the
53 marine environment (Blettler et al., 2018; Jambeck et al., 2015; Schmidt et al., 2017).

54 It has been postulated that few very large rivers could be the major sources of plastic
55 contamination of the oceanic contamination (Lebreton et al., 2017; Schmidt et al., 2017).
56 However, a recent modelling study, based on field observation, revealed that about 1000
57 rivers located in highly populated areas and mainly distributed in Asian countries can
58 cumulatively discharge annually between 0.8 and 2.7 million metric tons of macro-plastic
59 (i.e. plastic debris >5 mm) into the global oceans, which account for >80% of riverine
60 annual plastic emissions to the sea (Meijer et al., 2021).

61 Nevertheless, despite the current literature estimates, the real quantification of total
62 plastic transport from land to seas remains still largely uncertain because of operational
63 difficulties to obtain *in situ* measurements and due to the lack of standard observation
64 techniques (Broere et al., 2021; Edelson et al., 2021; van Emmerik and Schwarz, 2020;
65 Weiss et al., 2021). Intermittent rivers and ephemeral streams are common across Europe
66 and dominate river networks in Mediterranean regions (Skoulikidis et al., 2017;
67 Stubbington et al., 2018). As being ecosystems with unpredictably temporal dynamics of
68 water supply, the role of intermittent rivers in RAM transport to the sea has been almost
69 entirely ignored (Table S1).

70 Riverine Anthropogenic Macro-litter (RAM) refers to the fraction of solid waste (>5
71 mm) present in rivers and on riverbanks (González-Fernández et al., 2021; Schmidt et al.,
72 2017). Anthropogenic Macro-litter originates from mismanagement of urban waste, sewage
73 outlets from wastewater treatment plants, illegal dumping, loss of products from industrial

74 and agricultural activities (Bruge et al., 2018; Faure et al., 2015; Galafassi et al., 2019;
75 Kiessling et al., 2019).

76 Plastic debris are considered dominant in riverine contamination, whereas other
77 materials such as glass, metal are relatively minor contributors due to the intrinsic features
78 of the materials (e.g., specific weight, buoyancy) that likely prevent them to be dislocated
79 via river flow (Castro-Jiménez et al., 2019; Cesarini and Scalici, 2022; González-Fernández
80 et al., 2021; Rech et al., 2014). Besides the aesthetic, ethical and socio-economic damage
81 (Rochman et al., 2016; Williams and Simmons, 1996), the contamination of plastic in
82 riverine ecosystems can cause numerous negative consequences for biota and environment.
83 Plastic debris can potentially degrade into microplastics that could be ingested by aquatic
84 organisms, such as fishes and zooplankton (Galafassi et al., 2021; Rehse et al., 2018).
85 Moreover the presence and accumulation of the trapped plastic in riparian vegetation can
86 cause negative effects on the plant status and survival of trees (van Bijsterveldt et al.,
87 2021).

88 Abundance, composition and distribution of RAM are influenced by the cumulative
89 effects of an array of environmental characteristics, including: i) the presence of floating
90 (e.g., hyacinth) (Schreyers et al., 2021) and riparian vegetation (e.g., arboreal, reeds, etc.)
91 that act as a barrier, especially during flooding events (Cesarini and Scalici, 2022; Williams
92 and Simmons, 1996; Windsor et al., 2019); ii) tidal influence, iii) seasonal changes of the
93 water level, iv) flow rate (Battulga et al., 2019; Vriend et al., 2020b), and v) curvature and
94 shape of the river (Calcar and van Emmerik, 2019).

95 Moreover, RAM accumulation can be also influenced by several anthropogenic
96 pressures like land use (Cowger et al., 2019; McCormick and Hoellein, 2016), shipping
97 activities and the presence of fluvial infrastructures (irrigation and drainage channels,
98 wastewater treatment plants, dams and bridges) (Calcar and van Emmerik, 2019; Mihai,
99 2018; Schirinzi et al., 2020; Simon-Sánchez et al., 2019), human population and road
100 density (Battulga et al., 2019; Jambeck et al., 2015; McCormick and Hoellein, 2016).

101 The Mediterranean Sea is one of the most important accumulation zones of marine
102 litter worldwide (Cózar et al., 2015; Eriksen et al., 2014; Suaria et al., 2016). Information

103 about RAM inputs from permanent rivers associated with highly populated areas in the
104 Mediterranean Sea is available (e.g. Crosti *et al.*, 2018; Castro-Jiménez *et al.*, 2019;
105 Schirinzi *et al.*, 2020; Cesarini and Scalici, 2022), but, to date, information about RAM
106 from intermittent rivers is still almost absent (Table S1).

107 To address this knowledge gaps, by contending that RAM's role could be more
108 important than previously thought or hypothesized, this study aims: (1) to assess density
109 and composition of RAM in riverbanks of intermittent rivers in Sardinia; (2) to determine
110 the main factors affecting the occurrence, composition, and distribution of RAM.

111

112

113 **Materials and Methods**

114 *Study area*

115 The study area is in Sardinia (Italy), the second largest island (ca. 24,106 km²) in the
116 Mediterranean Sea. Sardinia, with a population of 1,630,474, corresponding to a density of
117 nearly 67.7 inhabitants km⁻², is one of the less densely populated regions in Italy.

118 Sardinia is characterized by complex topography with the presence of a long mountain
119 range (Sardinian-Corse Mountain System) that influence the local circulation and spatial
120 distribution of the rainfall (Marras et al., 2021). Bi-seasonal climatic features, with hot arid
121 summers, rainy autumn/winter seasons along with extreme precipitation events, determine
122 irregular flow and strong seasonal hydrological fluctuations (De Waele et al., 2010; Palmas
123 et al., 2020; Podda et al., 2020; Sabatini et al., 2018). The hydrographic network is
124 characterized by the dominance of non-perennial rivers (90% of the total) (Skoulikidis et
125 al., 2017). The recurrent temporal overlap of the dry season with a high water demand for
126 agriculture irrigation, industry and domestic purposes have led the construction of a total of
127 54 larger dams (Marchetto et al., 2009; Montaldo and Sarigu, 2017), that, interrupt the
128 continuity of perennial rivers (Tirso, Flumendosa and Coghinas), strongly influencing their
129 natural hydrological cycle (Moccia et al., 2020; Naselli-Flores et al., 2014). According to
130 available data provided by Autonomous Region of Sardinia, all dams have both surface and
131 bottom-discharge systems and an average height of 42.5 ± 3 m. While punctual data on the
132 volume of discharge per year was not available, it was possible to assert that the main
133 discharge system is from surface with an average of 884.5 ± 130 m³ sec⁻¹.

134 Furthermore, in Sardinia it has been predicted that the future reduction of mean
135 precipitation due to global warming, may further exacerbate droughts with a strong
136 decrease of mean runoff (Marras et al., 2021). For all reasons mentioned above, we
137 considered all investigated rivers as intermittent.

138

139 *Anthropogenic macrolitter occurrence, abundance, and composition*

140 Litter monitoring was conducted over a total of 133 sampling sites (belonging to 37 river
141 basins), covering different altitudinal zones and environmental conditions across 2018 and

142 2020 (Figure 1). Litter occurrence on the riverbanks was determined according to the
143 Rivers-OSPAR protocol (van Emmerik et al., 2020a) that was based on the OSPAR beach
144 litter guidelines (OSPAR, 2010). Sampling was carried out on both riverbanks of each river
145 and data of each river were cumulated. The methodological approach was based on the
146 count of macro-litter items (>5 mm) on 100 m long stretches of riverbanks parallel to the
147 waterline, from the waterline itself to the maximum level of floodplain landward (Figure
148 S1). All visible RAM items, deposited in the riverbanks and/or entrapped in the vegetation,
149 were counted within the entire sampling area and, for each station, litter density was
150 calculated as the number of items per kilometer of riverbank (items km⁻¹).

151 The items were collected and sorted according to the UNEP-Code master list
152 classification for beach litter items (Cheshire et al., 2009; Galgani et al., 2018). The list
153 comprised 128 sub-categories grouped into nine anthropogenic litter materials: artificial
154 polymer materials (PL), rubber (RB), cloth/textile (CL), paper/cardboard (PC),
155 processed/worked wood (WD), metal (ME), glass and ceramics (GC), other materials (OT),
156 undefined (UN).

157

158 ***Factors affecting the occurrence, composition, and distribution of RAM***

159 To determine factors affecting the occurrence, composition and distribution of RAM we
160 first identified from the literature an array of eleven potential variables, grouped into three
161 categories (geomorphology, land use and human pressure; Table S2), assumed to mostly
162 influence the occurrence of litter items (presence/absence).

163 Geo-morphological variables include: i) the sub-catchment area (km²) above the
164 sampling site as a proxy of catchment runoff; ii) the river order as a proxy of
165 upstream–downstream gradients (according to Strahler method's (Strahler, 1957); iii) the
166 stream slope as a proxy of potential water velocities. Season of sampling was also used as a
167 proxy of river discharge events, considering the peculiar abovementioned climatic features
168 of Sardinia. Land use data, obtained from the CORINE database, were merged in four
169 categories: natural use (which includes forests and semi-natural areas, among others, but
170 excludes recreational use of the territory), agricultural use, grazing use and urban use. Land

171 cover was expressed as the percentage (%) of each of these categories in the sub-catchment
172 area above each sampling site.

173 Human pressure proxies were estimated in terms of: i) road density (km km^{-2}), ii)
174 population density ($\text{population km}^{-2}$), iii) the presence/absence of river bridges immediately
175 above the sampling point. When bridges were present, the transects started at the bridge,
176 moving downstream from there. In addition, since the presence of weirs and dams could
177 negatively influence the presence and transport waste items in the riverbanks, the number
178 of dams above the sampling station were also considered.

179 Georeferenced datasets on hydrographic data, roadways, larger dams, and land use
180 were acquired from the Regional Land Information System of Sardinia.

181

182 *Statistical analyses*

183 Density from each station was then used to generate distribution maps of the most
184 important litter categories. The free Quantum GIS Desktop, version 2.18.3 (QGIS)
185 (<http://www.qgis.org/>) software was used for creating distribution maps and to extract the
186 exploratory variables. Sub-catchment area, slope as well as the stream order of each
187 sampling site were calculated based on 10-m resolution Digital Elevation Model (DEM).
188 For the entire river network generated by flow accumulation, stream order was derived with
189 the Strahler method's (Strahler, 1957).

190 Since the artificial polymers and metal materials (PL and ME, respectively) were the
191 most abundant anthropogenic litter material (cumulatively accounting for ca. 80% of total
192 litter), we tested whether and to which extent some potential explanatory variables were
193 putative drivers for PL and ME litter occurrence using a non-linear Boosted Regression Tree
194 model (BRT, Elith *et al.*, 2008). As we aimed at identifying the conditions that might
195 represent a threshold over which light (PL) and heavy materials (ME) could be found, we
196 used only presence/absence transformed data. BRT models have been used to analyse the
197 relationships between response and predictor variables in different fields of environmental
198 science (Ju *et al.*, 2021; Lagarde *et al.*, 2021; Lemm *et al.*, 2021; Saha *et al.*, 2021). BRT

199 models allow testing different types of predictive variables by fitting complex non-linear
200 relationships and handling interaction effect between predictors, while not depending on the
201 normality and homoscedasticity of the data (Déath, 2007; Elith et al., 2008).

202 To fit the BRT models, the learning rate (the importance of each iteration in the model)
203 and tree complexity were set through an iterative process to ensure that the final model
204 outcome consisted of at least 1000 decision trees (Elith *et al.*, 2008). The relative
205 importance of each predictor variable has been also calculated from the BRT model and
206 was visualised in partial dependence plots. BRT models were run with a Bernoulli link
207 function. The BTRs' performance was evaluated by the amount of total deviance explained
208 (DEV %) and by cross-validated correlation between model prediction and observed data
209 (R^2 of CV) (Derville et al., 2016; Ju et al., 2021; Nieto and Mélin, 2017; Saha et al., 2021).
210 The predictive performance of the BTRs were also tested and evaluated using the
211 threshold-independent Receiver-Operating Characteristic (ROC) curve and the estimation
212 of the area under ROC plot (AUC) (Amorim et al., 2016; Derville et al., 2016; Saha et al.,
213 2021; Wang et al., 2021). Collinearity among covariates was tested by computing pairwise
214 scatter plots among covariates. Covariates showing relevant Spearman's Rho ($\rho > 0.7$) were
215 discarded from the modelling. The Variance Inflation Factor (VIF) was also used to check
216 collinearities among explanatory variables; those showing VIF >3 were also discarded from
217 the analysis (Zuur et al., 2010).

218 All analyses were carried out using the statistical software package R (R Core Team,
219 2021). BRTs are estimated using the "dismo" library (Hijmans et al., 2011). The ROC
220 analysis was performed using the R package pROC (Robin et al., 2011).

221

222 **Results**

223 A total of 2078 RAM items were collected from the 37 river basins, covering ca. 22 linear
224 km of riverbanks. Out of 133 sampling stations, 114 (85.7%) showed the presence of litter
225 items, and only 19 were litter free. Overall, 28 sub-categories of litter items were found on
226 Sardinian rivers, even if the top 5 most abundant types of items represented most of the
227 litter found (~70%, Table S3). Artificial polymers materials (PL, 70.4%) were the most

228 abundant category, followed by Metal (ME, 9.3%), Cloth/Textile (CL, 8.5%) and Glass &
229 Ceramic (GC, 7.4%) (Figure 2a). Other materials (including rubber; RB), paper; PE, and
230 processed wood; WD) represented cumulatively 4.4% of all the litter. Artificial polymers
231 items consisted mostly of single-use plastic items such as bags (PL07, 60%), bottles caps
232 and lids (PL01, 16%) and small bottles (PL02, 11%) (Figure 2b). Metal dominant
233 categories were equally cans (ME03; 37%) and metallic objects larger than 50cm (ME10;
234 38%).

235 Overall, the mean litter density for all investigated riverbanks was 156 ± 19 items km^{-1}
236 (median value of 90 items km^{-1}). The highest litter mean density was measured in the river
237 basins of Flumini Mannu (R2_1) (393 ± 100 items km^{-1}), Mannu di San Sperate (R2_2)
238 (386 ± 168 items km^{-1}), Pelau (O3) (335 ± 335 items km^{-1}) and Flumendosa (P_1) (318 ± 263
239 items km^{-1}) (Figure 2c). With the exception of Pelau river basin (O3), where the highest
240 litter densities were measured at the most upstream location and mainly composed of glass
241 and ceramic items (GC) (670 items km^{-1}) (Figure 3a), likely too heavy to be transported
242 downstream as per other lighter materials. All other river basins were characterized by litter
243 dominated by PL, with the highest PL items densities recorded in downstream sampling
244 stations (Figure 3b).

245 Analysis of multi-collinearity among predictive variables revealed strong correlations
246 between natural use (Natural_use), agricultural use (Agricultural_use) ($\rho = -0.9$), number of
247 larger dams above the sampling station (Dams) and sub-catchment area (Sub_catchment) (ρ
248 = 0.7). After removing Natural_use and the Dams variables, the VIF values did not exceed
249 3.0.

250 The results of BTR model revealed that the presence of plastic litter (PL) is influenced
251 by the joint effect of geomorphological variables, land use and human pressures. The PL
252 model accounted for 40% of the total deviance and a CV correlation between predicted and
253 observed data of 0.70. The analysis of the relative importance of the different predictors
254 revealed that the sub-catchment area (Sub_catchment) (34.5%), urban use (Urban_use)
255 (12.3%), agricultural use (Agricultural_use) (12.0%), the presence of a bridge above the
256 sampling point (Near_bridge) (10.1%) and road density (Road_density) (7.5%) represented

257 the highest share in relative explained deviance (Table 1). The partial responses of single
258 predictors showed a predominantly positive linear trend with a plateau for Sub_catchment,
259 Urban_use, Road_density and Agricultural_use (Figure S2a). In particular, the
260 sub-catchment curve is steeper than that of all other predictors and reaches a peak at a
261 relatively small surface area (60 km^{-1}). The effects of land use variables were
262 approximately J-shaped, with the probability of occurrence of PL litter significantly
263 increased after 0.5% and 6.5% of coverage area for urban and agricultural use, respectively
264 (Figure S2a). The PL litter is also found most in stretches of rivers characterized by the
265 presence of a bridge. Road density concentration had a consistent positive relationship with
266 PL items' probability of occurrence (Figure S2a).

267 The BRT model applied to heavy materials (ME) explained 16% of the predicted
268 deviance and a CV correlation between predicted and observed data of 0.57. The most
269 significant predictor for ME occurrence were the sub catchment area above the sampling
270 station (22.6%) followed by the presence of a bridge (Near_bridge, 21.1%) and
271 population's density (Pop_density, 19,8%) (Table 1, Figure S2b).

272 The outcomes of the AUC of ROC curves confirmed that the machine learning (BRTs)
273 models have a good prediction capability for detecting the probability of occurrence of
274 either PL or ME (0.92 and 0.83, respectively) (Table 1).

275

276 **Discussion**

277 Our study provides evidence that a high frequency (~ 86%) of intermittent rivers in
278 Sardinia are contaminated by Riverine Anthropogenic Macro-litter (RAM). We show here
279 that RAM composition in Sardinian intermittent rivers is dominated by PL items (~70 %)
280 followed by metal (ME) and glass (GL) (9% and 7%, respectively). Our results confirm
281 previous studies worldwide which reported PL items as the globally dominant category in
282 riverbanks (Battulga et al., 2019; Bruge et al., 2018; Carpenter and Wolverton, 2017;
283 Hoellein et al., 2014; Kiessling et al., 2019; Rech et al., 2014; Schöneich-Argent et al.,
284 2020; van Emmerik et al., 2020a). Within the PL macro-category, single-use items such as
285 plastic bags (~60%) and bottle caps and/or covers (11%) are the most common items on

286 Sardinian riverbanks, as also observed in other European locations (Bernardini et al., 2020;
287 Winton et al., 2020).

288 The mean density of total litter for the whole study period is 156 items km⁻¹ (median
289 90 items km⁻¹) with a remarkable heterogeneity among different river basins, despite their
290 distance from each other. The Flumini Mannu is the most polluted basin (393± 100 items
291 km⁻¹), followed by the Mannu di San Sperate (386±168 items km⁻¹) and Flumendosa
292 (318±263 items km⁻¹). Differences in geographical locations, social and economic context
293 and sampling techniques make comparisons of RAM abundance among freshwater studies
294 very challenging (Blettler et al., 2018; van Emmerik et al., 2019b; Vriend et al., 2020a).
295 Nevertheless, considering those studies focused on riverbanks that used the same sampling
296 approach of this study, we report here that the RAMs' mean abundance in the intermittent
297 rivers in Sardinia are an order of magnitude lower than those reported for the Rhine and the
298 Meuse rivers (North Sea) (median value 2060 items km⁻¹) (van Emmerik et al., 2020a,
299 2020c). It is worth of notice that these systems are characterized by different hydrological
300 and anthropogenic factors such as discharge and population density.

301 The machine learning method (BRTs) identified the relative importance of a set of
302 different factors able to explain a significant proportions of variance observed for light
303 weight (PL) and heavy weight (ME) RAMs occurring along the Sardinian riverbanks. The
304 model applied to the PL macro-category has a relatively good explanatory power and
305 identifies, in decreasing order, the sub-catchment area surface, the urban and agricultural
306 percentage use of land and the presence/absence of bridges as the most important predictors
307 of RAM occurrence.

308 Our model predicted that the larger the sub-catchment area surface above the sampling
309 station, the more the occurrence of light (PL) RAMs can be expected. The primary role of
310 the sub-catchment area surface can be associated with the fact that most plastic waste
311 originates, generally, from land-based areas due to the littering or illegal landfill of waste
312 (Chae and An, 2018; Geyer et al., 2017). There, light (PL) RAMs may be washed away
313 from drainage areas by the additive effects of wind, heavy rainfall and floods (Bruge et al.,
314 2018; Carpenter and Wolverton, 2017; Windsor et al., 2019; Zylstra, 2013). Moreover, the

315 larger the sub-catchment area, the more riparian vegetation plays a role in litter
316 accumulation. In this regard, we report here that, for instance, plastic bags were mostly
317 found trapped in the vegetation at the riverside, which is known to act as a trap for floating
318 materials (Schöneich-Argent et al., 2020; Schreyers et al., 2021; van Emmerik and
319 Schwarz, 2020; Williams and Simmons, 1996). Moreover, given that the floodplains were,
320 sometimes, covered by dense arboreal, shrubs and herbaceous vegetation associations
321 occurrence and abundance of the smaller sized of PL items could be likely underestimated.

322 The same, though lower, explanatory power on litter occurrence is observed for the
323 agricultural and urban percentage use of the territory, the increase of which has been
324 already reported to positively affect PL RAMs' occurrence (Bruge et al., 2018; Carpenter
325 and Wolverton, 2017; Cowger et al., 2019; Glanville and Chang, 2015; Guerranti et al.,
326 2020; van der Wal et al., 2015). This effect appears particularly relevant in the subset of
327 Sardinian rivers (Flumini Mannu and Mannu di San Sperate) that run along the Campidano
328 plain, characterized by an almost entire occupancy of the territory by agricultural and dense
329 different urban uses.

330 The third explanatory factor of PL RAMs occurrence identified by our model is the
331 presence of bridges immediately above the sampling station. This result could depend on
332 the slowdown of the river flow which favors the accumulation of waste on the riverbanks
333 (Hoellein et al., 2014; Kiessling et al., 2019; Lebreton et al., 2017). Moreover, the
334 prevalence of plastic bags in those localities suggests the persistence of the incorrect
335 behavior of abandoning waste in places that, due to the landscape attractiveness of bridges,
336 makes them often used for refreshment breaks of tourists, motorists, and campers.

337 The sub-catchment area, the presence of bridges and the population density are the
338 most important predictors also of the occurrence of heavy materials (ME) in Sardinian
339 riverbanks. The highest abundance of discarded house appliances and aluminum drink cans
340 (38% and 37% of the total ME items, respectively) suggests that the illegal disposal and
341 dumping of ME are main sources of litter.

342 The 2019 report on Sardinian Urban Waste Management has estimated a total
343 production of urban waste of ~ 740 tons year⁻¹ with 454 kg/habitant/year (ARPAS, 2019).

344 Even if Sardinia is considered the second best performing region in Italy in terms of waste
345 management with a recycling rate of 73% (ARPAS, 2019), a certain fraction of the total
346 household waste generated is currently susceptible to uncontrolled disposal in unauthorized
347 landfill and river dumping. Our results reveal also that illegal landfills in proximity of
348 bridges and secondary roads are much more common in sub-catchment areas characterized
349 by higher population density. This result, again, fits with the observations made on
350 riverbanks of Chile, Wales and Romania, where the combination of illegal dumping and
351 human presence, more than the road density, have been identified as the main sources of
352 litter items occurrence (Cowger et al., 2019; Kiessling et al., 2019; Mihai, 2018; Rech et al.,
353 2014; Williams and Simmons, 1997).

354 Moreover, we cannot exclude that other factors, not included in this study, and
355 associated with hydrological (runoff, flow velocity, discharge, vegetation cover) and
356 anthropogenic factors (tourism and recreation activities, poor waste management practices
357 on land) could explain a certain portion of RAMs occurrence variance in riverbanks and
358 their potential transport to the sea (Bruge et al., 2018; Kiessling et al., 2019; Schirinzi et al.,
359 2020; Windsor et al., 2019). In this regard, heavy rain and extreme flooding events have
360 been suggested to affect the transport and accumulation of RAM items (Axelsson and van
361 Sebille, 2017; van Emmerik et al., 2020b, 2019b, 2019a). Despite the above biases, we
362 must notice here that, since we limited our analysis to macroscopic RAMs (> 5mm), the
363 potential severity of our results could rise when considering smaller size items
364 (Schöneich-Argent et al., 2020). The complex hydrological scenario that characterizes
365 Sardinian coastal marine waters (Olita et al., 2013; Palmas et al., 2017) does not allow to
366 infer about the actual linkages between the abundance of macro-litter observed across
367 marine coastlines (Alvito et al., 2018; Cau et al., 2022) and their occurrence in riverbanks,
368 since floating debris might end up well far from the source (Cózar et al., 2015;
369 González-Fernández et al., 2021).

370 Finally, it is worth of mention the higher quantities of litter items near bridges
371 indicated that these areas can act both as sinks and source of RAM. In this perspective,
372 more supportive infrastructures for the disposal of urban litter and monitoring and

373 surveillance measures should be envisaged. Surveillance of illegal waste disposal sites is
374 often complicated by the relative geographical isolation of disposal locations which hinders
375 the efficiency and cost effectiveness of intensive surveillance (Glanville and Chang, 2015;
376 Tasaki et al., 2007). Surveillance cameras aimed at detecting illegal dumping action (Yun et
377 al., 2019), could represent a powerful control method to be positioned in critical spots such
378 as bridges. Also, the identification of bridges as sink and source can represent a useful
379 insights for prioritizing mitigation actions foreseen by local authorities or cleanup activities.

380

381 **Conclusions**

382 Overall, the results of our study, though limited to a regional spatial scale, highlight that a
383 certain, not irrelevant, amount of RAM litter can accumulate along the banks of intermittent
384 rivers, confirming and posing light on the severity of riverine contamination and its
385 potential to transport waste into the sea. This issue would appear more crucial in the
386 southern Mediterranean Sea, in which most rivers have an intermittent water flow regime.

387 Finally, to fill existing knowledge gaps in understanding the severity of litter discharge
388 and accumulation, temporally and spatially replicated land-to-sea systematic monitoring
389 campaigns should be put in place.

390 We pinpoint also that further special effort should be also paid to optimize and
391 standardize protocols of identification, characterization and quantification of RAMS in
392 different environments (Bernardini *et al.*, 2020; van Emmerik *et al.*, 2020b; Vriend *et al.*,
393 2020b).

394

395

396 **Funding:** The research was a part of the Regional Fish Inventory supported by the Regione
397 Sardegna (Grant No. 27002-1 18/12/2015 and No 7304-12 29/03/2018).

398

399

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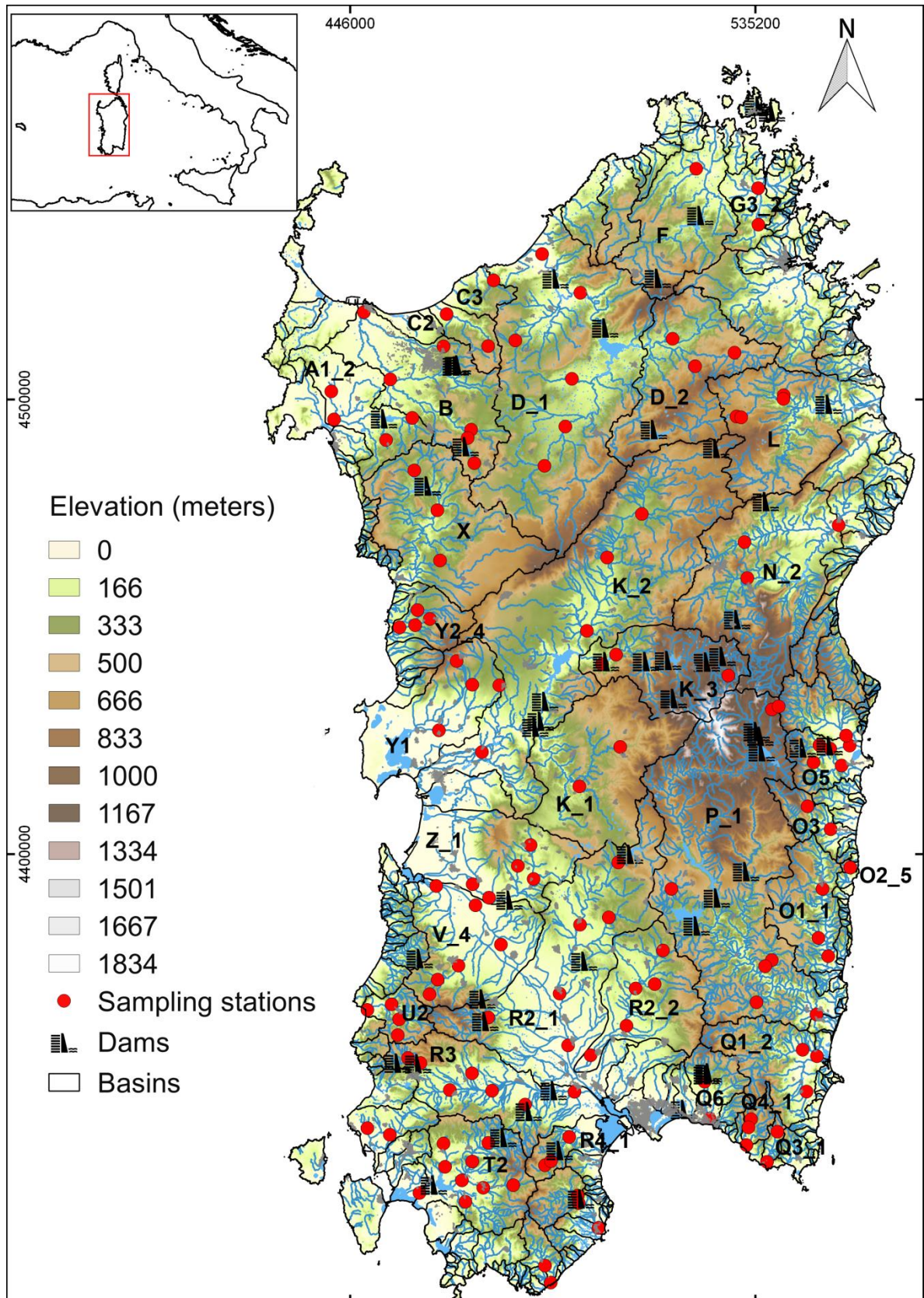
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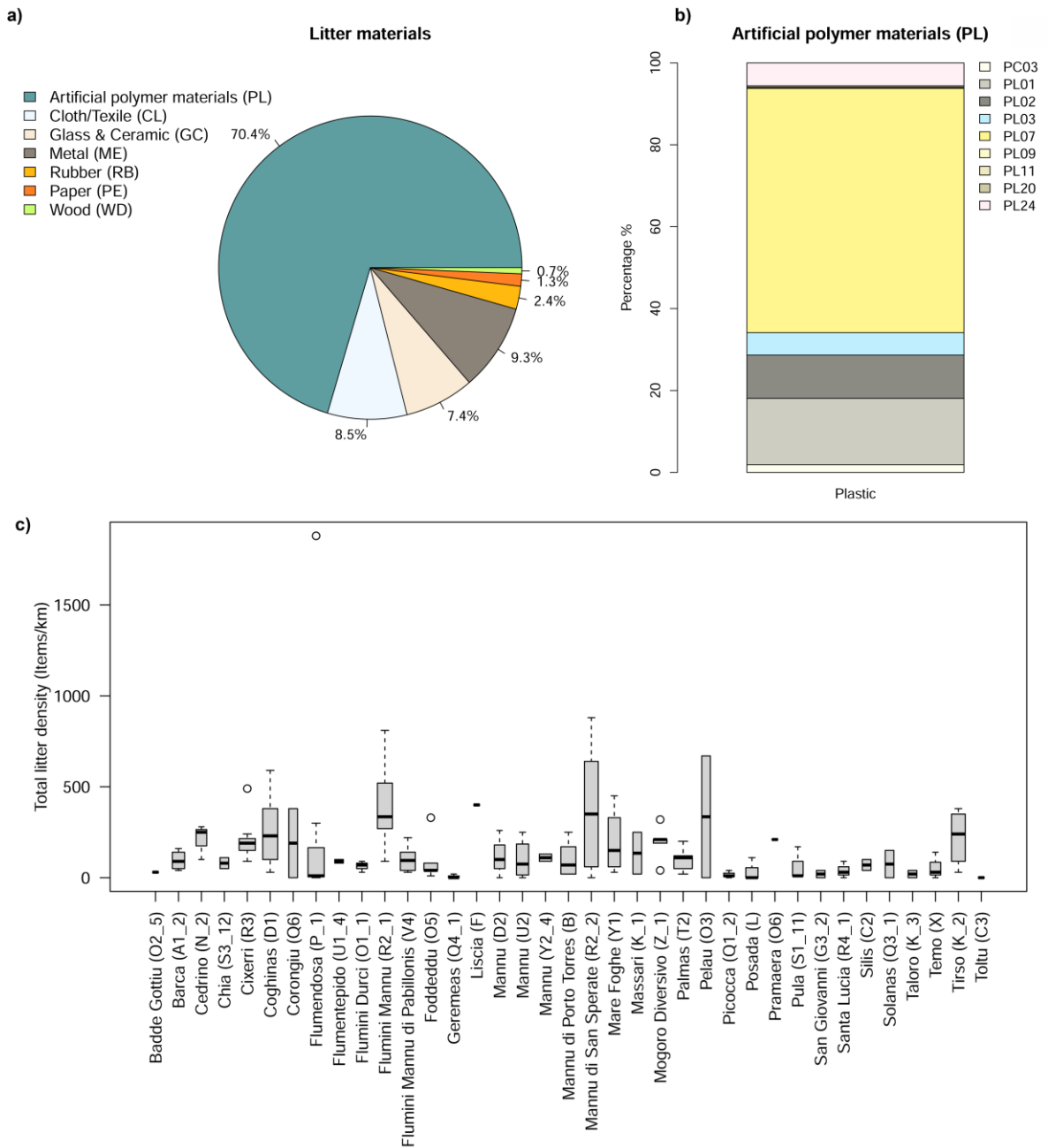


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Figure 1. Study area and sampling stations of the investigated rivers (study period 2018-2020). Upper case letters indicated the codes of rivers basins.

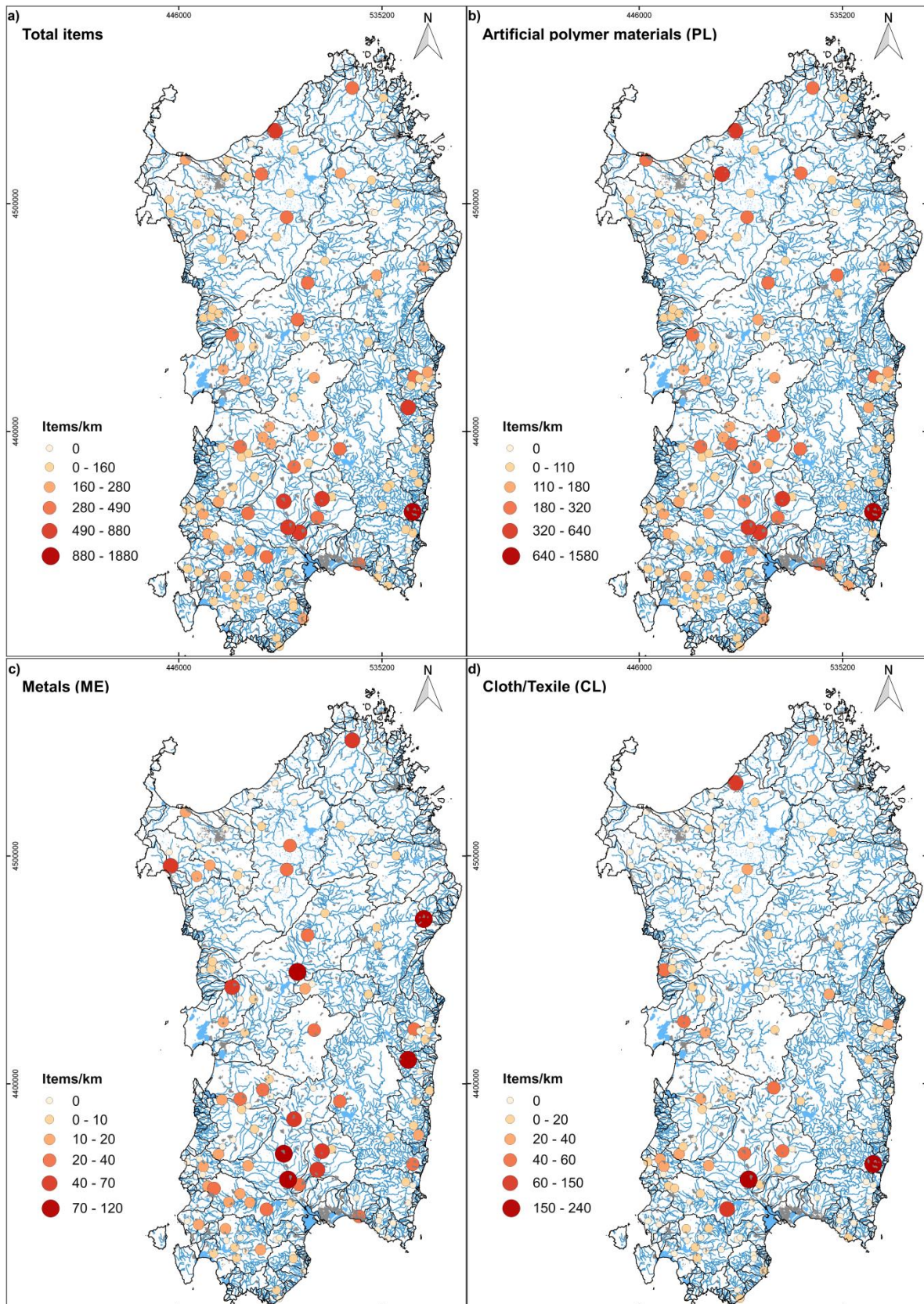


747

748 **Figure 2. (a) Percentage of the number of litter items in Sardinian rivers. (b)**
 749 **Percentage of artificial polymer materials (PL). UNEP code sub-categories: PC03**
 750 **(Cups, food trays, food wrappers, cigarette packs, drink containers), PL01 (Bottles**
 751 **caps and lids), PL02 (Bottles < 2 L), PL03 (Bottles, drums, jerrycans and buckets > 2L,**
 752 **PL07 (Plastic bags (opaque & clear)), PL09 (gloves), PL11 (cigarettes, butts and**
 753 **filters), PL20 (fishing net), PL24 (other plastic items). (c) Box-whisker plot**
 754 **representation of the total litterdensity (items/km) of 37 river basins of Sardinia.**

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757

758 **Figure 3. Geographical distribution and number of items km^{-1} of the total and of the**
 759 **most abundant litter materials.**

760

761 **Table 1. Summary of the relative contributions (%) of predictor variables for a boosted**
 762 **regression tree models developed for light (PL) and heavy (ME) materials. Total**
 763 **deviance explained by the model (DEV %), cross-validated correlation between model**
 764 **prediction and observed data (R² of CV) and area under the curve (AUC).**

Material type	Predictor	Relative contribution (%)	DEV (%)	R ² of CV	AUC
Light material (PL)	Sub_catchment	34.6			
	Urban_use	12.4			
	Agricultural_use	12.0			
	Near_bridge	10.1			
	Road_density	7.51			
	Slope	5.91	40%	0.70	0.92
	Pop_density	5.45			
	Distance_Mouth	4.57			
	Grazing_use	3.68			
	River_order	2.70			
	Season	1.02			
Heavy material (ME)	Sub_catchment	22.5			
	Near_bridge	21.1			
	Pop_density	19.8			
	Grazing_use	9.34			
	Distance_Mouth	7.91			
	Agricultural_use	5.66	16%	0.57	0.83
	Slope	5.06			
	Urban_use	2.92			
	Season	2.26			
	Road_density	2.12			
	River_order	1.25			

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767

769 **Table S1. Overview of studies carried out to date on the occurrence and distribution anthropogenic litter in rivers flowing in the**
 770 **Mediterranean Sea.**

Matrix	Unit	Material	River area	Reference
Water	Items m ⁻³ and items km ⁻²	Microplastic and macroplastic	Po River (NE Italy)	(van der Wal et al., 2015)
Sediment	Items kg ⁻¹	Microplastic, mesoplastic and macroplastic	Ombro, Osa and Albegna Rivers (Central W Italy)	(Guerranti et al., 2017)
Sediment	Items kg ⁻¹	Microplastic	Cecina River (Central W Italy)	(Blašković et al., 2018)
Water	Items km ⁻²	Microplastic	Rhone River (S France)	(Schmidt et al., 2018)
Sediment and water	Items kg ⁻¹ and items m ⁻³	Microplastic	Po River (NE Italy)	(Atwood et al., 2019)
Water	Items h ⁻¹	Multiple macrolitter materials	Tiber River (Central W Italy)	(Crosti et al., 2018)
Water	Items h ⁻¹	Multiple macrolitter materials	Rhone River (S France)	(Castro-Jiménez et al., 2019)
Sediment	Items kg ⁻¹	Microplastic	Ebro River (NE Spain)	(Simon-Sánchez et al., 2019)
Water	Items m ⁻³	Microplastic	Ofanto River (SE Italy)	(Campanale et al., 2020)
Water	Items m ⁻³	Microplastic	Rhone and Tet Rivers (S France)	(Constant et al., 2020)
Sediment and water	Items kg ⁻¹ and items L ⁻¹	Microplastic and mesoplastic	Mignone River (Central W Italy)	(Gallitelli et al., 2020)
Water	Items h ⁻¹	Multiple macrolitter materials	Catalonia (NE Spain)	(Schirinzi et al., 2020)
Water	Items m ⁻³	Microplastic	Kifissos and Pinios Rivers (E Grece)	(Zeri et al., 2021)
Water	ng L ⁻¹	Micro-nanoplastic	Ebro River (NE Spain)	(Llorca et al., 2021)
Water	m ³	Multiple macrolitter materials	Segura River (SE Spain)	(Rocamora et al., 2021)
River banks	Items m ⁻²	Multiple macrolitter materials	Latium Rivers (Central Italy)	(Cesarini and Scalici, 2022)
River banks	Items km ⁻¹	Multiple macrolitter materials	Sardinian Rivers	Present study

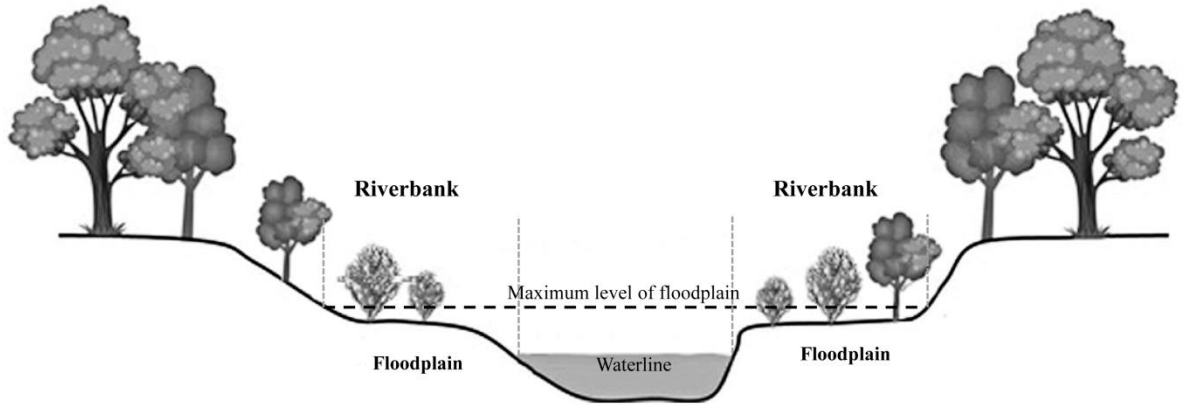
773 **Table S2. Summary of candidate predictor variables used in BTR models and determination methods.**

	Variables	Unit	Apparatus and Methods
Geomorphology	Sub_catchment	(km ²)	QGIS; Catchment area above the sampling site
	River_Order	Classified (1-5)	QGIS; Stream order <i>sensu</i> (Strahler, 1957)
	Slope	%	QGIS; Gradient of stretch
	Seasons	(Winter, Spring, Summer, Autumn)	As proxy of river runoff
Land use	Natural_use	%	QGIS; % in the catchment area above sampling site
	Agricultur_use	%	QGIS; % in the catchment area above sampling site
	Grazing_use	%	QGIS; % in the catchment area above sampling site
	Urban_use	%	QGIS; % in the catchment area above sampling site
Human pressures	Road_density	km km ⁻²	QGIS; Length of road in the catchment area above sampling site/catchment area above sampling site
	Polulation_density	Population km ⁻²	QGIS; Population in the catchment area above sampling site/catchment area above sampling site
	River_bridge	Presence/Absence (1-0)	QGIS; Bridge immediately above the sampling station
	Dams	Count	QGIS; Number of dams above the sampling station

774

775 **Table S3. Percentage of recorded items of 28 sub-categories found in Sardinian rivers.**

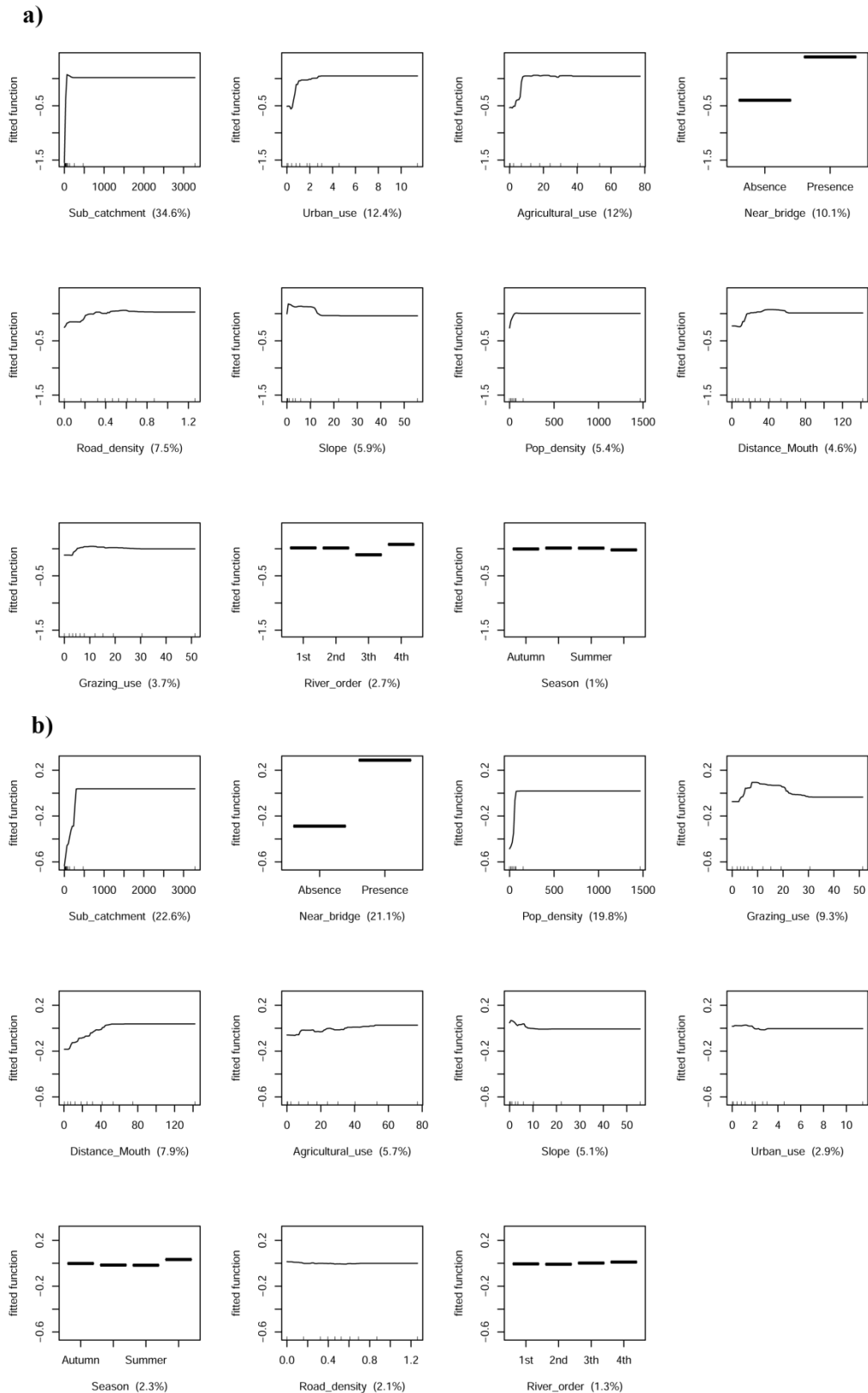
UNEP-Code	General name	Percentage of recorded items
PL07	Plastic bags (opaque & clear)	42.40
PL01	Bottles caps & lids	11.50
PL02	Bottles < 2 L	7.51
CL01	Clothing, shoes, hats & towel	6.02
GC02	Bottles & jars	4.62
PL24	Other plastic items	3.99
PL03	Bottles, drums, jerrycans & buckets > 2L	3.90
ME03	Aluminium drink cans	3.56
ME10	Appliances	3.56
GC01	Construction material	1.64
CL06	Other cloth	1.54
GC08	Other glass items	1.35
PC03	Bottles, drums, jerrycans & buckets > 2L	1.35
RB04	Tyres	1.35
RB08	Other rubber items	1.06
CL04	Rope & string	0.96
ME05	Gas bottles, drums % buckets (> 4 L)	0.77
WD04	Processed timber and pallet crates	0.67
ME09	Wire, wire mesh & barber wire	0.63
ME04	Other cans (< 4 L)	0.48
ME06	Foil wrappers	0.34
PL11	Cigarettes, butts & filters	0.19
PL20	Fishing net	0.19
ME02	Bottles caps, lids & pull tabs	0.14
CL05	Carpet & furnishing	0.10
RB05	Inner-tubes and rubber sheet	0.10
CL03	Canvas, sailcloth & sacking	0.05
PL09	Gloves	0.05



777

778 **Figure S1. Diagram of a typical stream cross-section encountered across the study area,**
779 **emphasizing the sampled area from the waterline to max. level of floodplain.**

780



781

782 **Figure S2. Partial dependence plots of the relationship between predictive variables**
 783 **and occurrence of light (PL) (a) and heavy (ME) (b) materials. The percentage**
 784 **indicates the relative contribution of each variable in the BRTs. Black lines represent**
 785 **the smoothed results.**