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

3

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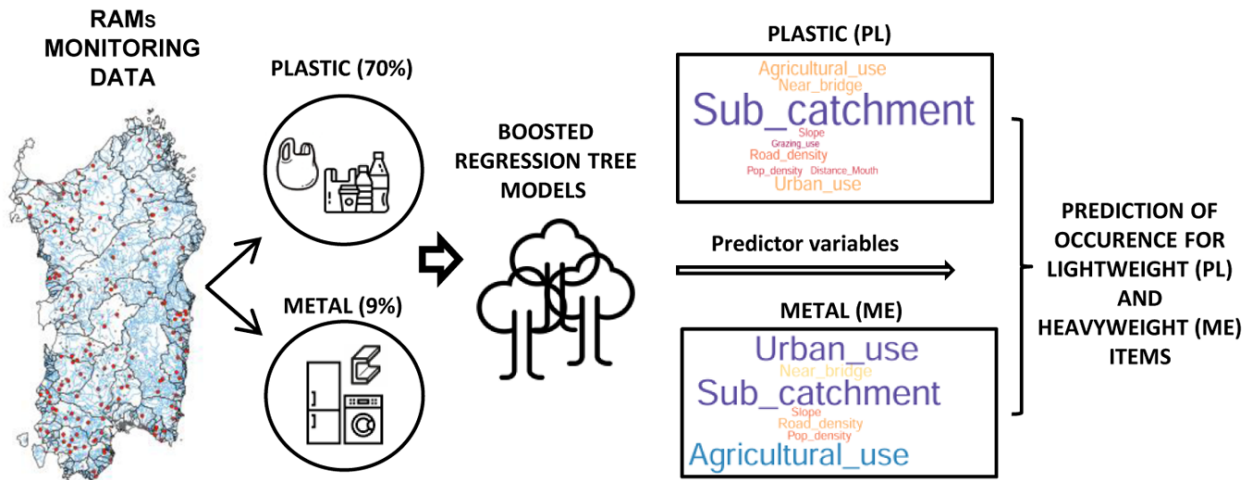
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Highlights

- Macrolitter (RAM) contamination was studied in Intermittent riverbanks.
- Collected RAM was dominated by Plastic (PL) and Metal (ME) items.
- The presence of PL items is influenced by urban and agricultural land use.
- ME occurrence is linked with presence of bridges and population density.
- Littering seems to be the main source of riverbank contamination.

14 **Graphical abstract**

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18

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. Non-linear Boosted Regression Tree models (BRTs)
30 revealed that occurrence of lightweight RAM (especially plastic) is mostly explained by
31 levels of urban and agricultural land use of the territory, whereas the proximity of bridges
32 to the sampling point and the local population density are best predictors of heavy weighted
33 RAM items (i.e., large metal items, appliances) occurrence. Our results confirm that RAM,
34 especially plastics, represent an important component of riverine waste and pinpoint that
35 actions aimed at abating litter contamination along riverbanks of intermittent rivers, apart
36 practices put in place to decrease plastic production and waste, should be localized on the
37 proximity of bridges, whatever the local population density. Finally, to fill existing
38 knowledge gaps in understanding the severity of litter discharge and accumulation in the
39 Mediterranean Sea, land-to-sea systematic monitoring campaigns at appropriate spatial and
40 temporal scales 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 some like
57 1000 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).

66 Riverine Anthropogenic Macro-litter (RAM) refers to the fraction of solid waste (>5
67 mm) present in rivers and on riverbanks (González-Fernández et al., 2021; Schmidt et al.,
68 2017). RAM originates from mismanagement of urban waste, sewage outlets from
69 wastewater treatment plants, illegal dumping, loss of products from industrial and
70 agricultural activities (Bruge et al., 2018; Faure et al., 2015; Galafassi et al., 2019;
71 Kiessling et al., 2019). RAM is typically dominated by plastics, whereas other materials
72 such as glass, metal and paper/cardboard are relatively minor contributors (Castro-Jiménez
73 et al., 2019; Cesarini and Scalici, 2022; González-Fernández et al., 2021; Rech et al., 2014).

74 Abundance, composition and distribution of RAM are influenced by the cumulative
75 effects of an array of environmental characteristics including, among the others, the
76 presence of floating and riparian vegetation (Cesarini and Scalici, 2022; Schreyers et al.,
77 2021; Williams and Simmons, 1997; Windsor et al., 2019), tidal influence, seasonal
78 changes of the water level and flow rate (Battulga et al., 2019; Vriend et al., 2020b), and
79 curvature and shape of the river (Calcar and van Emmerik, 2019). Moreover, RAM
80 accumulation can be also influenced by several anthropogenic pressures like land use
81 (Cowger et al., 2019; McCormick and Hoellein, 2016), fluvial navigation and the presence
82 of hydraulic infrastructures (Calcar and van Emmerik, 2019; Mihai, 2018; Schirinzi et al.,
83 2020), human population and road density (Battulga et al., 2019; Jambeck et al., 2015;
84 McCormick and Hoellein, 2016).

85 The Mediterranean Sea is one of the most important accumulation zones of marine
86 litter worldwide (Cózar et al., 2015; Eriksen et al., 2014; Suaria et al., 2016). Information
87 about RAM inputs from permanent rivers associated with highly populated areas in the
88 Mediterranean Sea is available (e.g. Crosti *et al.*, 2018; Castro-Jiménez *et al.*, 2019;
89 Schirinzi *et al.*, 2020; Cesarini and Scalici, 2022), but, to date, information about RAM
90 from intermittent rivers is still scant (Table S1). Intermittent rivers and ephemeral streams
91 are common across Europe and dominate river networks in the Southern basin and islands
92 of the Mediterranean Sea (Stubbington et al., 2018), including Sardinia and Sicily. As being
93 ecosystems with unpredictably temporal dynamics of water supply, the role of intermittent
94 rivers in RAM transport to the sea has been almost entirely ignored.

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

99
100

101 **Materials and Methods**

102 *Study area*

103 The study area is in Sardinia (Italy), the second largest island (ca. 24,106 km²) in the
104 Mediterranean Sea. Sardinia, with a population of 1,630,474, corresponding to a density of
105 nearly 67.7 inhabitants km⁻², is one of the less densely populated regions in Italy.
106 Bi-seasonal climatic features, with hot arid summers and smoothed rainy autumn/winter
107 seasons determine irregular flow and strong seasonal hydrological fluctuations (De Waele
108 et al., 2010; Palmas et al., 2020; Podda et al., 2020; Sabatini et al., 2018). The recurrent
109 temporal overlap of the dry season with a high water demand for agriculture irrigation,
110 industry and domestic purposes have led the construction of a total of 54 larger dams (> 15
111 m) (Marchetto et al., 2009; Montaldo and Sarigu, 2017), that, interrupting the rivers'
112 continuity, have strongly influenced their annual hydrological cycle (Moccia et al., 2020;
113 Naselli-Flores et al., 2014).

114

115 *RAM occurrence, abundance, and composition*

116 Litter monitoring was conducted over a total of 133 sampling sites (belonging to 37 river
117 basins), covering different altitudinal zones and environmental conditions across 2018 and
118 2020 (Figure 1). Litter occurrence on the riverbanks was determined according to the
119 Rivers-OSPAR protocol(van Emmerik et al., 2020a) that was based on the OSPAR beach
120 litter guidelines (OSPAR, 2010). Sampling was carried out on both riverbanks of each river
121 and data of each river were cumulated. Briefly, macro-litter items (>5 mm) on 100-m long
122 stretches of riverbanks parallel to the waterline, from the waterline itself and the maximum
123 level of flood plain landward, were counted and, for each station, litter density was
124 calculated as the number of items per kilometer of riverbank (items km⁻¹).

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

130 undefined (UN). The frequency of occurrence for each litter category was estimated as the
131 percentage of stations containing items over the total number of stations.

132

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

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

138 Geo-morphological variables include: i) the sub-catchment area (km²) above the
139 sampling site as a proxy of catchment runoff; ii) the river order as a proxy of
140 upstream–downstream gradients; iii) the stream slope as a proxy of potential water
141 velocities. Season of sampling was also used as a proxy of river discharge events,
142 considering the peculiar abovementioned climatic features of Sardinia.

143 Land use data, obtained from the CORINE database, were merged in four categories:
144 natural use, agricultural use, grazing use and urban use. Land cover was expressed as the
145 percentage (%) of each of these categories in the sub-catchment area above each sampling
146 site.

147 Human pressure proxies were estimated in terms of: i) road density (km km⁻²), ii)
148 population density (population km⁻²), iii) the presence/absence of river bridges immediately
149 above the sampling point. In addition, since the presence of weirs and dam could negatively
150 influence the presence and transport waste items in the riverbanks, the number of dams
151 above the sampling station were also considered.

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

154

155 ***Statistical analyses***

156 Density from each station was then used to generate distribution maps of the most
157 important litter categories. The free Quantum GIS Desktop, version 2.18.3 (QGIS)

158 (<http://www.qgis.org/>) software was used for creating distribution maps and to extract the
159 exploratory variables. Sub-catchment area, slope as well as the stream order of each
160 sampling site were calculated based on 10-m resolution Digital Elevation Model (DEM).
161 For the entire river network generated by flow accumulation, stream order was derived with
162 the Hack method's (Hack, 1957).

163 Since the artificial polymers and metal materials (PL and ME, respectively) were the
164 most abundant anthropogenic litter material (cumulatively accounting for ca. 80% of total
165 litter), we tested whether and to which extent some potential explanatory variables were
166 putative drivers for PL and ME litter occurrence using a non-linear Boosted Regression Tree
167 model (BRT, Elith *et al.*, 2008). As we aimed at identifying the conditions that might
168 represent a threshold over which light (PL) and heavy materials (ME) could be found, we
169 used only presence/absence transformed data. BRT models have been used to analyse the
170 relationships between response and predictor variables in different fields of environmental
171 science (Ju *et al.*, 2021; Lagarde *et al.*, 2021; Lemm *et al.*, 2021; Saha *et al.*, 2021). BRT
172 models allow testing different types of predictive variables by fitting complex non-linear
173 relationships and handling interaction effect between predictors, while not depending on the
174 normality and homoscedasticity of the data (De'ath, 2007; Elith *et al.*, 2008).

175 To fit the BRT models, the learning rate (the importance of each iteration in the model)
176 and tree complexity were set through an iterative process to ensure that the final model
177 outcome consisted of at least 1000 decision trees (Elith *et al.*, 2008). The relative
178 importance of each predictor variable has been also calculated from the BRT model and
179 was visualised in partial dependence plots. BRT models were run with a Bernoulli link
180 function. The BTRs' performance was evaluated by the amount of total deviance explained
181 (DEV %) and by cross-validated correlation between model prediction and observed data
182 (R^2 of CV) (Derville *et al.*, 2016; Ju *et al.*, 2021; Nieto and Mélin, 2017; Saha *et al.*, 2021).
183 The predictive performance of the BTRs were also tested and evaluated using the
184 threshold-independent Receiver-Operating Characteristic (ROC) curve and the estimation
185 of the area under ROC plot (AUC) (Amorim *et al.*, 2016; Derville *et al.*, 2016; Saha *et al.*,
186 2021; Wang *et al.*, 2021). Collinearity among covariates was tested by computing pairwise

187 scatter plots among covariates. Covariates showing relevant Spearman's Rho ($\rho > 0.7$) were
188 discarded from the modelling. The Variance Inflation Factor (VIF) was also used to check
189 collinearities among explanatory variables; those showing VIF >3 were also discarded from
190 the analysis (Zuur et al., 2010).

191 All analyses were carried out using the statistical software package R (R Core Team,
192 2021). BRTs are estimated using the "dismo" library (Hijmans et al., 2011). The ROC
193 analysis was performed using the R package pROC (Robin et al., 2011).

194

195 **Results**

196 A total of 2078 RAM items were collected from the 37 river basins, covering ca. 22 linear
197 km of riverbanks. Out of 133 sampling stations, 114 (85.7%) showed the presence of litter
198 items, and only 19 were litter free. Overall, 28 sub-categories of litter items were found on
199 Sardinian rivers, even if the top 5 most abundant types of items represented most of the
200 litter found (~70%, Table S3). Artificial polymers materials (PL, 70.4%) were the most
201 abundant category, followed by Metal (ME, 9.3%), Cloth/Textile (CL, 8.5%) and Glass &
202 Ceramic (GC, 7.4%) (Figure 2a). Other materials (including rubber; RB), paper; PE, and
203 processed wood; WD) represented cumulatively 4.4% of all the litter. Artificial polymers
204 items consisted mostly of single-use plastic items such as bags (PL07, 60%), bottles caps
205 and lids (PL01, 16%) and small bottles (PL02, 11%) (Figure 2b).

206 Overall, the mean litter density for all investigated riverbanks was 156 ± 19 items km^{-1}
207 (median value of 90 items km^{-1}). This absence of significant differences was observed
208 using both aggregated data and dominant categories, treated singularly. The highest litter
209 mean density was measured in the river basins of Flumini Mannu (R2_1) (393 ± 100 items
210 km^{-1}), Mannu di San Sperate (R2_2) (386 ± 168 items km^{-1}), Pelau (O3) (335 ± 335 items
211 km^{-1}) and Flumendosa (P_1) (318 ± 263 items km^{-1}) (Figure S1). With the exception of
212 Pelau river basin (O3), where the highest litter densities were measured at the most
213 upstream location and mainly composed of glass and ceramic items (GC) (670 items km^{-1})
214 (Figure 3a), all other river basins were characterized by litter dominated by PL, with the
215 highest PL items densities recorded in downstream sampling stations (Figure 3b).

216 Analysis of multi-collinearity among predictive variables revealed strong correlations
217 between natural use (Natural_use) and agricultural use (Agricultural_use) ($\rho = -0.9$) and
218 sub-catchment area (Sub_catchment) and number of larger dams above the sampling station
219 (Dams) ($\rho = 0.7$) (Figure S2). After removing Natural_use and the Dams variables, the
220 VIF values did not exceed 3.0 (Table S4).

221 The results of BTR model revealed that the presence of plastic litter (PL) is influenced
222 by the joint effect of geomorphological variables, land use and human pressures. The PL
223 model accounted for 40% of the total deviance and a CV correlation between predicted and
224 observed data of 0.70. The analysis of the relative importance of the different predictors
225 revealed that the sub-catchment area (Sub_catchment) (34.5%), urban use (Urban_use)
226 (12.3%), agricultural use (Agricultural_use) (12.0%), the presence of a bridge above the
227 sampling point (Near_bridge) (10.1%) and road density (Road_density) (7.5%) represented
228 the highest share in relative explained deviance (Table 1). The partial responses of single
229 predictors showed a predominantly positive linear trend with a plateau for Sub_catchment,
230 Urban_use, Agricultural_use, and Road_density (Figure 4). In particular, the sub-catchment
231 curve is steeper than that of all other predictors and reaches a peak at a relatively small
232 surface area (60 km^2). The effects of land use variables were approximately J-shaped, with
233 the probability of occurrence of PL litter significantly increased after 0.5% and 6.5% of
234 coverage area for urban and agricultural use, respectively (Figure 4). The PL litter is most
235 found also in stretches of rivers characterized by the presence of a bridge. Road density
236 concentration had a consistent positive relationship with PL items' probability of occurrence
237 (Figure 4).

238 The BRT model applied to heavy materials (ME) explained 16% of the predicted
239 deviance and a CV correlation between predicted and observed data of 0.57. The most
240 significant predictor for ME occurrence were the sub catchment area above the sampling
241 station (22.6%) followed by the presence of a bridge (Near_bridge, 21.1%) and
242 population's density (Pop_density, 19.8%) (Table 1).

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

246

247 **Discussion and Conclusions**

248 Our study provides evidence that a high frequency (~ 86%) of intermittent rivers in
249 Sardinia is contaminated by Riverine Anthropogenic Macro-litter (RAM). We show here
250 that RAM composition in Sardinian intermittent rivers is dominated by PL items (~70 %)
251 followed by metal (ME) and glass (GL) (9% and 7%, respectively). Our results confirm
252 previous studies worldwide which reported PL items as the globally dominant category in
253 riverbanks (Battulga et al., 2019; Bruge et al., 2018; Carpenter and Wolverton, 2017;
254 Hoellein et al., 2014; Kiessling et al., 2019; Rech et al., 2014; Schöneich-Argent et al.,
255 2020; van Emmerik et al., 2020a). Within the PL macro-category, single-use items such as
256 plastic bags (~60%) and tips (11%) are the most common items on Sardinian riverbanks, as
257 also observed in other European locations (Bernardini et al., 2020; Winton et al., 2020).

258 The mean density of total litter for the whole study period is 156 items km⁻¹ (median
259 90 items km⁻¹) with a remarkable heterogeneity among different river basins, despite their
260 distance from each other. The Flumini Mannu is the most polluted basin (393± 100 items
261 km⁻¹), followed by the Mannu di San Sperate (386±168 items km⁻¹) and Flumendosa
262 (318±263 items km⁻¹). Differences in geographical locations, social and economic context
263 and sampling techniques make comparisons of RAM abundance among freshwater studies
264 very challenging (Blettler et al., 2018; van Emmerik et al., 2019; Vriend et al., 2020a).
265 Nevertheless, considering those studies focused on riverbanks that used the same sampling
266 approach of this study, we report here that the RAMs' mean abundance in the intermittent
267 rivers in Sardinia are an order of magnitude lower than those reported for the Rhine and the
268 Meuse rivers (North Sea) (median value 2060 items km⁻¹) (van Emmerik et al., 2020a,
269 2020b). It is worth of notice that these systems are characterized by different hydrological
270 and anthropogenic factors such as discharge and population density.

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

278 Our model predicted that the larger the sub-catchment area surface above the sampling
279 station, the more the occurrence of light (PL) RAMs can be expected. The primary role of
280 the sub-catchment area surface can be associated with the fact that most plastic waste
281 originates, generally, from land-based areas due to the littering or illegal landfill of waste
282 (Chae and An, 2018; Geyer et al., 2017). There, light (PL) RAMs may be washed away
283 from drainage areas by the synergic effects of wind, heavy rainfall and floods (Bruge et al.,
284 2018; Carpenter and Wolverton, 2017; Windsor et al., 2019; Zylstra, 2013). Moreover, the
285 larger the sub-catchment area, the more riparian vegetation plays a role in litter
286 accumulation. In this regard, we report here that, for instance, plastic bags were mostly
287 found trapped in the vegetation at the riverside, which is known to act as a trap for floating
288 materials (Schöneich-Argent et al., 2020; Schreyers et al., 2021; van Emmerik and
289 Schwarz, 2020; Williams and Simmons, 1997).

290 The same, though lower, explanatory power on litter occurrence is observed for the
291 agricultural and urban percentage use of the territory, the increase of which has been
292 already reported to positively affect PL RAMs' occurrence (Bruge et al., 2018; Carpenter
293 and Wolverton, 2017; Cowger et al., 2019; Glanville and Chang, 2015; Guerranti et al.,
294 2020; van der Wal et al., 2015). This effect appears particularly relevant in the subset of
295 Sardinian rivers (Flumini Mannu and Mannu di San Sperate) that run along the Campidano
296 plain, characterized by an almost entire occupancy of the territory by agricultural and dense
297 different urban uses.

298 The third explanatory factor of PL RAMs occurrence identified by our model is the
299 presence of bridges immediately above the sampling station. This result could depend on

300 the slowdown of the river flow which favors the accumulation of waste on the riverbanks
301 (Hoellein et al., 2014; Kiessling et al., 2019; Lebreton et al., 2017). Moreover, the
302 prevalence of plastic bags in those localities suggests the persistence of the incorrect
303 behavior of abandoning waste in places that, due to the landscape attractiveness of bridges,
304 makes them often used for refreshment breaks of tourists, motorists, and campers.

305 The sub-catchment area, the presence of bridges and the population density are the
306 most important predictors also of the occurrence of heavy materials (ME) in Sardinian
307 riverbanks. The highest abundance of discarded house appliances and aluminum drink cans
308 (38% and 37% of the total ME items, respectively) suggests that the illegal disposal and
309 dumping of ME are main sources of litter. Moreover, our results reveal also that illegal
310 landfills in proximity of bridges and secondary roads are much more common in
311 sub-catchment areas characterized by higher population density. This result, again, fits with
312 the observations made on riverbanks of Chile, Wales and Romania, where the combination
313 of illegal dumping and human presence, more than the road density, have been identified as
314 the main sources of litter items occurrence (Williams and Simmons, 1997b; Rech *et al.*,
315 2014; Mihai, 2018; Kiessling *et al.*, 2019; Cowger *et al.*, 2019).

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

321 At a regional scale, our study highlights that the occurrence of light (plastics) and
322 heavy (metal) waste materials largely depends upon the combined effects of river natural
323 geomorphology, the presence of bridges and the land use. The complex hydrological
324 scenario that characterizes Sardinian coastal marine waters (Olita et al., 2013; Palmas et al.,
325 2017) does not allow to infer about the actual linkages between the abundance of
326 macro-litter observed across marine coastlines (Alvito et al., 2018; Cau et al., 2022) and
327 their occurrence in riverbanks, since floating debris might end up well far from the source
328 (Cózar et al., 2015; González-Fernández et al., 2021). Moreover, we cannot exclude that

329 other factors, not included in this study, and associated with hydrological (runoff, flow
330 velocity, discharge, vegetation cover) and anthropogenic factors (tourism and recreation
331 activities, poor waste management practices on land) could explain a certain portion of
332 RAMs occurrence variance in riverbanks and their potential transport to the sea (Bruge et
333 al., 2018; Kiessling et al., 2019; Schirinzi et al., 2020; Windsor et al., 2019). Despite the
334 above biases, we must notice here that, since we limited our analysis to macroscopic RAMs
335 (> 5mm), the potential severity of our results could rise when considering smaller size
336 items (Schöneich-Argent et al., 2020).

337 Finally, we contend here that, to fill existing knowledge gaps in understanding the
338 severity of litter discharge and accumulation in the Mediterranean Sea, land-to-sea
339 systematic monitoring campaigns at appropriate spatial and temporal scales should be put
340 in place, always considering the complex array of factors influencing litter sources,
341 distribution, accumulation, and transport (van Emmerik et al., 2020b; Vriend et al., 2020a).

342 We pinpoint also that further special effort should be also paid to optimize and
343 standardize protocols of identification, characterization and quantification of RAMS in
344 different environments (Bernardini *et al.*, 2020; van Emmerik *et al.*, 2020b; Vriend *et al.*,
345 2020b).

346

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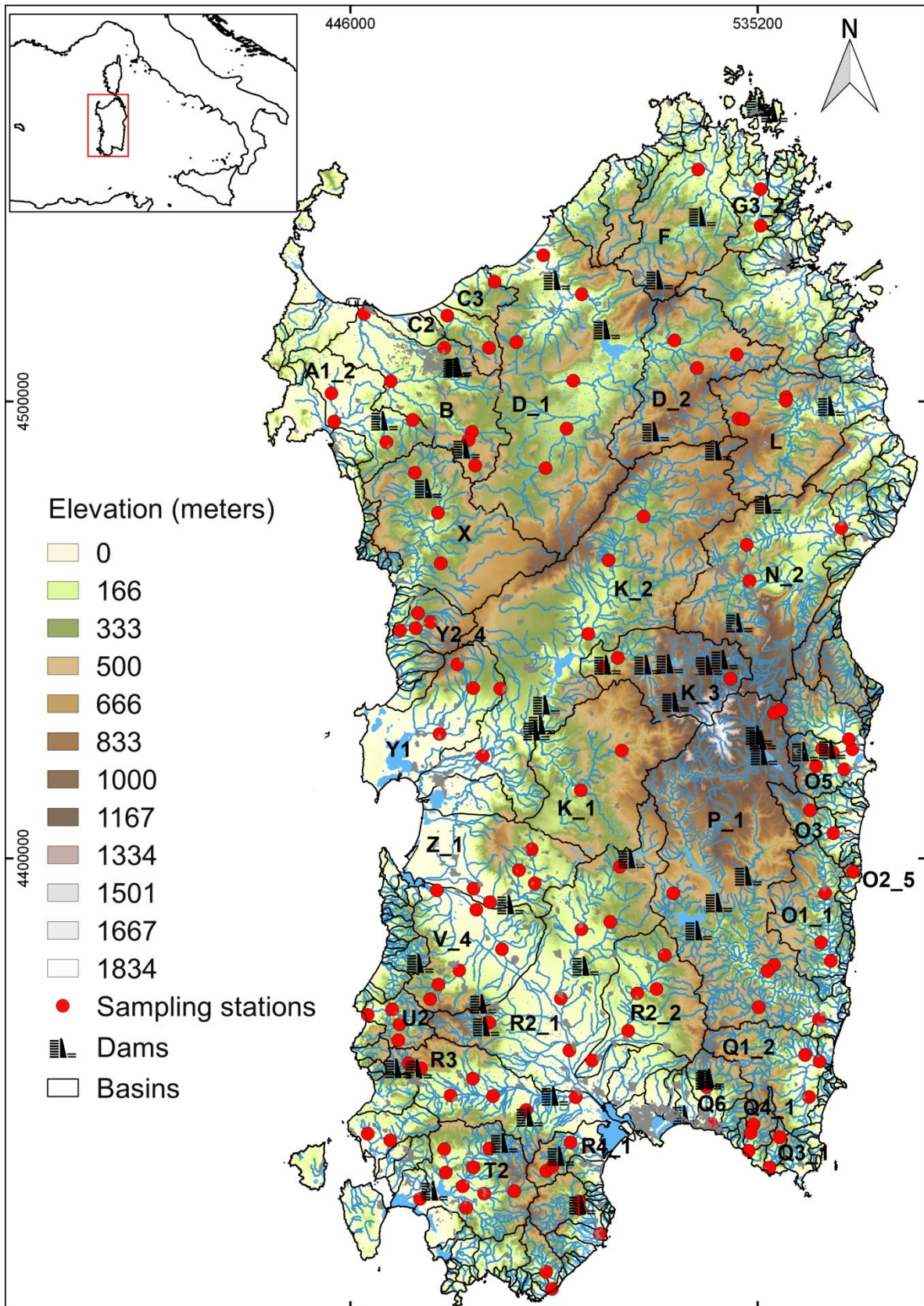
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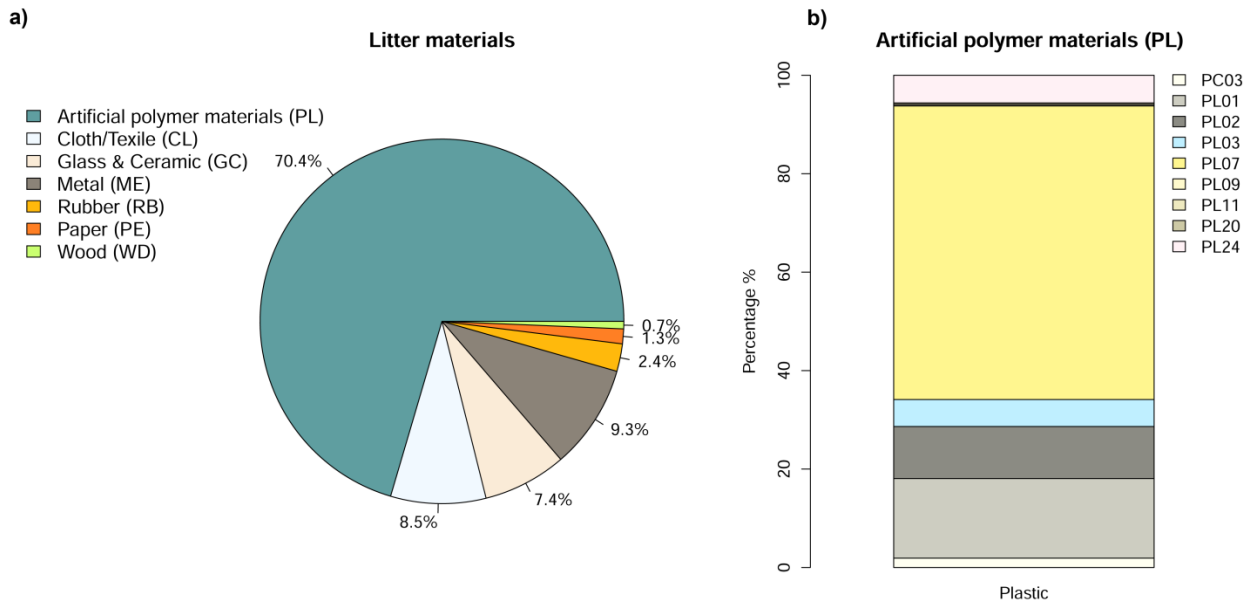
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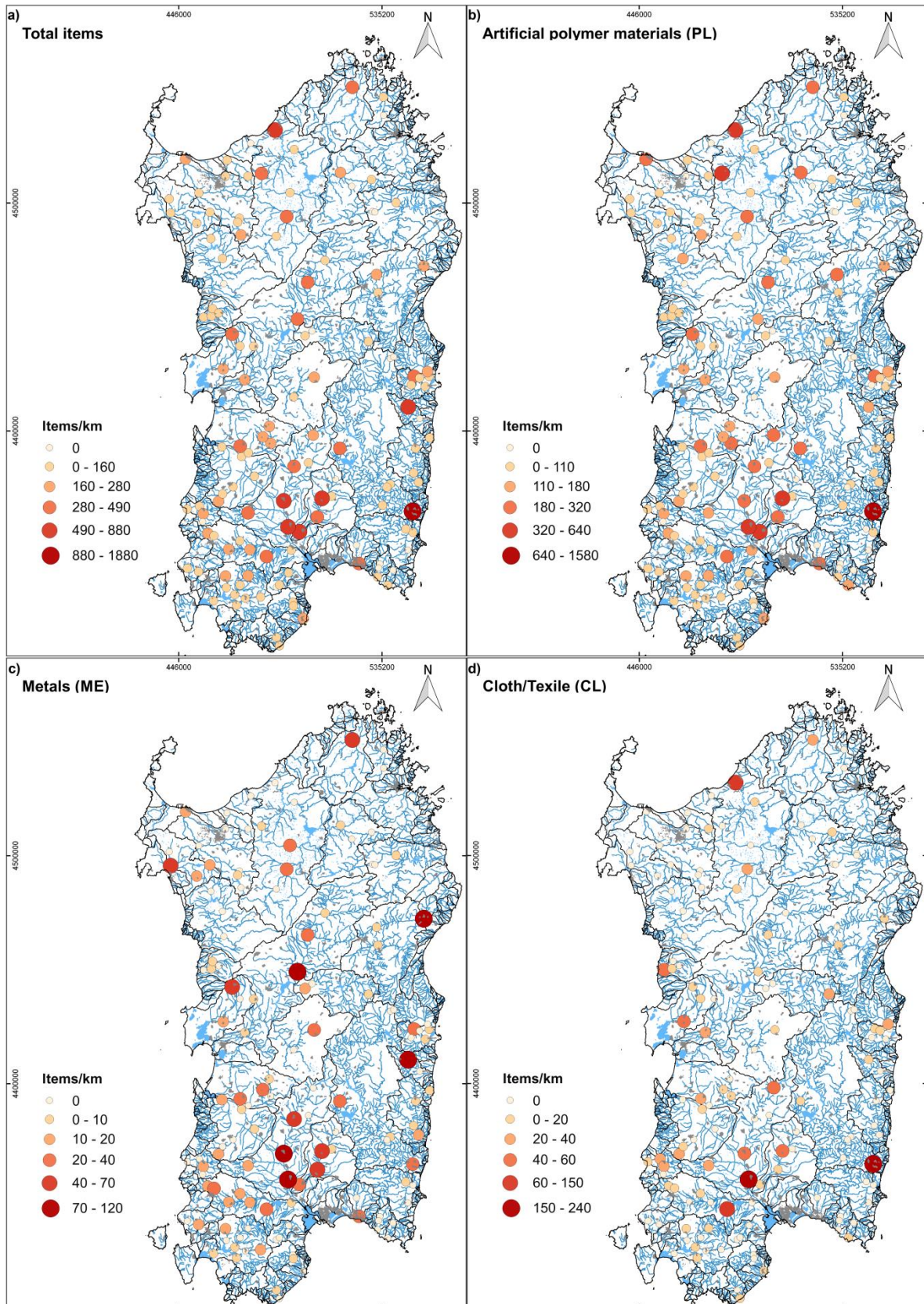
653 **Figure 1. Study area and sampling stations of the investigated rivers. Upper case letters**
 654 **indicated the codes of rivers basins.**



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656 **Figure 2. (a) Percentage of anthropogenic litter materials in Sardinian rivers. (b)**
 657 **Percentage of artificial polymer materials (PL). UNEP code sub-categories: PC03**
 658 **(Cups, food trays, food wrappers, cigarette packs, drink containers), PL01 (Bottles**
 659 **caps and lids), PL02 (Bottles < 2 L), PL03 (Bottles, drums, jerrycans and buckets > 2L,**
 660 **PL07 (Plastic bags (opaque & clear)), PL09 (gloves), PL11 (cigarettes, butts and**
 661 **filters), PL20 (fishing net), PL24 (other plastic items).**

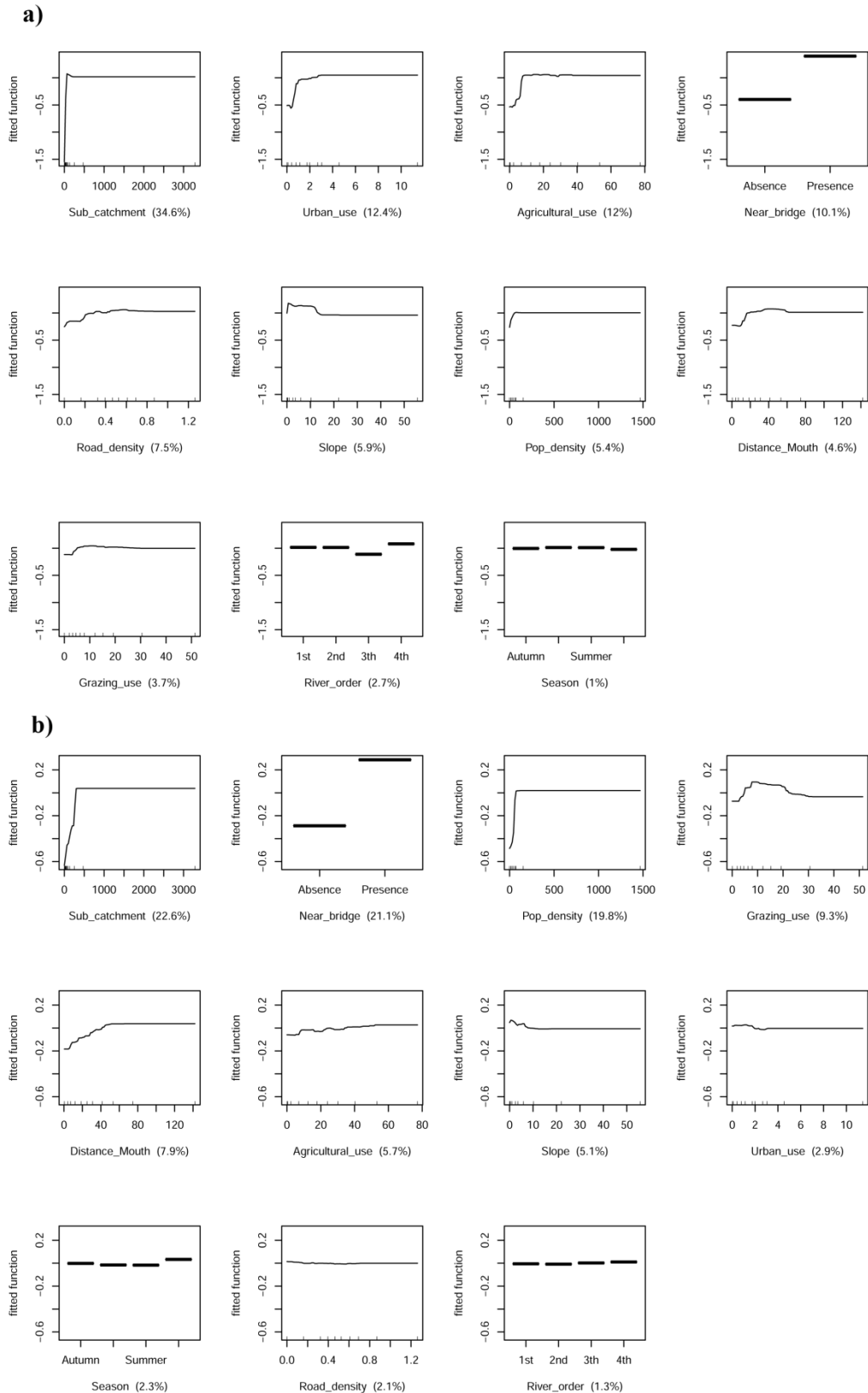
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664 **Figure 3. Geographical distribution and number of items km^{-1} of the total and of the**
 665 **most abundant litter materials.**

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668 **Figure 4. Partial dependence plots of the relationship between predictive variables and**
 669 **occurrence of light (PL) (a) and heavy (ME) (b) materials. The percentage indicates the**
 670 **relative contribution of each variable in the BRTs. Black lines represent the**
 671 **smoothed results.**

672 **Table 1. Summary of the relative contributions (%) of predictor variables for a boosted**
673 **regression tree models developed for light (PL) and heavy (ME) materials. Total**
674 **deviance explained by the model (DEV %), cross-validated correlation between model**
675 **prediction and observed data (R² of CV) and**

Material type	Predictor	Relative contribution (%)	DEV (%)	R ² of CV	AUC
Light material (PL)	Sub_catchment	34.584			
	Urban_use	12.399			
	Agricultural_use	12.042			
	Near_bridge	10.132			
	Road_density	7.510			
	Slope	5.911	40%	0.70	0.92
	Pop_density	5.449			
	Distance_Mouth	4.572			
	Grazing_use	3.680			
	River_order	2.704			
	Season	1.017			
Heavy material (ME)	Sub_catchment	22.550			
	Near_bridge	21.097			
	Pop_density	19.832			
	Grazing_use	9.340			
	Distance_Mouth	7.908			
	Agricultural_use	5.663	16%	0.57	0.83
	Slope	5.057			
	Urban_use	2.919			
	Season	2.258			
	Road_density	2.122			
	River_order	1.255			

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680 **Table S1. Overview of studies carried out to date on the occurrence and distribution anthropogenic litter in rivers flowing in the**
 681 **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

684 **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 sensu Hack (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

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686 **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

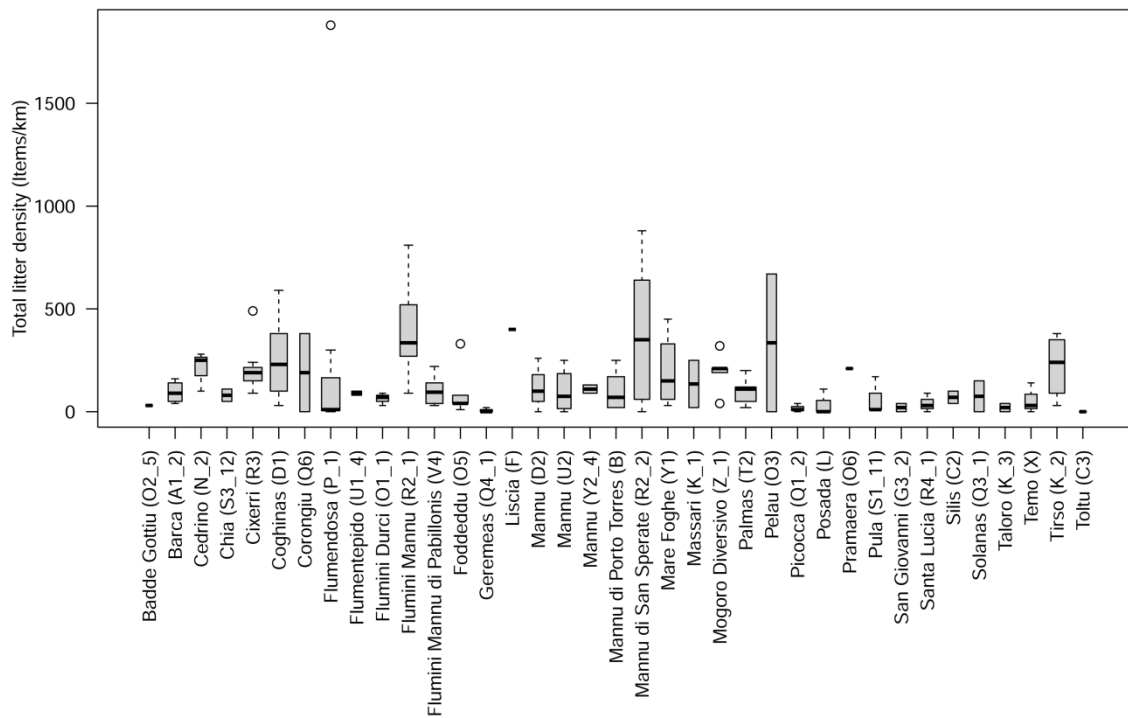
688 **Table S4. Result of collinearity test.**

Selected variables	Variance inflation factor (VIF)
Sub_catchment	1.070739
Slope	1.14281
Distance_Mouth	1.163589
Agricultural_use	1.660203
Grazing_use	1.300179
Urban_use	1.848669
Road_density	1.47726
Pop_density	1.298474

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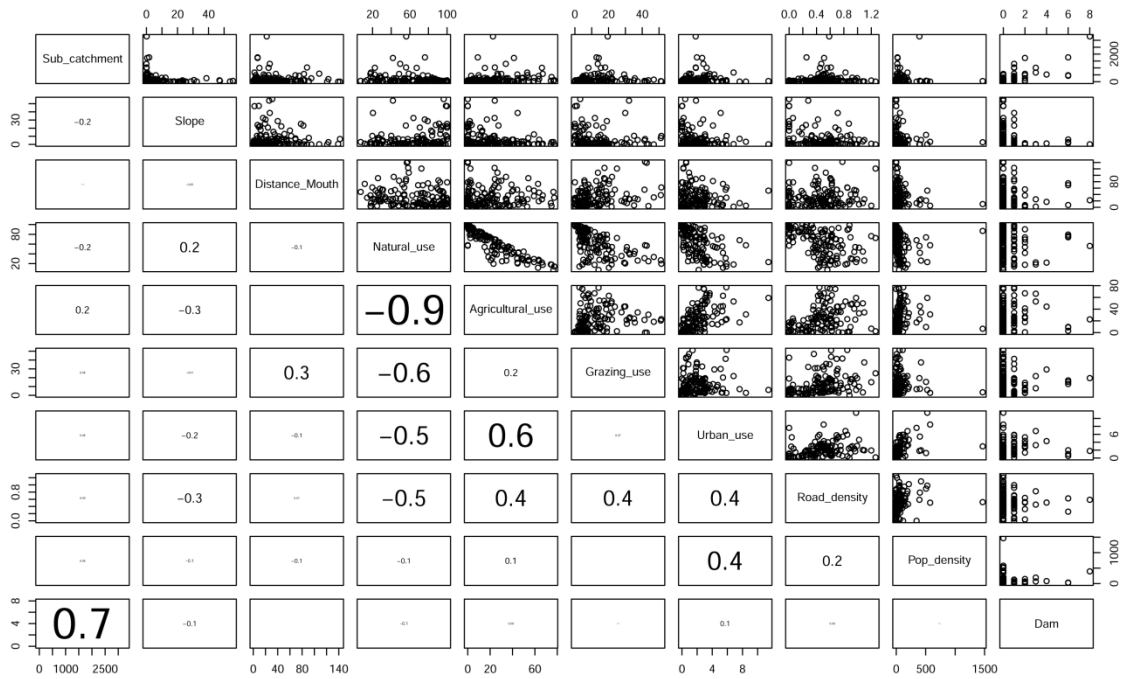


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693 **Figure S1.** Box-whisker plot representation of the total litter density of 37 river basins of
 694 Sardinia.

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698 **Figure S2.** Scatter plot matrix depicting the bivariate relationships among potential
 699 explanatory variables. Upper diagonal: scatter plots. Lower diagonal: Spearman's correlation
 700 coefficients among variables.

CRedit authorship contribution statement

F. Palmas: Conceptualization, Field sampling, Data curation, Software, Formal analysis, Formal Writing - Original draft. **Al. Cau:** Writing—review & editing. **C. Podda:** Field sampling, Software, Formal analysis, Writing—review & editing. **A. Musu:** Field sampling, Writing—review & editing. **A. Pusceddu:** Supervision, Writing – review & editing. **M. Serra:** Field sampling, Writing—review & editing. **A. Sabatini:** Project administration, Supervision, Writing - review & editing.