

See discussions, stats, and author profiles for this publication at: <https://www.researchgate.net/publication/354767473>

What You Net Depends on if You Grab: A Meta-analysis of Sampling Method's Impact on Measured Aquatic Microplastic Concentration

Article in *Environmental Science and Technology* · September 2021

DOI: 10.1021/acs.est.1c03019

CITATIONS

2

READS

67

3 authors, including:



Lisa Watkins

Cornell University

6 PUBLICATIONS 130 CITATIONS

[SEE PROFILE](#)



M. Todd Walter

Cornell University

273 PUBLICATIONS 7,486 CITATIONS

[SEE PROFILE](#)

Some of the authors of this publication are also working on these related projects:



CHALPI - FLOW [View project](#)



Role of biofilms in mediating the biogeochemistry of stream ecosystems [View project](#)

1 **This manuscript has been peer reviewed and accepted to *Environmental***
2 ***Science & Technology*. It is available online at doi.org/10.1021/acs.est.1c03019.**
3 **Supporting Information for this article, including associated data, is available**
4 **free of charge at the journal link above.**
5
6
7
8
9
10
11

12 What you net depends on if you grab: A meta-analysis of sampling method's
13 impact on measured aquatic microplastic concentration
14
15
16
17
18
19

20 *Lisa Watkins*^{1*}, *Patrick J. Sullivan*², *M. Todd Walter*¹
21

22 ¹Department of Biological and Environmental Engineering, Cornell University, Ithaca, New
23 York, 14853, USA

24 ²Department of Natural Resources and the Environment, Cornell University, Ithaca, New York,
25 14853, USA

26 *Corresponding author: ltw35@cornell.edu
27
28
29
30
31
32

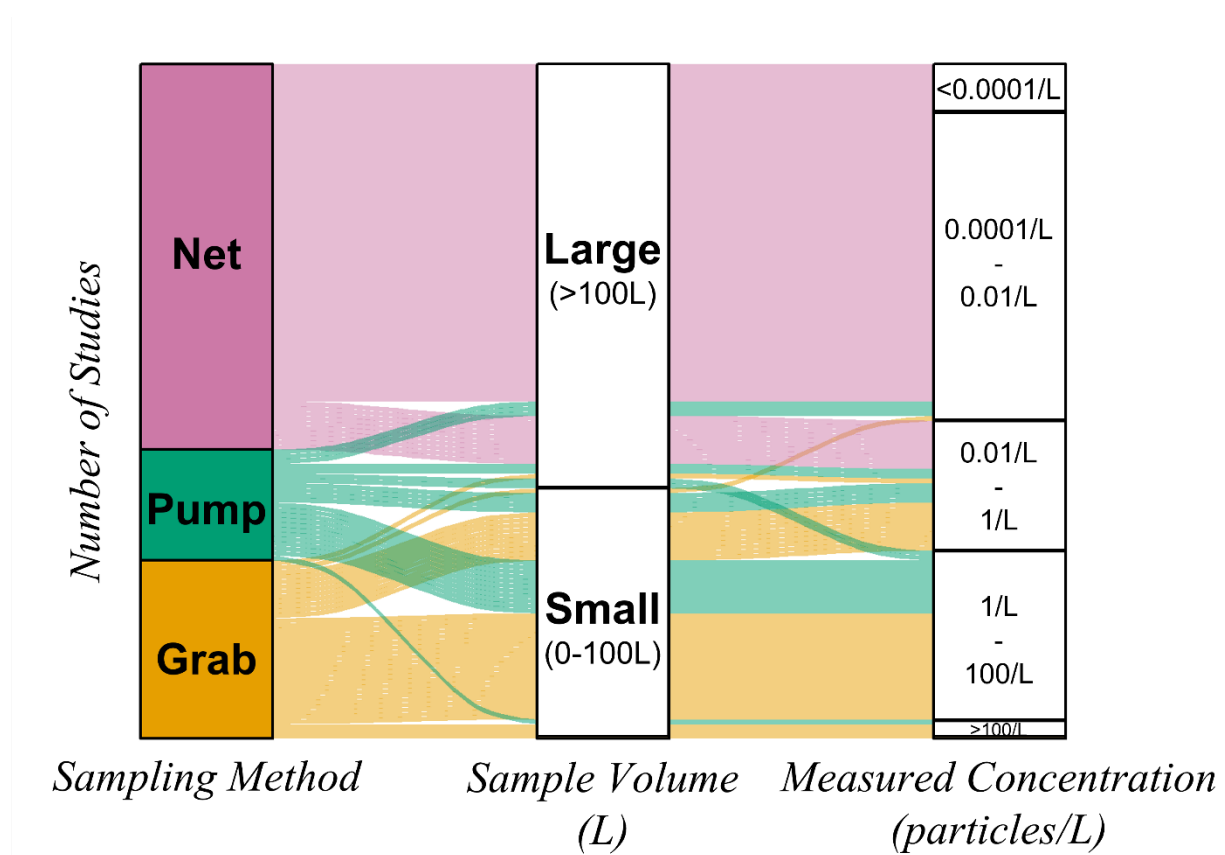
33 **Keywords**

34 Plastic, pollution, surface water, net, grab, pump, contamination, mesh size
35
36
37

38 **Suggested Citation:**

39 Watkins L, Sullivan PJ, Walter MT. 2021. What you net depends on if you grab: A meta-analysis
40 of sampling method's impact on measured aquatic microplastic concentration. *Environmental*
41 *Science & Technology*. 55(19): 12930–12942.

42 TOC/Abstract Graphic:



43

44 **Abstract**

45 Microplastic pollution is measured with a variety of sampling methods. Field experiments
46 indicate that commonly used sampling methods, including net, pump and grab samples, do not
47 always result in equivalent measured concentration. We investigate the comparability of these
48 methods through a meta-analysis of 121 surface water microplastic studies. We find systematic
49 relationships between measured concentration and sampled volume, method of collection, mesh
50 size used for filtration, and waterbody sampled. Most significantly, a strong log-linear
51 relationship exists between sample volume and measured concentration, with small-volume grab
52 samples measuring up to 10^4 particles/L higher concentrations than larger volume net samples,
53 even when sampled concurrently. Potential biasing factors explored included filtration size ($\pm 10^2$
54 particles/L), net volume overestimation ($\pm 10^1$ particles/L), fiber loss through net mesh (unknown
55 magnitude), intersample variability ($\pm 10^1$ particles/L), and contamination, the potential factor
56 with an effect large enough ($\pm 10^3$ particles/L) to explain the observed differences. Based on
57 these results, we caution against comparing concentrations across multiple studies or combining
58 multiple study results to identify regional patterns. Additionally, we emphasize the importance of
59 contamination reduction and quantification strategies, namely that blank samples from all stages
60 of field sampling be collected and reported as a matter of course for all studies.

61

62 **Synopsis**

63 This work uncovers a literature-wide bias in microplastic concentrations, related to sampling
64 method, with steps to remedy the comparability error.

65

66

67 **1. Introduction**

68 Microplastics, plastic particles less than 5mm in size, have been detected in water
69 worldwide including systems as pristine as those in the Pyrenees¹, as remote as the deep ocean²,
70 and seemingly everywhere in between³. These particles are either manufactured at sizes less than
71 5 mm or are the result of breakdown from UV exposure and physical abrasion of larger plastics.
72 Microplastics are of concern because of their observed and hypothesized effects on aquatic
73 organisms⁴⁻⁶. In particular, the concern comes from microplastics' propensity to introduce
74 chemical additives into and transport adsorbed contaminants within aquatic environments and
75 organisms^{7,8}.

76 The extent of microplastic pollution remains a fundamental question for the field. To
77 answer this, study results from spatial surveys are commonly aggregated to create regional and
78 global pictures of hotspots and average concentrations^{3,9-11}. Unfortunately, studies follow a
79 variety of evolving methodologies, and the comparability of results from studies that rely on
80 differing methodologies is generally unknown. Before regulations can be based on an
81 aggregation of regional results, it is imperative to understand how methodological choices affect
82 microplastic measurements.

83 In this study, we focus on how three different, but commonly used, field sampling
84 methods affect microplastic quantification: nets, bottles, and pumps. These methods largely
85 mimic those used for neustonic plankton sampling, due in part to microplastic contamination
86 being first reported by plankton researchers^{12,13}.

87 Net sampling deploys nets for a constant distance (if the net is moving) or time (if water
88 is flowing). Sample volume, typically ~10,000L, varies based the area of submerged net mouth
89 and the stream velocity or length-of-tow (in non-flowing waters). To avoid clogging the net with

90 organic material during sampling, a relatively large mesh size is used, often ~ 0.333 mm¹⁴.
91 Samples are collected at the base of the net, in a removable “cod end”, made of the same material
92 as the net, typically nylon. Because of their large size, cleaning nets between samples can be
93 difficult. Currently, they are still the most common sampling equipment used in oceanic settings,
94 as well as in lakes and large streams¹⁵.

95 Contrastingly, bottles are used to collect grab, or “bulk”, samples. These samples collect
96 much smaller volumes than a net sample, often 1-10L, but have the benefit of being able to
97 collect even the smallest particles. Small particles are most relevant to ecotoxicity questions,
98 adding special value to methods that allow such particles to be retained¹⁶. Compared to nets,
99 bottles are a less expensive, more intuitive, and faster method for sample collection, transport
100 and storage. These factors mean they are a frequent choice for citizen science projects, an
101 important approach to research that allows for a greater quantity of data to be collected while
102 also providing opportunities for science education and community dialogue.

103 For this analysis, we also include studies that use an emerging third option, pumps. These
104 allow for much larger volumes of water than grab samples but can be fitted with or convey water
105 through sieves, which allow them to capture smaller particles than typical net samples. The
106 sieves, tubing, and other pump components may themselves be made of plastic materials and
107 require a source of energy to power them in the field, making them a more challenging choice for
108 some sampling locations.

109 Several previous studies have reported dramatically different microplastic concentrations
110 from samples collected using differing methods¹⁷⁻²², as well as preliminary evidence to suggest
111 systematic trends^{16,23}. Here we take a wide and thorough look across the literature of surface
112 water studies, including those that pair methods and others that do not, to see how method choice

113 affects measured microplastic concentration. We then use the relationships uncovered to itemize
114 and quantify potential sources of systematic bias in sampling method.

115 The objective of this analysis is not to identify the best performing sampling method.
116 Each method is currently in use due to their own context-specific advantages. Our hope, instead,
117 is to shed light on the misalignment of the resulting concentration measurements and help move
118 the microplastics field one step closer to harmonizing methods and creating a comparable,
119 reliable body of literature for policymakers and researchers alike.

120

121 **2. Methods**

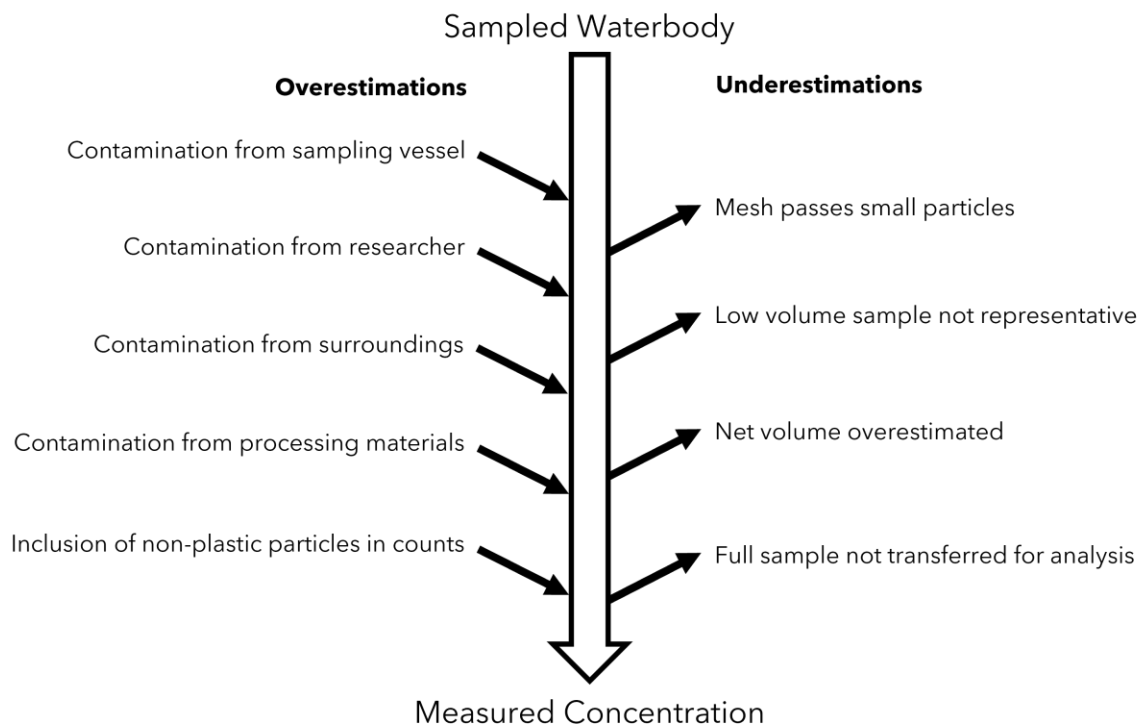
122 We rely on a multi-faceted approach to investigate potential biasing factors of
123 concentration measurement differences. These include (1) a systematic literature review of
124 surface water microplastic samples, (2) a closer look at samples collected in pairs of differing
125 methods, (3) controlled field studies and related works that isolate for particular potential biasing
126 factors, (4) statistical analysis and back-of-the-envelope calculations to identify reasonable
127 bounds on the magnitude or contribution of potential biasing factors.

128 *2.1 Literature review:*

129 We performed a literature search of surface water microplastic studies published prior to
130 October 2020. The review was conducted in September and October 2020. Studies were
131 retrieved from Google Scholar searches of the words: “microplastic” + “surface water”, along
132 with (individually) “net”, “pump”, “bulk”, “discrete” and “grab”. Of the returned results, we
133 included all studies that both sampled within the top 1m of a waterbody (deeper samples were
134 excluded) and reported volume sampled or a means of, at least roughly, calculating volume
135 sampled (e.g. net dimensions and tow distance or speed and time). This strategy of post-hoc

136 volume calculation accounted for about 1/3 of the included studies. For studies that sampled
137 multiple waterbodies or used multiple methods, results were included for each unique
138 combination of method and waterbody-type. For example, if multiple rivers in a region were
139 sampled with the same method, their results were averaged, while the results of pumping and net
140 methods on a single river were considered separate entries.

141 Additionally, we identified 15 datasets that measured microplastic concentrations using
142 paired samples of two or more methods (“paired-method”) at a single sampling time and
143 location. All but three of these studies, which were omitted due to insufficient data or
144 incompatible sampling depth, were also included in the analysis of literature-wide trends. One of
145 these datasets was collected specifically for this study (Section 2.2).



147

148 Figure 1. A conceptual diagram of the pathways that may increase (left) or decrease (right)

149 measured concentration, from the sampling of a waterbody to transferring and processing a

150 sample to the quantification of particles in the sample.

151

152 2.2 Field samples:

153 To include in the paired-method sample analysis with the forementioned published

154 datasets (n=14), we also collected paired grab and net samples in 4 streams (watershed areas:

155 35km, 73km, 101km, 320km) in Tompkins County, New York. These samples were filtered

156 through the equal size meshes to fill a gap in the literature of paired grab and net samples with

157 equivalent lower-size bounds.

158 We collected these samples across multiple flow conditions, sampling each river 1-3

159 times. A grab sample (mean volume: 1.8L) and a neuston net (10min deployment, 1m wide x

160 0.5m tall x 3m long, 0.335mm mesh; Sea-Gear, Melbourne, FL) were used sequentially to collect
161 microplastics at the surface in the region of highest flow in each river. In the lab, grab samples
162 were poured through a 0.335mm mesh to match the lower size constraint of the net samples.
163 Field data for these samples, as well as further laboratory processing details, particle
164 identification, Raman confirmation, and contamination reduction are included in the
165 supplementary information.

166 Alongside field samples, deionized water was run through each laboratory processing
167 step, including filtration, drying, digestion, separation, and counting, as procedural blanks to
168 measure potential contamination from laboratory materials. Air blanks were also collected by
169 exposing filter paper to laboratory air for 24 hours. Additionally, we collected a set of
170 “maximum reasonable procedural blanks”. These blanks were collected by passing deionized
171 water through single-rinsed mesh, sieves, and beakers. They were designed as “worst-case”
172 blank samples and were intended to quantify an upper-bound on “reasonable” potential
173 contamination levels to compare against concentration discrepancies across sampling method.
174 We collected these blanks after the completion of all laboratory work and after the lab space and
175 equipment had been used extensively for laboratory courses and demonstrations. Results of
176 blanks are included in Table S1. Average air and procedural blank values have been subtracted
177 from reported concentrations.

178

179 *2.3 Statistical Analysis*

180 We identified *a priori* a variety of potential factors influencing the concentration trends
181 observed through literature review and solicitation of hypotheses from field experts (Figure 1).
182 We use multiple linear regression as a tool to organize these hypotheses and identify which of

183 these reasonable factors may be more relevant in explaining concentration differences than
184 others. The regression included the following 6 factors: 1) sampled volume; 2) sampling method;
185 3) filtration or mesh size; 4) sampled waterbody (freshwater vs. marine); 5) whether visual
186 particle counts were confirmed with a more advanced technique to confirm polymer content; and
187 6) whether measured contamination was subtracted. To avoid problems of collinearity among
188 these predictor variables, the degree of correlation between variables was checked visually and
189 by examining the magnitude of their pairwise correlation. The multiple linear regression used to
190 ascertain the relationship between the response and predictor variables was run in R version
191 4.0.3²⁴, with statistical assumptions of normality and homogeneity of variance checked
192 graphically. To determine whether the percentage of fibers differed between paired-method
193 samples, a Wilcoxon signed-rank test was used. For all statistical tests, we used a p-value upper-
194 bound of 0.05 to determine statistical significance.

195 As we explore potential factors influencing concentration differences, we use a simple
196 equation (Equation 1) to determine a rough magnitude estimate for an additive factor, such as
197 contamination, that may be affecting measured concentration. Equation 1 provides a rough
198 estimate of the number of contaminating particles, or other additive factor, needed to equate two
199 paired-method sample concentrations:

200

$$201 \quad \frac{n_1 - k}{V_1} = \frac{n_2 - k}{V_2} \quad \text{Equation 1}$$

202

203 where n is the number of particles counted in the sample, k is the number of introduced particles
204 due to an additive factor (such as contamination), V is the volume of the sample, and subscripts
205 denote each sample of a pair.

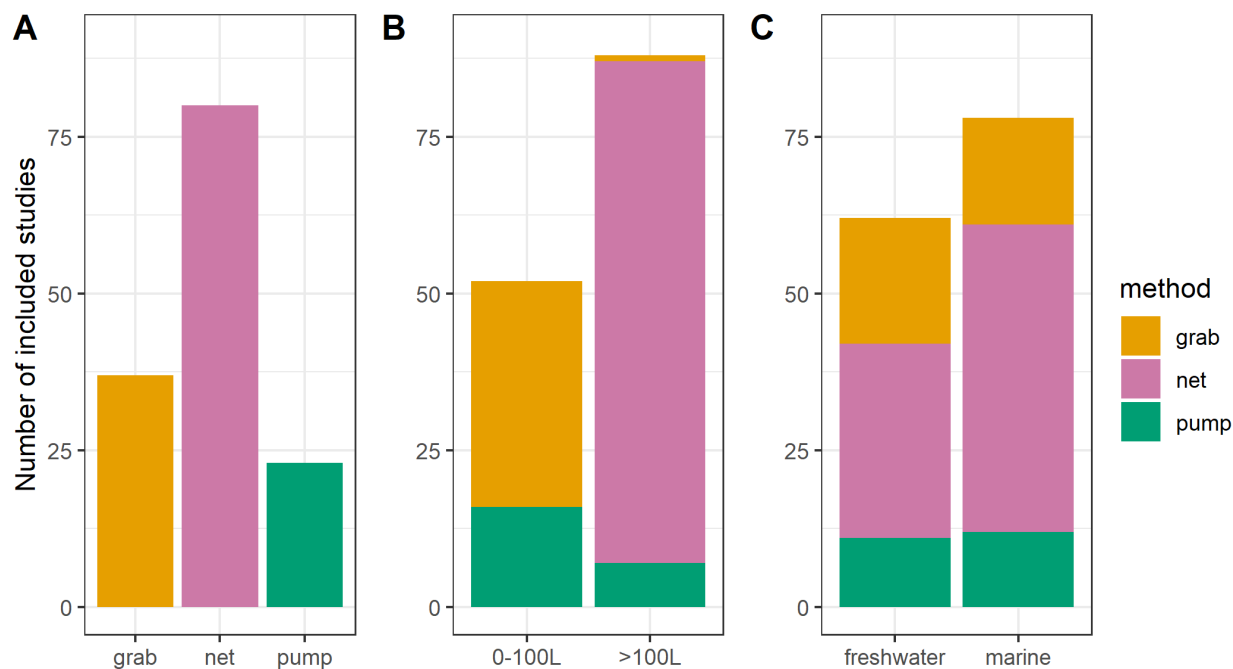
206 Equation 1 assumes that contamination affects samples collected and processed together
207 in a similar way. It also assumes that there is a true environmental concentration that would be
208 reported equivalently by any paired-method samples. This equation includes two major
209 simplifications: 1) that intersample variability is zero (we know side-by-side samples to vary up
210 to $9x^{25}$); and 2) that the number of introduced particles of contamination will be equal across all
211 samples (more precisely, k 's would be sampled from a given distribution). The equation
212 therefore represents the case where an additive effect, like contamination, is the sole factor
213 affecting concentration differences between measurements and volume the sole factor
214 influencing sampling intensity.

215

216 **3. Results and Discussion**

217 A total of 118 studies were included in the analysis of literature-wide trends. Due to
218 studies that include results from the use of more than one sampling method or sample more than
219 one type of waterbody, 140 unique entries were included (Figure 2). This total includes 37
220 instances of a grab method^{11,16–18,22,26–55}, 80 using a net method^{8,10,13,17,21,22,25,27–30,35,36,44,50,56–117},
221 and 23 of a pump method^{21,22,47,62,69,91,103,118–130} to collect their samples. Of the unique entries,
222 44% were freshwater (including 39 riverine and 22 limnic systems) and 56% were marine
223 (including 12 estuarine and 65 oceanic systems).

224



225

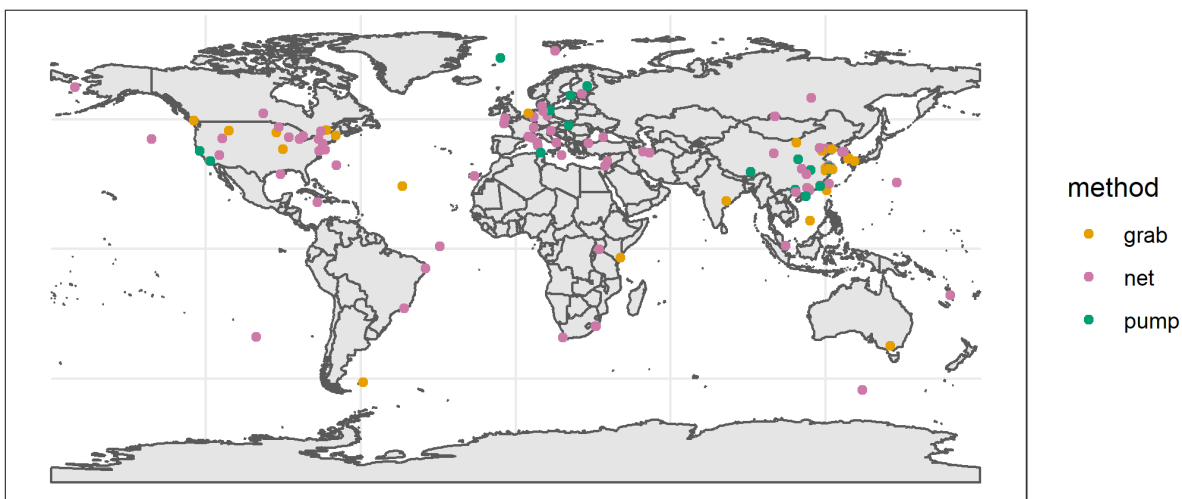
226 Figure 2. Summary of the unique entries included in this literature review, including sampling

227 method used (A), binned sample volume (B), and sampled waterbody type (C).

228

229 These studies span the globe (Figure 3). They also include samples from the 1970's,
 230 2000's and 2010's, with publication dates ranging from 1971-2020 (Figure S1). The studies also
 231 rely on a variety of laboratory techniques. Some use wet peroxide oxidation and density
 232 separations to first isolate particles, while others simply examine all contents of a sample.
 233 Fourier transform-infrared (FTIR), Raman, Nile Red staining and simple visual inspection were
 234 all represented.

235



236

237 Figure 3. Global¹³¹ distribution of samples included in this analysis.

238

239 To help narrow exploration into the main factors that may cause the systematic
 240 concentration differences, we use a multiple linear regression run on the overall dataset (Table
 241 1). Across the literature analyzed, volume sampled, mesh size, and waterbody sampled were
 242 significant predictors of measured concentration.

243

244 Table 1. Summary of coefficients for the multiple linear regression^a fit to the literature-wide data
 245 to predict \log_{10} of measured concentration.

Parameter	Estimate	Standard Error	t value	p-value
Intercept	1.04	0.28	3.75	$3 \cdot 10^{-4}$
$\text{Log}_{10}(\text{Volume})$	-0.50	0.10	-5.20	$9 \cdot 10^{-7}$
Method = Net	-1.10	0.44	-2.52	0.01
Method = Pump	-0.10	0.27	0.40	0.69
Mesh size	-1.51	0.67	-2.24	0.03
Waterbody = Marine	-0.35	0.17	-2.04	0.04
Confirmed polymer	0.20	0.21	0.99	0.33
Subtracted blanks	0.05	0.17	0.28	0.78

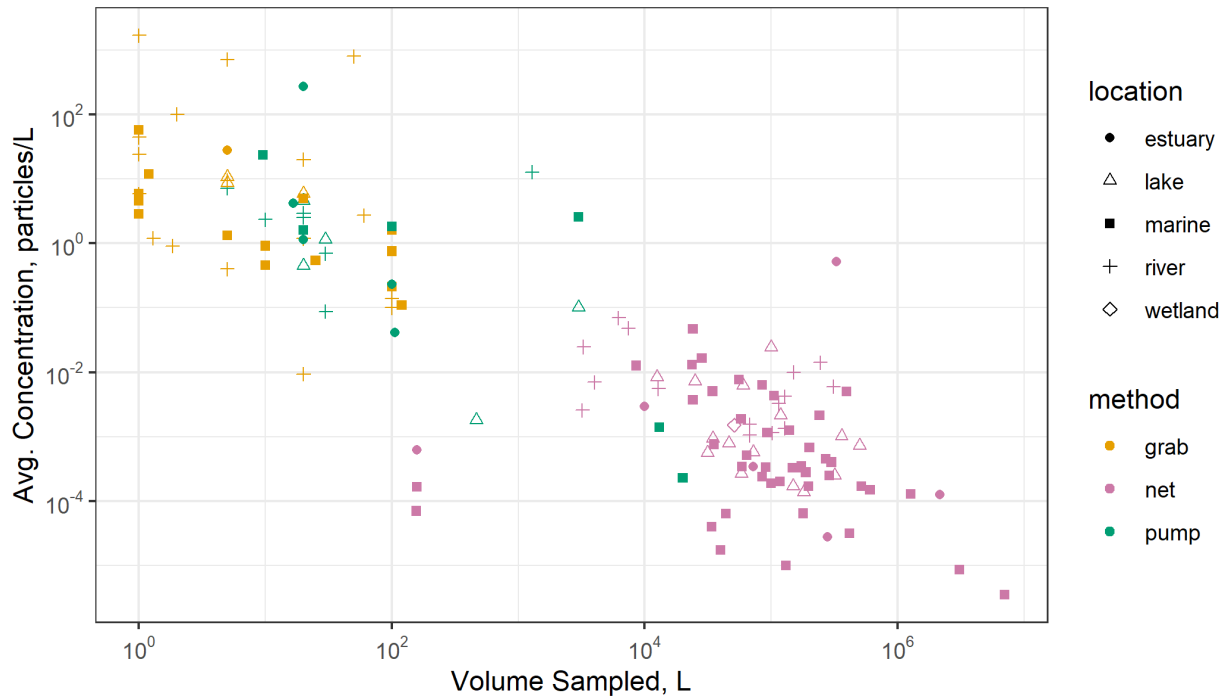
246

247

248

^aAdjusted R squared value for this regression model is 0.77, with an F-statistic of 69 on 7 & 132 degrees of freedom and a p-value of $2 \cdot 10^{-16}$

249 Volume sampled was the most significant predictor (Figure 4). Grab samples (10^0 - 10^2 L)
 250 systematically resulted in higher microplastic concentrations than net samples (10^2 - 10^7 L).
 251 Pumped samples (10^1 - 10^4 L) represented concentrations (10^{-4} - 10^2 L $^{-1}$) that overlapped with and
 252 fell between grab (10^{-3} - 10^3 L $^{-1}$) and net (10^{-6} - 10^{-1} L $^{-1}$) sample concentrations.
 253



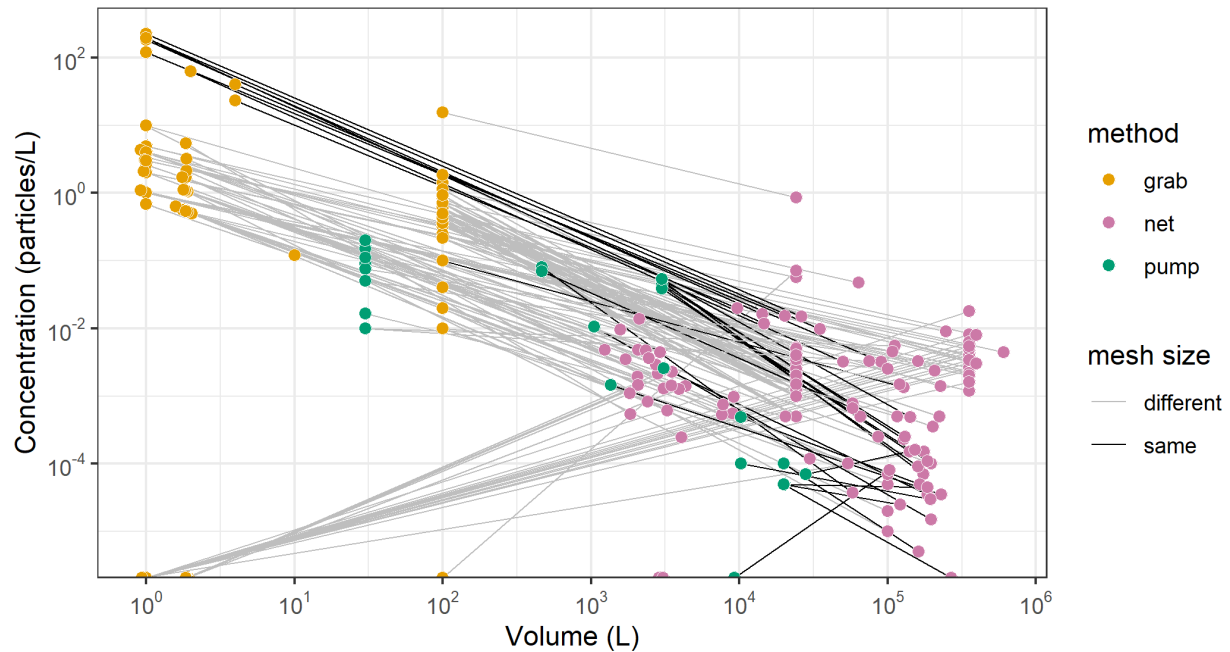
254
 255 Figure 4. Average volume sampled in studies relying on differing methods (color) and in
 256 differing waterbodies (shape) and the average concentration measured in each of those studies.

257
 258 Method and mesh size, though correlated with sample volume, were found to include
 259 enough independent information to also be significant factors in predicting concentration.
 260 Correlation between mesh size and volume, for example, as measured with Kendall's Rank
 261 Correlation, yields tau = 0.5. Enough variability exists in the relationship between volume and
 262 mesh size (Figure S3) that these factors can be examined independently. Pump and grab method

263 estimates were not statistically significant from one another but were both different from net
264 method estimates after accounting for all other factors (Table 1). For all methods, volume
265 appears to be the driving predictor of measured concentration, more so than the method itself
266 (Figure 4).

267 The regression also indicated that sampled waterbody type was a significant predictor,
268 with marine samples tending to measure lower concentrations. While the included net samples
269 do tend to be of marine environments (61%) and grab samples of freshwater environments
270 (54%), a further analysis of paired-method samples (Figure 5) highlights that even in the same
271 sampling environment, with the same anthropogenic pressures, the relationship between method
272 and concentration remains. The model fit suggested that visual identification and the use of
273 blanks, as implemented using current, highly variable methods, were not among the strongest
274 predictors of measured concentration.

275 Among the subset of paired-method studies ($n = 15$), which sampled at the same time and
276 location with differing methods (Figure 5), the same concentration trend is apparent: low volume
277 samples tend to measure orders of magnitude higher concentrations than high volume samples. A
278 few sample pairs (28 out of 310 paired-method samples) show the opposite trend, specifically
279 when smaller volume sample concentrations are zero, but we believe this to be a demonstration
280 of one of the shortcomings of small sample sizes: that they may miss particles altogether and
281 falsely report zero concentration due to undersampling the system. Koelmans et al. take note of
282 this shortcoming in their review and recommend a minimum sample volume in surface waters of
283 500L¹³². Replicates of low volume samples can also help mitigate this issue.



284

285 Figure 5. Paired-method samples from 15 datasets, including field data collected for this study.

286 Lines connect sample pairs collected at the same time and location. Black lines indicate pairs
 287 were filtered through the same mesh size, while gray lines indicate pairs used two different mesh
 288 sizes. Zero concentration is adjusted to 10^{-6} particles/L to account for log-scale limitations and is
 289 plotted along the x-axis.

290

291 One explanation of the volume-concentration relationship disproven by paired-method
 292 sample results is that researchers may be intentionally choosing to sample larger volumes when
 293 they visit areas where lower concentrations are anticipated. What these paired-method samples
 294 show instead is that even at the same time and location, higher sample volumes measure lower
 295 concentrations, just as in the literature-wide trend (Figure 4).

296 One important note is that samples containing a large volume of water naturally will
 297 contain more particles, thus diminishing the influence of any single particle captured on the
 298 overall mean per volume. For small sample volumes like 1L grab samples, concentration values

309 have a resolution of 1 particle/L. Contrastingly, for larger sample volumes, such as 10,000L net
300 samples, this resolution is much finer; each additional plastic particle would alter the
301 concentration by only 0.0001 particle/L. Larger volumes also have the effect of reducing
302 variability between samples¹³³.

303

304 *3.1 Potential contribution of mesh size differences*

305 Mesh size differences alone are insufficient to explain the orders of magnitude
306 differences in concentration alone. Paired-method studies that use the same mesh size still find
307 small volumes to measure lower concentrations than large volumes (black lines, Figure 5). When
308 filtered through the same size mesh, net samples measured an average of 10⁴-times lower
309 concentrations than those sampled by a paired grab and 10³-times lower concentrations than a
310 paired pump.

311 Net samples tend to use larger filtration sizes than grab and pump samples do. This is
312 largely an intentional design choice to avoid clogging. However, only a few mesh sizes are
313 commonly used for sampling nets, which allows volume, which varies widely, to remain
314 somewhat independent of mesh size within a given sampling method and, therefore, be examined
315 separately (Figure S3). For example, the studies from our broader literature review that sampled
316 using a net with 300-350 μ m mesh still show a strong volume-concentration relationship (Figure
317 S4). In contrast, when looking only at grab samples with 1L sample volume, there is no evident
318 mesh size – concentration relationship (Figure S5).

319 To some extent, mesh size certainly does matter: when you allow smaller particles to be
320 in your sample, you will likely collect more particles overall^{21,62,134,135}. Smaller volume samples,
321 specifically grab and some pump samples, allow for a smaller mesh or filtration size to be used

322 without the issue of clogging. In the included studies, grab sample mesh or filtration sizes ranged
323 from 0.4-335 μm , pump samples from 4-300 μm and net samples from 50-947 μm .

324 Existing field measurements in the literature have quantified the change in measured
325 concentration when volume is controlled for while varying mesh sizes (Table S2). In the case of
326 Lindeque et al., 100 μm mesh measured 10x higher concentration than 500 μm mesh¹³⁴. Based on
327 the literature, Koelmans et al. develop a concentration conversion factor to account for mesh size
328 differences¹³⁶. At its maximum, to convert from a measured particle size range of 333-5000 μm ,
329 as is common to net samples, to a broader range of 1-5000 μm , the calculated conversion is a
330 factor of 40. Paired-method samples in our analysis, where at a given time and location samples
331 of differing methods are collected, measure concentrations that differ by 2 orders of magnitude
332 or more. Accounting for mesh size is therefore an important, but insufficient step in rectifying
333 measured concentration differences between methods.

334

335 *3.2 Potential contribution of overestimated net volumes*

336 One possible explanation for net samples measuring lower concentrations is due to how
337 sample volumes are measured. Grab sample and some pumped sample volumes can be precisely
338 measured based on the sampling vessel. In contrast, net samples, and some pump sampling
339 techniques, require calibrated flowmeters for accurate sample volume measurement. Without
340 one, net volumes are prone to overestimation. Overestimated sample volumes result in measured
341 concentrations lower than true system concentrations.

342 Karlsson et al. found that net sample volumes calculated without a flowmeter incorporate
343 a volume error of at least 1%¹²². They observed that the water level in the net mouth fluctuates
344 during towing, making sampled depth an inconsistent metric. One in three studies included in our

345 analysis lacked flowmeter results and required us to calculate volume sampled, by relying on
346 average mouth depth and a given boat speed or GPS distance. We found that calculated volumes
347 did have a steeper volume-concentration relationship than studies with volumes given (Figure
348 S6); however, removing studies that required ad-hoc volume calculations did not affect the
349 predictors included in a best-fitting regression model.

350 Another way that a net volume calculation can be inaccurate is due to fluid dynamic
351 principles, which result in water bypassing the net due to flow resistance (drag) from the mesh
352 itself. A typical strategy for calculating volume is to multiply a tow length by net dimensions (for
353 river samples, tow length is time of deployment multiplied by river velocity). This provides a
354 theoretical volume that ignores drag, assuming no water bypasses the net. The relationship
355 between actual volume sampled and theoretical volume sampled is known as “filtration
356 efficiency”. This factor can change dramatically even for the same equipment. It is affected by
357 the speed at which water is being forced through the net, the mesh size and the abundance of
358 biological material in the sampled water body.

359 At a filtration efficiency of 85%, which is an acceptable value in plankton tows¹⁴,
360 measured concentration would be underestimated from “true” system concentration by 15%.
361 While noteworthy, this percent decrease is dwarfed by the differences observed in the paired-
362 method studies analyzed, where net concentrations were 75-100% lower than grab concentration
363 and 45-100% lower than pump concentrations. This indicates filtration efficiency, while
364 important, cannot solely explain the concentration differences observed.

365

366 *3.3 Potential for fiber loss between sampling and processing*

367 It is aspirational to assume that all particles that enter through the net are captured and
368 collected in the cod-end. Likely some particles, fibers especially, may be trapped in the mesh
369 itself or pass through the net entirely and return to the surrounding waters. The majority of
370 included studies found fibers to be the most prevalent particle-type captured, followed by
371 fragments. When examined by method, however, fibers' dominance was only true for grab and
372 pump samples; in the majority of net samples, fragments were the most prevalent particle-type
373 (Figure S7). This points to one of two potential hypotheses. The first is that a significant portion
374 of fibers are being lost from the net. Lusher et al. provide evidence for this by putting sieves in
375 series and discovering particles in secondary and tertiary sieves, an indication that some number
376 slip through a primary sieve¹³⁷. Another way fibers may be lost from net samples is if they are
377 captured during sampling, but not transferred to the vessel processed in the lab. Too few studies
378 have looked for and quantified residual microplastic particles in the net mesh for this work to
379 investigate whether lower net concentrations could be caused by this kind of particle loss. We
380 encourage future studies to examine net mesh before and after sampling to add to this body of
381 knowledge. The second hypothesis relies on the observation that contaminating particles are
382 largely fibers^{22,77}. If small-volume sample counts are dominated by contamination, perhaps
383 fibers' dominance in grab and pump samples reflects contamination and not environmental
384 conditions. Without knowing the true, relative prevalence of particle-types in the sampled
385 environment and because this meta-analysis looks at studies from across the globe, we cannot
386 confirm either hypothesis with this dataset.

387 Analysis of the paired-method data within this study, where differing methods are being
388 sampled from the same environment, finds no significant evidence of fiber loss from net

389 samples: paired studies that included particle-type showed statistically similar percentages of
390 fibers between samples of differing method (Wilcoxon signed-rank test, p-value = 0.37). Perhaps
391 this points to an additional interaction where fibers are not only lost from the net, but also gained
392 in similar quantities when fibers small enough to escape through net mesh are actually retained²².
393 Controlled field studies will be needed to fully understand and quantify fiber losses and gains
394 through mesh.

395

396 *3.4 Potential contribution of intersample variability*

397 Given that all waterbodies are heterogeneous to some extent, it is reasonable to assume
398 that no two water samples will hold the exact same contents. For this reason, one suggested
399 explanation for differing concentrations holds that it is actually the replication and not
400 necessarily the methods themselves that create the variance observed in paired grab-net studies.
401 To some extent, this is disproven by the systematic differences observed across unpaired studies
402 of varying methods (Figure 4).

403 To investigate whether this may, however, play even a minor role in the differences observed
404 in paired-method studies, we rely on existing studies which have measured the concentration
405 variance between replicate studies of the same sampling method. Lindeque et al. towed two nets
406 (0.333mm mesh-size manta trawls) in parallel and found no significant difference between the
407 measured concentrations (0.54 and 0.46 microplastics m⁻³)¹³⁴. Schmidt et al. found triplicate net
408 samples taken within 2 hours of each other varied up to 9x²⁵. Hung et al. found duplicate net
409 samples had a standard deviation of less than 15%, while duplicate grab samples varied by 2x²².
410 From this evidence, we conclude that heterogeneity plays only a negligible role in the multiple

411 orders of magnitude concentration differences observed among methods reported in this analysis
412 (e.g., Figure 4).

413 Another possible influence of heterogeneity is in patchy distribution of particles at the
414 sampling site, where researchers may be selecting for higher concentration areas. In rivers, for
415 example, many researchers choose to sample the thalweg, but in smaller streams, a net may
416 sample additional flow outside of the true thalweg, while a pump or grab sample would be able
417 to sample more narrowly from only that zone. This may result in higher concentrations from
418 more focused sampling methods. Additional investigations into the patchiness the distribution of
419 particles at various sampling locations will help quantify the role of location selection in the
420 volume-concentration relationship.

421

422 *3.5 Potential contribution of contamination*

423 Contamination as an explanation fits the systematic concentration differences observed
424 due to the relationship between count, volume, and concentration. When a count is inflated in a
425 small volume sample, the concentration is affected much more dramatically than if the count of a
426 larger volume sample were inflated by the same number of particles.

427 A wide range of approaches were used to measure contamination. Of the datasets
428 included in the literature-wide review, 28% neglected to run or report any blank samples
429 alongside field samples. Less than 5% measured for contamination throughout the sampling
430 processing (including from field instruments, ambient air, and laboratory processing methods
431 and supplies). The most common type of blanks run was “procedural”, or “method”, blanks,
432 where some quantity of filtered water was run through laboratory equipment and processes in
433 parallel with field samples. Beyond the measurement of blanks, strategies for reporting and

434 accounting for the measured contamination varied widely. Of those that measured blanks, 16%
435 failed to report how many particles were found during the process and only 28% removed
436 contamination, if found, from reported concentrations. This is an improvement from previous
437 reviews, such as Hanvey et al., who found only 7% of the microplastic studies included
438 procedural blanks¹³⁸.

439 Despite the increasing prevalence of measuring contamination in the laboratory
440 processes, not all potential pathways are being quantified. For example, field blanks are still
441 uncommon²². As one rare example, Ryan et al. used a neuston net fitted with mesh at its mouth
442 to exclude introduced particles and still captured 28 particles ($0.1/\text{m}^3$), assumed to be originating
443 from the plastic net itself¹³⁹. The contribution of sampler's clothing, pump tubing, or plastic lids
444 on grab sample vessels are all still potential sources that require more investigation. Other
445 laboratory-based sources of error are possible and understudied as well. Recent work by Witzig
446 et al. indicates that even plastic gloves used for personal protection during lab work may be
447 contributing to an overestimation of sample polymer content¹⁴⁰.

448 Inflated counts unrelated to contamination are also a concern. For instance, when visual
449 counts are used to calculate concentrations, counts are often inflated by organic materials
450 mistaken as plastics. 23% of the studies included in our overall literature analysis did not use any
451 advanced microscopy or material identification methods to confirm polymer content of counted
452 particles. As an example of the shortcoming of visual counting technique, Lenz et al. visually
453 identified 1279 items as plastic but found through Raman spectroscopy that only 64% matched
454 known polymer signals¹⁴¹. While visual overestimation would influence concentrations in a
455 volume-sensitive way, we do not see systematic differences between studies that confirm particle

456 material. Additional unexpected pathways of inflated counts, false positives and contamination
457 should be an emphasis of future work.

458 Contamination in the laboratory is typically minimal, but regularly present. It derives
459 from fibers settling out of laboratory air (in our own 24-hour air blanks, we detected an average
460 of 6 particles, all fibers), contamination of reagents, and particles in or on the variety of
461 equipment and containers that typical multi-step processing requires. Procedural contamination
462 would be consistent across all sample types run in a lab, but the same number of introduced
463 particles would alter the concentration of a small volume sample to a greater extent than of a
464 large volume sample.

465 In the literature, reported contamination ranges from zero particles in a blank to dozens.
466 The actual number of particles measured in a blank likely depends not only on laboratory
467 protocols, but also on the volume of water processed for a given blank, the duration of
468 processing and the number of vessel transfers performed. It also is highly dependent on exactly
469 what parts of the sampling, processing, and counting the blank undergoes.

470 Because of these inputs, it is difficult to compare blank values across studies directly.
471 More commonly, they are put in the context of sample counts. For example, while Cable et al.
472 measured an average of 42 particles in three blanks, mean sample counts in their high volume net
473 samples ranged from 8 particles to 17,146 particles⁷⁷. Similarly Scircle et al. detected an average
474 of 35 particles in nine procedural blank samples, compared against particle counts within grab
475 samples that ranged from 0 to 151 particles³¹. Hung et al. chose to omit all pumped samples from
476 their analysis because of how similar blank and measured particle counts were (287 blank
477 particles vs. 192 sample particles)²². For context, when we attempted to create and measure a

478 highest reasonable bound of procedural contamination by avoiding the careful cleaning and
479 protections typical across the literature, we measured as many as 66 particles.

480

481 *3.6 Lessons from a related field: plankton population research*

482 Much of the sampling methods used for microplastics were adapted from plankton
483 sampling. There are many parallels in terms of particle shape, size and distribution between the
484 two sample targets. The results of plankton studies that perform similar paired-method
485 comparisons, on plankton concentrations instead of plastic ones, report mixed results. Some,
486 such as Cada and Loar, find no difference between ichthyoplankton (4-10mm) densities sampled
487 with net (100,000L) or with pump (16,700L) despite the pumped samples allowing smaller
488 particles¹⁴². While ichthyoplankton differ from microplastics in that they are able to actively avoid
489 net capture, this comparison took place at night when avoidance is minimal. Others, such as
490 Masson et al., report zooplankton (>0.053mm) concentrations being somewhat, though not
491 statistically, higher when sampled with a pump (2-20L) vs. nets (10-220L) of the same mesh
492 size¹⁴³. And still others, such as Appel, found about two orders of magnitude higher
493 concentration for zooplankton (>0.061mm) collected pumps (12L) or grab samples (2L) as
494 opposed to those collected with nets (5,000-11,500L)¹⁴⁴.

495 We were unable to find any plankton method comparison studies with orders of
496 magnitude concentration differences comparable to those we see in microplastics research (Table
497 2). This suggests the concentration differences in microplastic research are largely from factors
498 unique to plastics. Contamination is one such explanation that fits. It is, for example, much easier
499 to discern between zooplankton and lake debris than between a sampled plastic particle and a
500 contaminating one. More targeted research is required to know for certain whether the

501 contributing factor truly is more easily concealed contamination, unique interactions with
 502 sampling equipment or another factor not yet identified.

503

504 Table 2. (A) A summary of the concentration ranges observed across the synthesized literature in
 505 this study, as well as (B) the observed and calculated concentration differences produced by
 506 potential biasing factors.

507

(A)	
Method	Measured Concentration (particles/L)
Grab	$9.3 \cdot 10^{-3} - 1.7 \cdot 10^3$
Pump	$2.3 \cdot 10^{-4} - 2.7 \cdot 10^2$
Net	$3.5 \cdot 10^{-6} - 5.1 \cdot 10^{-1}$

(B)	
Potential Biasing Factor	Orders of magnitude explained ^{a,b}
Mesh/filtration size ^{21,62,103,134,135}	0-10 ²
Net Volume Overestimation ^{14,122}	0-10 ¹
Particles that enter net not captured in sample	Insufficient data
Intersample variability ^{22,25,134}	0-10 ¹
Contamination ^c	0-10 ³

508 ^a Note: Values included for each biasing factor are not necessarily independent. Each assumes the entire
 509 observed concentration difference is due to a single factor, when in reality, no study method fully isolates
 510 for the tested factor. For example, concentration differences from two side-by-side samples may be driven
 511 by the patchiness of the sampled waterbody, but may also be driven by contamination additionally.

512 ^b Values are the ratio of concentrations from paired-method samples collected at same time and location
 513 from various published studies.

514 ^c Calculated using Equation 1 on paired-method samples included in (A) to find concentration differences
 515 that could be accounted for with a reasonable k (i.e. $k < \text{sample count}$).

516

517 3.7 Assessment

518 We use Equation 1 to find the value of k that explains the difference in concentrations for
 519 paired-method studies (Figure 5). We start by looking only at reported values not yet corrected
 520 by blank measurements. We find that for the majority of published, non-blank-corrected paired-
 521 method studies, the introduction of only a few particles can explain the difference between grab
 522 and net concentrations (median: 3.4, mean±standard error: 39±1.4) and between pump and net

523 concentrations (median: 3.9, mean±standard error: 36±1.9). These values for the theoretical
524 number of introduced particles (k), even at their highest, are well within the range of values
525 reported in the literature (Section 3.5). The skewed results for k , however, reinforces the
526 observation that the number of introduced particles varies substantially among studies.

527 For a more study-specific test of our contamination-alone assumption of Equation 1 and
528 to assess whether k is reasonable within individual studies, we focus on 11 of the paired-method
529 studies that both ran blanks and report the number of particles found in those blanks. For each
530 study, we compare the particle counts measured in blanks run within the given study against the
531 theoretical number of introduced particles (k) needed to satisfy Equation 1. For the seven grab-
532 net studies and the two pump-net studies with available blank counts, theoretical contamination
533 differed from actual measured blank counts by less than one particle (an average of 0.57 particles
534 and 0.60 particles, respectively). These preliminary values indicate contamination alone (or in
535 conjunction with another additive affect) can explain nearly all of the observed concentration
536 differences observed between samples of differing methods and volumes. It also suggests,
537 however, that current contamination quantification methods are not universally sufficient for
538 identifying and removing contamination introduced into each sample, given studies like Hung et
539 al., which remove a standard blank count from sample counts and still find incompatible
540 concentrations²².

541 A combination of the examined factors, including contamination, could also be at play.
542 Though the values included in Table 2B are not fully independent of each other, in sum and at
543 their extreme, they can cumulatively account for the full concentration discrepancies observed.
544 To determine with certainty the factors at play and identify adequate methodological

545 interventions to correct for them, these biasing factors must be isolated further through targeted
546 research.

547

548 *3.8 Recommendations*

549 Differentiating between plastics from environmental samples and from contamination is
550 impossible with current methods, which makes precautions to avoid contamination at all times
551 and measuring blanks throughout processing imperative to reliable results.

552 Based on limited existing data, we can recommend that blanks be (1) run repeatedly
553 throughout the processing of a pool of related samples, (2) run through all items and spaces in
554 contact with samples, including mesh and steps completed in the field (3) adjusted, when
555 reported, for relevance to sampled volumes, exposure times, and particle counts, and (4)
556 thoroughly described such that a true “methodological peer” can be identified by future studies
557 for concentration comparisons.

558 Cross-study or multi-method comparisons and compilations should be avoided when
559 possible, until specific experiments can be performed to isolate and remedy the systematic
560 differences in concentration observed. This has broader implications in terms of policy decisions
561 that rely on a compilation of various studies; describing regional trends from a combination of
562 individual studies or creating forecasting models based on disparate studies is a risky endeavor at
563 this time. We also encourage study designs that allow relative abundance comparisons within a
564 sampling campaign, as this analysis strategy can control for biasing factors and avoid misleading
565 inter-study concentration comparisons.

566 Until standard methods for contamination quantification are developed, we recommend
567 large sampling volumes be used, regardless of method choice, to mitigate the influence particle

568 count inflation can have on overall sampling volume. We, unfortunately, were unable to detect a
569 volume threshold above which samples were unaffected, and thus, we are unable to recommend
570 a specific volume. Correcting for mesh size¹³⁶, collecting repeat samples, selecting sampling
571 sites randomly, and confirming visual counts with advanced techniques are all important steps to
572 accurate microplastic quantification, as well. We remain hopeful that researchers and citizen
573 scientists will continue to be able to use sampling methods that best suit their needs so long as
574 proper corrections, considerations, and contamination quantification protocols are followed.

575

576 **Acknowledgements**

577 The authors would like to acknowledge the important contributions of Susan McGrattan, Anna-
578 Katharina von Krauland, Alexis Weaver, Gray Ryan, Emma Mosier, Whitney Denison, Elizabeth
579 Dean, Leah Balkin, Jack Novak, and Xiaoman (Sharon) Zhang whose explorations in the field
580 and lab laid the groundwork for this study. This work was bolstered by helpful discussions with
581 attendees of AGU 2019 and MICRO2020, as well as with Dr. Nelson Hairston, Dr. Bruce
582 Monger, Dr. Todd Cowen and Dr. Lars Rudstam, bolstered this work. Statistical support from
583 Dr. Jack Hessel was instrumental in the analysis. Funding: Lisa Watkins was supported by the
584 National Science Foundation Graduate Research Fellowship under Grant No. 2017228528. This
585 work made use of Cornell Center for Materials Research Shared Facilities which are supported
586 through the NSF MRSEC program DMR-1719875.

587

588 **Supporting Information**

589 1) Document of details on field methods used and blank sample measurements, as well as
590 additional visualizations of the synthesized data from this review (all included tables and figures
591 are referenced directly in text). 2) Data used for this study in zipped folder, including our own

592 field sample results and a spreadsheet of all studies and attributes used for this meta-analysis.
593 This information is available free of charge via the Internet at <http://pubs.acs.org>.

- 595 (1) Allen, S.; Allen, D.; Phoenix, V. R.; Le Roux, G.; Durántez Jiménez, P.; Simonneau, A.;
596 Binet, S.; Galop, D. Atmospheric Transport and Deposition of Microplastics in a Remote
597 Mountain Catchment. *Nat. Geosci.* **2019**, *12* (5), 339–344. [https://doi.org/10.1038/s41561-](https://doi.org/10.1038/s41561-019-0335-5)
598 019-0335-5.
- 599 (2) Peng, X.; Chen, M.; Chen, S.; Dasgupta, S.; Xu, H.; Ta, K.; Du, M.; Li, J.; Guo, Z.; Bai, S.
600 Microplastics Contaminate the Deepest Part of the World’s Ocean. *Geochem. Perspect.*
601 *Lett.* **2018**, 1–5. <https://doi.org/10.7185/geochemlet.1829>.
- 602 (3) Li, J.; Liu, H.; Paul Chen, J. Microplastics in Freshwater Systems: A Review on
603 Occurrence, Environmental Effects, and Methods for Microplastics Detection. *Water Res.*
604 **2018**, *137*, 362–374. <https://doi.org/10.1016/j.watres.2017.12.056>.
- 605 (4) Rochman, C. M.; Tahir, A.; Williams, S. L.; Baxa, D. V.; Lam, R.; Miller, J. T.; Teh, F.-
606 C.; Werorilangi, S.; Teh, S. J. Anthropogenic Debris in Seafood: Plastic Debris and Fibers
607 from Textiles in Fish and Bivalves Sold for Human Consumption. *Sci. Rep.* **2015**, *5* (1),
608 14340. <https://doi.org/10.1038/srep14340>.
- 609 (5) Critchell, K.; Hoogenboom, M. O. Effects of Microplastic Exposure on the Body
610 Condition and Behaviour of Planktivorous Reef Fish (*Acanthochromis Polyacanthus*).
611 *PLOS ONE* **2018**, *13* (3), e0193308. <https://doi.org/10.1371/journal.pone.0193308>.
- 612 (6) Foley, C. J.; Feiner, Z. S.; Malinich, T. D.; Höök, T. O. A Meta-Analysis of the Effects of
613 Exposure to Microplastics on Fish and Aquatic Invertebrates. *Sci. Total Environ.* **2018**,
614 *631–632*, 550–559. <https://doi.org/10.1016/j.scitotenv.2018.03.046>.
- 615 (7) Rios, L. M.; Moore, C.; Jones, P. R. Persistent Organic Pollutants Carried by Synthetic
616 Polymers in the Ocean Environment. *Mar. Pollut. Bull.* **2007**, *54* (8), 1230–1237.
617 <https://doi.org/10.1016/j.marpolbul.2007.03.022>.
- 618 (8) Wang, W.; Ge, J.; Yu, X. Bioavailability and Toxicity of Microplastics to Fish Species: A
619 Review. *Ecotoxicol. Environ. Saf.* **2020**, *189*, 109913.
620 <https://doi.org/10.1016/j.ecoenv.2019.109913>.
- 621 (9) Rezania, S.; Park, J.; Md Din, M. F.; Mat Taib, S.; Talaiekhosani, A.; Kumar Yadav, K.;
622 Kamyab, H. Microplastics Pollution in Different Aquatic Environments and Biota: A
623 Review of Recent Studies. *Mar. Pollut. Bull.* **2018**, *133*, 191–208.
624 <https://doi.org/10.1016/j.marpolbul.2018.05.022>.
- 625 (10) Zhang, K.; Shi, H.; Peng, J.; Wang, Y.; Xiong, X.; Wu, C.; Lam, P. K. S. Microplastic
626 Pollution in China’s Inland Water Systems: A Review of Findings, Methods,
627 Characteristics, Effects, and Management. *Sci. Total Environ.* **2018**, *630*, 1641–1653.
628 <https://doi.org/10.1016/j.scitotenv.2018.02.300>.
- 629 (11) Crew, A.; Gregory-Eaves, I.; Ricciardi, A. Distribution, Abundance, and Diversity of
630 Microplastics in the Upper St. Lawrence River. *Environ. Pollut.* **2020**, 113994.
631 <https://doi.org/10.1016/j.envpol.2020.113994>.
- 632 (12) Buchanan, J. B. Pollution by Synthetic Fibres. *Mar. Pollut. Bull.* **1971**, *2* (2), 23.
633 [https://doi.org/10.1016/0025-326X\(71\)90136-6](https://doi.org/10.1016/0025-326X(71)90136-6).
- 634 (13) Carpenter, E. J.; Smith, K. L. Plastics on the Sargasso Sea Surface. *Science* **1972**, *175*
635 (4027), 1240–1241. <https://doi.org/10.1126/science.175.4027.1240>.
- 636 (14) Jacobs, F.; Grant, G. C. *Guidelines for Zooplankton Sampling in Quantitative Baseline*
637 *and Monitoring Programs*; Ecological Research Series; Special Scientific Report 83;

- 638 United States Environmental Protection Agency: Virginia Institute of Marine Science,
639 1978; p 61.
- 640 (15) Klein, S.; Dimzon, I. K.; Eubeler, J.; Knepper, T. P. Analysis, Occurrence, and
641 Degradation of Microplastics in the Aqueous Environment. In *Freshwater Microplastics :
642 Emerging Environmental Contaminants?*; Wagner, M., Lambert, S., Eds.; The Handbook
643 of Environmental Chemistry; Springer International Publishing: Cham, 2018; pp 51–67.
644 https://doi.org/10.1007/978-3-319-61615-5_3.
- 645 (16) Covernton, G. A.; Pearce, C. M.; Gurney-Smith, H. J.; Chastain, S. G.; Ross, P. S.;
646 Dower, J. F.; Dudas, S. E. Size and Shape Matter: A Preliminary Analysis of Microplastic
647 Sampling Technique in Seawater Studies with Implications for Ecological Risk
648 Assessment. *Sci. Total Environ.* **2019**, *667*, 124–132.
649 <https://doi.org/10.1016/j.scitotenv.2019.02.346>.
- 650 (17) Barrows, A. P. W.; Neumann, C. A.; Berger, M. L.; Shaw, S. D. Grab vs. Neuston Tow
651 Net: A Microplastic Sampling Performance Comparison and Possible Advances in the
652 Field. *Anal. Methods* **2017**, *9* (9), 1446–1453. <https://doi.org/10.1039/C6AY02387H>.
- 653 (18) Green, D. S.; Kregting, L.; Boots, B.; Blockley, D. J.; Brickle, P.; da Costa, M.; Crowley,
654 Q. A Comparison of Sampling Methods for Seawater Microplastics and a First Report of
655 the Microplastic Litter in Coastal Waters of Ascension and Falkland Islands. *Mar. Pollut.
656 Bull.* **2018**, *137*, 695–701. <https://doi.org/10.1016/j.marpolbul.2018.11.004>.
- 657 (19) Karlsson, T. M.; Kärrman, A.; Rotander, A.; Hassellöv, M. Comparison between Manta
658 Trawl and in Situ Pump Filtration Methods, and Guidance for Visual Identification of
659 Microplastics in Surface Waters. *Environ. Sci. Pollut. Res.* **2019**.
660 <https://doi.org/10.1007/s11356-019-07274-5>.
- 661 (20) McEachern, K.; Alegria, H.; Kalagher, A. L.; Hansen, C.; Morrison, S.; Hastings, D.
662 Microplastics in Tampa Bay, Florida: Abundance and Variability in Estuarine Waters and
663 Sediments. *Mar. Pollut. Bull.* **2019**, *148*, 97–106.
664 <https://doi.org/10.1016/j.marpolbul.2019.07.068>.
- 665 (21) Tamminga, M.; Stoewer, S.-C.; Fischer, E. K. On the Representativeness of Pump Water
666 Samples versus Manta Sampling in Microplastic Analysis. *Environ. Pollut.* **2019**, *254*,
667 112970. <https://doi.org/10.1016/j.envpol.2019.112970>.
- 668 (22) Hung, C.; Klasios, N.; Zhu, X.; Sedlak, M.; Sutton, R.; Rochman, C. M. Methods Matter:
669 Methods for Sampling Microplastic and Other Anthropogenic Particles and Their
670 Implications for Monitoring and Ecological Risk Assessment. *Integr. Environ. Assess.
671 Manag.* **2020**, *00* (00). <https://doi.org/10.1002/ieam.4325>.
- 672 (23) Prata, J. C.; da Costa, J. P.; Duarte, A. C.; Rocha-Santos, T. Methods for Sampling and
673 Detection of Microplastics in Water and Sediment: A Critical Review. *TrAC Trends Anal.
674 Chem.* **2019**, *110*, 150–159. <https://doi.org/10.1016/j.trac.2018.10.029>.
- 675 (24) R Core Team. *R: A Language and Environment for Statistical Computing*; R Foundation
676 for Statistical Computing: Vienna, Austria, 2020.
- 677 (25) Schmidt, N.; Thibault, D.; Galgani, F.; Paluselli, A.; Sempéré, R. Occurrence of
678 Microplastics in Surface Waters of the Gulf of Lion (NW Mediterranean Sea). *Prog.
679 Oceanogr.* **2018**, *163*, 214–220. <https://doi.org/10.1016/j.pocean.2017.11.010>.
- 680 (26) Barrows, A. P. W.; Christiansen, K. S.; Bode, E. T.; Hoellein, T. J. A Watershed-Scale,
681 Citizen Science Approach to Quantifying Microplastic Concentration in a Mixed Land-
682 Use River. *Water Res.* **2018**, *147*, 382–392. <https://doi.org/10.1016/j.watres.2018.10.013>.

- 683 (27) Chae, D.-H.; Kim, I.-S.; Kim, S.-K.; Song, Y. K.; Shim, W. J. Abundance and Distribution
684 Characteristics of Microplastics in Surface Seawaters of the Incheon/Kyeonggi Coastal
685 Region. *Arch. Environ. Contam. Toxicol.* **2015**, *69* (3), 269–278.
686 <https://doi.org/10.1007/s00244-015-0173-4>.
- 687 (28) Kang, J.-H.; Kwon, O. Y.; Lee, K.-W.; Song, Y. K.; Shim, W. J. Marine Neustonic
688 Microplastics around the Southeastern Coast of Korea. *Mar. Pollut. Bull.* **2015**, *96* (1),
689 304–312. <https://doi.org/10.1016/j.marpolbul.2015.04.054>.
- 690 (29) Tamminga, M.; Hengstmann, E.; Fischer, E. K. Microplastic Analysis in the South Funen
691 Archipelago, Baltic Sea, Implementing Manta Trawling and Bulk Sampling. *Mar. Pollut.*
692 *Bull.* **2018**, *128*, 601–608. <https://doi.org/10.1016/j.marpolbul.2018.01.066>.
- 693 (30) Kapp, K. J.; Yeatman, E. Microplastic Hotspots in the Snake and Lower Columbia Rivers:
694 A Journey from the Greater Yellowstone Ecosystem to the Pacific Ocean. *Environ. Pollut.*
695 **2018**, *241*, 1082–1090. <https://doi.org/10.1016/j.envpol.2018.06.033>.
- 696 (31) Scircle, A.; Cizdziel, J. V.; Missling, K.; Li, L.; Vianello, A. Single-Pot Method for the
697 Collection and Preparation of Natural Water for Microplastic Analyses: Microplastics in
698 the Mississippi River System during and after Historic Flooding. *Environ. Toxicol. Chem.*
699 **2020**, *39* (5), 986–995. <https://doi.org/10.1002/etc.4698>.
- 700 (32) Barrows, A. P. W.; Cathey, S. E.; Petersen, C. W. Marine Environment Microfiber
701 Contamination: Global Patterns and the Diversity of Microparticle Origins. *Environ.*
702 *Pollut.* **2018**, *237*, 275–284. <https://doi.org/10.1016/j.envpol.2018.02.062>.
- 703 (33) Leslie, H. A.; Brandsma, S. H.; van Velzen, M. J. M.; Vethaak, A. D. Microplastics En
704 Route: Field Measurements in the Dutch River Delta and Amsterdam Canals, Wastewater
705 Treatment Plants, North Sea Sediments and Biota. *Environ. Int.* **2017**, *101*, 133–142.
706 <https://doi.org/10.1016/j.envint.2017.01.018>.
- 707 (34) Alam, F. C.; Sembiring, E.; Muntalif, B. S.; Suendo, V. Microplastic Distribution in
708 Surface Water and Sediment River around Slum and Industrial Area (Case Study:
709 Ciwalengke River, Majalaya District, Indonesia). *Chemosphere* **2019**, *224*, 637–645.
710 <https://doi.org/10.1016/j.chemosphere.2019.02.188>.
- 711 (35) Song, Y. K.; Hong, S. H.; Jang, M.; Kang, J.-H.; Kwon, O. Y.; Han, G. M.; Shim, W. J.
712 Large Accumulation of Micro-Sized Synthetic Polymer Particles in the Sea Surface
713 Microlayer. *Environ. Sci. Technol.* **2014**, *48* (16), 9014–9021.
714 <https://doi.org/10.1021/es501757s>.
- 715 (36) Vermaire, J. C.; Pomeroy, C.; Herczegh, S. M.; Haggart, O.; Murphy, M. Microplastic
716 Abundance and Distribution in the Open Water and Sediment of the Ottawa River,
717 Canada, and Its Tributaries. *FACETS* **2017**. <https://doi.org/10.1139/facets-2016-0070>.
- 718 (37) Yan, M.; Nie, H.; Xu, K.; He, Y.; Hu, Y.; Huang, Y.; Wang, J. Microplastic Abundance,
719 Distribution and Composition in the Pearl River along Guangzhou City and Pearl River
720 Estuary, China. *Chemosphere* **2019**, *217*, 879–886.
721 <https://doi.org/10.1016/j.chemosphere.2018.11.093>.
- 722 (38) Zhao, W.; Huang, W.; Yin, M.; Huang, P.; Ding, Y.; Ni, X.; Xia, H.; Liu, H.; Wang, G.;
723 Zheng, H.; Cai, M. Tributary Inflows Enhance the Microplastic Load in the Estuary: A
724 Case from the Qiantang River. *Mar. Pollut. Bull.* **2020**, *156*, 111152.
725 <https://doi.org/10.1016/j.marpolbul.2020.111152>.
- 726 (39) Lin, L.; Zuo, L.-Z.; Peng, J.-P.; Cai, L.-Q.; Fok, L.; Yan, Y.; Li, H.-X.; Xu, X.-R.
727 Occurrence and Distribution of Microplastics in an Urban River: A Case Study in the

- 728 Pearl River along Guangzhou City, China. *Sci. Total Environ.* **2018**, *644*, 375–381.
729 <https://doi.org/10.1016/j.scitotenv.2018.06.327>.
- 730 (40) Simmerman, C. B.; Wasik, J. K. C. The Effect of Urban Point Source Contamination on
731 Microplastic Levels in Water and Organisms in a Cold-Water Stream. *Limnol. Oceanogr.*
732 *Lett.* **2020**, *5* (1), 137–146. <https://doi.org/10.1002/lol2.10138>.
- 733 (41) Watkins, L.; McGrattan, S.; Sullivan, P. J.; Walter, M. T. The Effect of Dams on River
734 Transport of Microplastic Pollution. *Sci. Total Environ.* **2019**, *664*, 834–840.
735 <https://doi.org/10.1016/j.scitotenv.2019.02.028>.
- 736 (42) Tien, C.-J.; Wang, Z.-X.; Chen, C. S. Microplastics in Water, Sediment and Fish from the
737 Fengshan River System: Relationship to Aquatic Factors and Accumulation of Polycyclic
738 Aromatic Hydrocarbons by Fish. *Environ. Pollut.* **2020**, *265*, 114962.
739 <https://doi.org/10.1016/j.envpol.2020.114962>.
- 740 (43) Kabir, A. H. M. E.; Sekine, M.; Imai, T.; Yamamoto, K. Microplastics Pollution in the
741 Seto Inland Sea and Sea of Japan Surrounded Yamaguchi Prefecture Areas, Japan:
742 Abundance, Characterization and Distribution, and Potential Occurrences. *J. Water*
743 *Environ. Technol.* **2020**, *18* (3), 175–194. <https://doi.org/10.2965/jwet.19-127>.
- 744 (44) Liu, Y.; Zhang, J.; Cai, C.; He, Y.; Chen, L.; Xiong, X.; Huang, H.; Tao, S.; Liu, W.
745 Occurrence and Characteristics of Microplastics in the Haihe River: An Investigation of a
746 Seagoing River Flowing through a Megacity in Northern China - ScienceDirect. *Environ.*
747 *Pollut.* **2020**. <https://doi.org/10.1016/j.envpol.2020.114261>.
- 748 (45) Mao, R.; Hu, Y.; Zhang, S.; Wu, R.; Guo, X. Microplastics in the Surface Water of
749 Wuliangshai Lake, Northern China. *Sci. Total Environ.* **2020**, *723*, 137820.
750 <https://doi.org/10.1016/j.scitotenv.2020.137820>.
- 751 (46) Nan, B.; Su, L.; Kellar, C.; Craig, N. J.; Keough, M. J.; Pettigrove, V. Identification of
752 Microplastics in Surface Water and Australian Freshwater Shrimp *Paratya Australiensis* in
753 Victoria, Australia. *Environ. Pollut.* **2020**, *259*, 113865.
754 <https://doi.org/10.1016/j.envpol.2019.113865>.
- 755 (47) Wu, N.; Zhang, Y.; Zhang, X.; Zhao, Z.; He, J.; Li, W.; Ma, Y.; Niu, Z. Occurrence and
756 Distribution of Microplastics in the Surface Water and Sediment of Two Typical Estuaries
757 in Bohai Bay, China. *Environ. Sci. Process. Impacts* **2019**, *21* (7), 1143–1152.
758 <https://doi.org/10.1039/C9EM00148D>.
- 759 (48) Han, M.; Niu, X.; Tang, M.; Zhang, B.-T.; Wang, G.; Yue, W.; Kong, X.; Zhu, J.
760 Distribution of Microplastics in Surface Water of the Lower Yellow River near Estuary.
761 *Sci. Total Environ.* **2020**, *707*, 135601. <https://doi.org/10.1016/j.scitotenv.2019.135601>.
- 762 (49) Zhang, J.; Zhang, C.; Deng, Y.; Wang, R.; Ma, E.; Wang, J.; Bai, J.; Wu, J.; Zhou, Y.
763 Microplastics in the Surface Water of Small-Scale Estuaries in Shanghai. *Mar. Pollut.*
764 *Bull.* **2019**, *149*, 110569. <https://doi.org/10.1016/j.marpolbul.2019.110569>.
- 765 (50) Su, L.; Xue, Y.; Li, L.; Yang, D.; Kolandhasamy, P.; Li, D.; Shi, H. Microplastics in Taihu
766 Lake, China. *Environ. Pollut.* **2016**, *216*, 711–719.
767 <https://doi.org/10.1016/j.envpol.2016.06.036>.
- 768 (51) Zhu, L.; Bai, H.; Chen, B.; Sun, X.; Qu, K.; Xia, B. Microplastic Pollution in North
769 Yellow Sea, China: Observations on Occurrence, Distribution and Identification. *Sci.*
770 *Total Environ.* **2018**, *636*, 20–29. <https://doi.org/10.1016/j.scitotenv.2018.04.182>.
- 771 (52) Kosore, C.; Ojwang, L.; Maghanga, J.; Kamau, J.; Kimeli, A.; Omukoto, J.; Ngisiag'e, N.;
772 Mwaluma, J.; Ong'ada, H.; Magori, C.; Ndirui, E. Occurrence and Ingestion of
773 Microplastics by Zooplankton in Kenya's Marine Environment: First Documented

- 774 Evidence. *Afr. J. Mar. Sci.* **2018**, *40* (3), 225–234.
775 <https://doi.org/10.2989/1814232X.2018.1492969>.
- 776 (53) Huang, Y.; Yan, M.; Xu, K.; Nie, H.; Gong, H.; Wang, J. Distribution Characteristics of
777 Microplastics in Zhubi Reef from South China Sea. *Environ. Pollut.* **2019**, *255*, 113133.
778 <https://doi.org/10.1016/j.envpol.2019.113133>.
- 779 (54) Zhang, D.; Cui, Y.; Zhou, H.; Jin, C.; Yu, X.; Xu, Y.; Li, Y.; Zhang, C. Microplastic
780 Pollution in Water, Sediment, and Fish from Artificial Reefs around the Ma'an
781 Archipelago, Shengsi, China. *Sci. Total Environ.* **2020**, *703*, 134768.
782 <https://doi.org/10.1016/j.scitotenv.2019.134768>.
- 783 (55) Hu, L.; Chernick, M.; Hinton, D. E.; Shi, H. Microplastics in Small Waterbodies and
784 Tadpoles from Yangtze River Delta, China. *Environ. Sci. Technol.* **2018**, *52* (15), 8885–
785 8893. <https://doi.org/10.1021/acs.est.8b02279>.
- 786 (56) Colton, J. B.; Knapp, F. D.; Burns, B. R. Plastic Particles in Surface Waters of the
787 Northwestern Atlantic. *Science* **1974**, *185* (4150), 491–497.
- 788 (57) Ryan, P. G. The Characteristics and Distribution of Plastic Particles at the Sea-Surface off
789 the Southwestern Cape Province, South Africa. *Mar. Environ. Res.* **1988**, *25* (4), 249–273.
790 [https://doi.org/10.1016/0141-1136\(88\)90015-3](https://doi.org/10.1016/0141-1136(88)90015-3).
- 791 (58) Lattin, G. L.; Moore, C. J.; Zellers, A. F.; Moore, S. L.; Weisberg, S. B. A Comparison of
792 Neustonic Plastic and Zooplankton at Different Depths near the Southern California
793 Shore. *Mar. Pollut. Bull.* **2004**, *49* (4), 291–294.
794 <https://doi.org/10.1016/j.marpolbul.2004.01.020>.
- 795 (59) Ivar do Sul, J. A.; Costa, M. F.; Barletta, M.; Cysneiros, F. J. A. Pelagic Microplastics
796 around an Archipelago of the Equatorial Atlantic. *Mar. Pollut. Bull.* **2013**, *75* (1), 305–
797 309. <https://doi.org/10.1016/j.marpolbul.2013.07.040>.
- 798 (60) Doyle, M. J.; Watson, W.; Bowlin, N. M.; Sheavly, S. B. Plastic Particles in Coastal
799 Pelagic Ecosystems of the Northeast Pacific Ocean. *Mar. Environ. Res.* **2011**, *71* (1), 41–
800 52. <https://doi.org/10.1016/j.marenvres.2010.10.001>.
- 801 (61) Collignon, A.; Hecq, J.-H.; Glagani, F.; Voisin, P.; Collard, F.; Goffart, A. Neustonic
802 Microplastic and Zooplankton in the North Western Mediterranean Sea. *Mar. Pollut. Bull.*
803 **2012**, *64* (4), 861–864. <https://doi.org/10.1016/j.marpolbul.2012.01.011>.
- 804 (62) Schönlau, C.; Karlsson, T. M.; Rotander, A.; Nilsson, H.; Engwall, M.; van Bavel, B.;
805 Kärman, A. Microplastics in Sea-Surface Waters Surrounding Sweden Sampled by Manta
806 Trawl and in-Situ Pump. *Mar. Pollut. Bull.* **2020**, *153*, 111019.
807 <https://doi.org/10.1016/j.marpolbul.2020.111019>.
- 808 (63) Yonkos, L. T.; Friedel, E. A.; Perez-Reyes, A. C.; Ghosal, S.; Arthur, C. D. Microplastics
809 in Four Estuarine Rivers in the Chesapeake Bay, U.S.A. *Environ. Sci. Technol.* **2014**, *48*
810 (24), 14195–14202. <https://doi.org/10.1021/es5036317>.
- 811 (64) Eriksen, M.; Maximenko, N.; Thiel, M.; Cummins, A.; Lattin, G.; Wilson, S.; Hafner, J.;
812 Zellers, A.; Rifman, S. Plastic Pollution in the South Pacific Subtropical Gyre. *Mar.*
813 *Pollut. Bull.* **2013**, *68* (1), 71–76. <https://doi.org/10.1016/j.marpolbul.2012.12.021>.
- 814 (65) de Lucia, G. A.; Caliani, I.; Marra, S.; Camedda, A.; Coppa, S.; Alcaro, L.; Campani, T.;
815 Giannetti, M.; Coppola, D.; Cicero, A. M.; Panti, C.; Bainsi, M.; Guerranti, C.; Marsili, L.;
816 Massaro, G.; Fossi, M. C.; Matiddi, M. Amount and Distribution of Neustonic Micro-
817 Plastic off the Western Sardinian Coast (Central-Western Mediterranean Sea). *Mar.*
818 *Environ. Res.* **2014**, *100*, 10–16. <https://doi.org/10.1016/j.marenvres.2014.03.017>.

- 819 (66) Lima, A. R. A.; Costa, M. F.; Barletta, M. Distribution Patterns of Microplastics within
820 the Plankton of a Tropical Estuary. *Environ. Res.* **2014**, *132*, 146–155.
821 <https://doi.org/10.1016/j.envres.2014.03.031>.
- 822 (67) Panti, C.; Giannetti, M.; Bainsi, M.; Rubegni, F.; Minutoli, R.; Fossi, M. C. Occurrence,
823 Relative Abundance and Spatial Distribution of Microplastics and Zooplankton NW of
824 Sardinia in the Pelagos Sanctuary Protected Area, Mediterranean Sea. *Environ. Chem.*
825 **2015**, *12* (5), 618. <https://doi.org/10.1071/EN14234>.
- 826 (68) Setälä, O.; Magnusson, K.; Lehtiniemi, M.; Norén, F. Distribution and Abundance of
827 Surface Water Microlitter in the Baltic Sea: A Comparison of Two Sampling Methods.
828 *Mar. Pollut. Bull.* **2016**, *110* (1), 177–183.
829 <https://doi.org/10.1016/j.marpolbul.2016.06.065>.
- 830 (69) Zhao, S.; Zhu, L.; Wang, T.; Li, D. Suspended Microplastics in the Surface Water of the
831 Yangtze Estuary System, China: First Observations on Occurrence, Distribution. *Mar.*
832 *Pollut. Bull.* **2014**, *86* (1), 562–568. <https://doi.org/10.1016/j.marpolbul.2014.06.032>.
- 833 (70) Faure, F.; Demars, C.; Wieser, O.; Kunz, M.; Alencastro, L. F. de. Plastic Pollution in
834 Swiss Surface Waters: Nature and Concentrations, Interaction with Pollutants. *Environ.*
835 *Chem.* **2015**, *12* (5), 582–591. <https://doi.org/10.1071/EN14218>.
- 836 (71) Free, C. M.; Jensen, O. P.; Mason, S. A.; Eriksen, M.; Williamson, N. J.; Boldgiv, B.
837 High-Levels of Microplastic Pollution in a Large, Remote, Mountain Lake. *Mar. Pollut.*
838 *Bull.* **2014**, *85* (1), 156–163. <https://doi.org/10.1016/j.marpolbul.2014.06.001>.
- 839 (72) Cózar, A.; Sanz-Martín, M.; Martí, E.; González-Gordillo, J. I.; Ubeda, B.; Gálvez, J. Á.;
840 Irigoien, X.; Duarte, C. M. Plastic Accumulation in the Mediterranean Sea. *PLOS ONE*
841 **2015**, *10* (4), e0121762. <https://doi.org/10.1371/journal.pone.0121762>.
- 842 (73) Frère, L.; Paul-Pont, I.; Rinnert, E.; Petton, S.; Jaffré, J.; Bihannic, I.; Soudant, P.;
843 Lambert, C.; Huvet, A. Influence of Environmental and Anthropogenic Factors on the
844 Composition, Concentration and Spatial Distribution of Microplastics: A Case Study of
845 the Bay of Brest (Brittany, France). *Environ. Pollut.* **2017**, *225*, 211–222.
846 <https://doi.org/10.1016/j.envpol.2017.03.023>.
- 847 (74) Gajšt, T.; Bizjak, T.; Palatinus, A.; Liubartseva, S.; Kržan, A. Sea Surface Microplastics in
848 Slovenian Part of the Northern Adriatic. *Mar. Pollut. Bull.* **2016**, *113* (1), 392–399.
849 <https://doi.org/10.1016/j.marpolbul.2016.10.031>.
- 850 (75) Mani, T.; Hauk, A.; Walter, U.; Burkhardt-Holm, P. Microplastics Profile along the Rhine
851 River. *Sci. Rep.* **2015**, *5* (1), 1–7. <https://doi.org/10.1038/srep17988>.
- 852 (76) Zhang, K.; Gong, W.; Lv, J.; Xiong, X.; Wu, C. Accumulation of Floating Microplastics
853 behind the Three Gorges Dam. *Environ. Pollut.* **2015**, *204*, 117–123.
854 <https://doi.org/10.1016/j.envpol.2015.04.023>.
- 855 (77) Cable, R. N.; Beletsky, D.; Beletsky, R.; Wigginton, K.; Locke, B. W.; Duhaime, M. B.
856 Distribution and Modeled Transport of Plastic Pollution in the Great Lakes, the World's
857 Largest Freshwater Resource. *Front. Environ. Sci.* **2017**, *5*.
858 <https://doi.org/10.3389/fenvs.2017.00045>.
- 859 (78) Naidoo, T.; Glassom, D.; Smit, A. J. Plastic Pollution in Five Urban Estuaries of
860 KwaZulu-Natal, South Africa. *Mar. Pollut. Bull.* **2015**, *101* (1), 473–480.
861 <https://doi.org/10.1016/j.marpolbul.2015.09.044>.
- 862 (79) Lusher, A. L.; Tirelli, V.; O'Connor, I.; Officer, R. Microplastics in Arctic Polar Waters:
863 The First Reported Values of Particles in Surface and Sub-Surface Samples. *Sci. Rep.*
864 **2015**, *5* (1), 14947. <https://doi.org/10.1038/srep14947>.

- 865 (80) Aytan, U.; Valente, A.; Senturk, Y.; Usta, R.; Esensoy Sahin, F. B.; Mazlum, R. E.;
866 Agirbas, E. First Evaluation of Neustonic Microplastics in Black Sea Waters. *Mar.*
867 *Environ. Res.* **2016**, *119*, 22–30. <https://doi.org/10.1016/j.marenvres.2016.05.009>.
- 868 (81) Castro, R. O.; Silva, M. L.; Marques, M. R. C.; de Araújo, F. V. Evaluation of
869 Microplastics in Jurujuba Cove, Niterói, RJ, Brazil, an Area of Mussels Farming. *Mar.*
870 *Pollut. Bull.* **2016**, *110* (1), 555–558. <https://doi.org/10.1016/j.marpolbul.2016.05.037>.
- 871 (82) van der Hal, N.; Ariel, A.; Angel, D. L. Exceptionally High Abundances of Microplastics
872 in the Oligotrophic Israeli Mediterranean Coastal Waters. *Mar. Pollut. Bull.* **2017**, *116* (1),
873 151–155. <https://doi.org/10.1016/j.marpolbul.2016.12.052>.
- 874 (83) Sadri, S. S.; Thompson, R. C. On the Quantity and Composition of Floating Plastic Debris
875 Entering and Leaving the Tamar Estuary, Southwest England. *Mar. Pollut. Bull.* **2014**, *81*
876 (1), 55–60. <https://doi.org/10.1016/j.marpolbul.2014.02.020>.
- 877 (84) Baldwin, A. K.; Corsi, S. R.; Mason, S. A. Plastic Debris in 29 Great Lakes Tributaries:
878 Relations to Watershed Attributes and Hydrology. *Environ. Sci. Technol.* **2016**, *50* (19),
879 10377–10385. <https://doi.org/10.1021/acs.est.6b02917>.
- 880 (85) Estahbanati, S.; Fahrenfeld, N. L. Influence of Wastewater Treatment Plant Discharges on
881 Microplastic Concentrations in Surface Water. *Chemosphere* **2016**, *162*, 277–284.
882 <https://doi.org/10.1016/j.chemosphere.2016.07.083>.
- 883 (86) Isobe, A.; Uchiyama-Matsumoto, K.; Uchida, K.; Tokai, T. Microplastics in the Southern
884 Ocean. *Mar. Pollut. Bull.* **2017**, *114* (1), 623–626.
885 <https://doi.org/10.1016/j.marpolbul.2016.09.037>.
- 886 (87) Wang, T.; Zou, X.; Li, B.; Yao, Y.; Li, J.; Hui, H.; Yu, W.; Wang, C. Microplastics in a
887 Wind Farm Area: A Case Study at the Rudong Offshore Wind Farm, Yellow Sea, China.
888 *Mar. Pollut. Bull.* **2018**, *128*, 466–474. <https://doi.org/10.1016/j.marpolbul.2018.01.050>.
- 889 (88) Zhang, W.; Zhang, S.; Wang, J.; Wang, Y.; Mu, J.; Wang, P.; Lin, X.; Ma, D. Microplastic
890 Pollution in the Surface Waters of the Bohai Sea, China. *Environ. Pollut.* **2017**, *231*, 541–
891 548. <https://doi.org/10.1016/j.envpol.2017.08.058>.
- 892 (89) Anderson, P. J.; Warrack, S.; Langen, V.; Challis, J. K.; Hanson, M. L.; Rennie, M. D.
893 Microplastic Contamination in Lake Winnipeg, Canada. *Environ. Pollut.* **2017**, *225*, 223–
894 231. <https://doi.org/10.1016/j.envpol.2017.02.072>.
- 895 (90) Lenaker, P. L.; Baldwin, A. K.; Corsi, S. R.; Mason, S. A.; Reneau, P. C.; Scott, J. W.
896 Vertical Distribution of Microplastics in the Water Column and Surficial Sediment from
897 the Milwaukee River Basin to Lake Michigan. *Environ. Sci. Technol.* **2019**, *53* (21),
898 12227–12237. <https://doi.org/10.1021/acs.est.9b03850>.
- 899 (91) Uurasjärvi, E.; Hartikainen, S.; Setälä, O.; Lehtiniemi, M.; Koistinen, A. Microplastic
900 Concentrations, Size Distribution, and Polymer Types in the Surface Waters of a Northern
901 European Lake. *Water Environ. Res.* **2020**, *92* (1), 149–156.
902 <https://doi.org/10.1002/wer.1229>.
- 903 (92) Naidoo, T.; Glassom, D. Sea-Surface Microplastic Concentrations along the Coastal Shelf
904 of KwaZulu–Natal, South Africa. *Mar. Pollut. Bull.* **2019**, *149*, 110514.
905 <https://doi.org/10.1016/j.marpolbul.2019.110514>.
- 906 (93) Tan, X.; Yu, X.; Cai, L.; Wang, J.; Peng, J. Microplastics and Associated PAHs in Surface
907 Water from the Feilaixia Reservoir in the Beiji River, China. *Chemosphere* **2019**, *221*,
908 834–840. <https://doi.org/10.1016/j.chemosphere.2019.01.022>.
- 909 (94) Tang, G.; Liu, M.; Zhou, Q.; He, H.; Chen, K.; Zhang, H.; Hu, J.; Huang, Q.; Luo, Y.; Ke,
910 H.; Chen, B.; Xu, X.; Cai, M. Microplastics and Polycyclic Aromatic Hydrocarbons

- 911 (PAHs) in Xiamen Coastal Areas: Implications for Anthropogenic Impacts. *Sci. Total*
912 *Environ.* **2018**, *634*, 811–820. <https://doi.org/10.1016/j.scitotenv.2018.03.336>.
- 913 (95) Wagner, S.; Klöckner, P.; Stier, B.; Römer, M.; Seiwert, B.; Reemtsma, T.; Schmidt, C.
914 Relationship between Discharge and River Plastic Concentrations in a Rural and an Urban
915 Catchment. *Environ. Sci. Technol.* **2019**, *53* (17), 10082–10091.
916 <https://doi.org/10.1021/acs.est.9b03048>.
- 917 (96) Mataji, A.; Taleshi, M. S.; Balimoghaddas, E. Distribution and Characterization of
918 Microplastics in Surface Waters and the Southern Caspian Sea Coasts Sediments. *Arch.*
919 *Environ. Contam. Toxicol.* **2020**, *78* (1), 86–93. [https://doi.org/10.1007/s00244-019-](https://doi.org/10.1007/s00244-019-00700-2)
920 [00700-2](https://doi.org/10.1007/s00244-019-00700-2).
- 921 (97) Rose, D.; Webber, M. Characterization of Microplastics in the Surface Waters of Kingston
922 Harbour. *Sci. Total Environ.* **2019**, *664*, 753–760.
923 <https://doi.org/10.1016/j.scitotenv.2019.01.319>.
- 924 (98) Watkins, L.; Sullivan, P. J.; Walter, M. T. A Case Study Investigating Temporal Factors
925 That Influence Microplastic Concentration in Streams under Different Treatment Regimes.
926 *Environ. Sci. Pollut. Res.* **2019**, *26* (21), 21797–21807. [https://doi.org/10.1007/s11356-](https://doi.org/10.1007/s11356-019-04663-8)
927 [019-04663-8](https://doi.org/10.1007/s11356-019-04663-8).
- 928 (99) Hendrickson, E.; Minor, E. C.; Schreiner, K. Microplastic Abundance and Composition in
929 Western Lake Superior As Determined via Microscopy, Pyr-GC/MS, and FTIR. *Environ.*
930 *Sci. Technol.* **2018**, *52* (4), 1787–1796. <https://doi.org/10.1021/acs.est.7b05829>.
- 931 (100) Mu, J.; Zhang, S.; Qu, L.; Jin, F.; Fang, C.; Ma, X.; Zhang, W.; Wang, J. Microplastics
932 Abundance and Characteristics in Surface Waters from the Northwest Pacific, the Bering
933 Sea, and the Chukchi Sea. *Mar. Pollut. Bull.* **2019**, *143*, 58–65.
934 <https://doi.org/10.1016/j.marpolbul.2019.04.023>.
- 935 (101) Tunçer, S.; Artüz, O. B.; Demirkol, M.; Artüz, M. L. First Report of Occurrence,
936 Distribution, and Composition of Microplastics in Surface Waters of the Sea of Marmara,
937 Turkey. *Mar. Pollut. Bull.* **2018**, *135*, 283–289.
938 <https://doi.org/10.1016/j.marpolbul.2018.06.054>.
- 939 (102) Mai, L.; Bao, L.-J.; Shi, L.; Liu, L.-Y.; Zeng, E. Y. Polycyclic Aromatic Hydrocarbons
940 Affiliated with Microplastics in Surface Waters of Bohai and Huanghai Seas, China.
941 *Environ. Pollut.* **2018**, *241*, 834–840. <https://doi.org/10.1016/j.envpol.2018.06.012>.
- 942 (103) Zhang, L.; Liu, J.; Xie, Y.; Zhong, S.; Yang, B.; Lu, D.; Zhong, Q. Distribution of
943 Microplastics in Surface Water and Sediments of Qin River in Beibu Gulf, China. *Sci.*
944 *Total Environ.* **2020**, *708*, 135176. <https://doi.org/10.1016/j.scitotenv.2019.135176>.
- 945 (104) Zhang, X.; Leng, Y.; Liu, X.; Huang, K.; Wang, J. Microplastics' Pollution and Risk
946 Assessment in an Urban River: A Case Study in the Yongjiang River, Nanning City, South
947 China. *Expo. Health* **2020**, *12* (2), 141–151. <https://doi.org/10.1007/s12403-018-00296-3>.
- 948 (105) Campanale, C.; Stock, F.; Massarelli, C.; Kochleus, C.; Bagnuolo, G.; Reifferscheid, G.;
949 Uricchio, V. F. Microplastics and Their Possible Sources: The Example of Ofanto River in
950 Southeast Italy. *Environ. Pollut.* **2020**, *258*, 113284.
951 <https://doi.org/10.1016/j.envpol.2019.113284>.
- 952 (106) Wong, G.; Löwemark, L.; Kunz, A. Microplastic Pollution of the Tamsui River and Its
953 Tributaries in Northern Taiwan: Spatial Heterogeneity and Correlation with Precipitation.
954 *Environ. Pollut.* **2020**, *260*, 113935. <https://doi.org/10.1016/j.envpol.2020.113935>.
- 955 (107) Herrera, A.; Raymond, E.; Martínez, I.; Álvarez, S.; Canning-Clode, J.; Gestoso, I.; Pham,
956 C. K.; Ríos, N.; Rodríguez, Y.; Gómez, M. First Evaluation of Neustonic Microplastics in

- 957 the Macaronesian Region, NE Atlantic. *Mar. Pollut. Bull.* **2020**, *153*, 110999.
 958 <https://doi.org/10.1016/j.marpolbul.2020.110999>.
- 959 (108) Baldwin, A. K.; Spanjer, A. R.; Rosen, M. R.; Thom, T. Microplastics in Lake Mead
 960 National Recreation Area, USA: Occurrence and Biological Uptake. *PLOS ONE* **2020**, *15*
 961 (5), e0228896. <https://doi.org/10.1371/journal.pone.0228896>.
- 962 (109) Syakti, A. D.; Hidayati, N. V.; Jaya, Y. V.; Siregar, S. H.; Yude, R.; Suhendy; Asia, L.;
 963 Wong-Wah-Chung, P.; Doumenq, P. Simultaneous Grading of Microplastic Size
 964 Sampling in the Small Islands of Bintan Water, Indonesia. *Mar. Pollut. Bull.* **2018**, *137*,
 965 593–600. <https://doi.org/10.1016/j.marpolbul.2018.11.005>.
- 966 (110) Bakir, A.; Desender, M.; Wilkinson, T.; Van Hoytema, N.; Amos, R.; Airahui, S.;
 967 Graham, J.; Maes, T. Occurrence and Abundance of Meso and Microplastics in Sediment,
 968 Surface Waters, and Marine Biota from the South Pacific Region. *Mar. Pollut. Bull.* **2020**,
 969 *160*, 111572. <https://doi.org/10.1016/j.marpolbul.2020.111572>.
- 970 (111) Kazour, M.; Jemaa, S.; Issa, C.; Khalaf, G.; Amara, R. Microplastics Pollution along the
 971 Lebanese Coast (Eastern Mediterranean Basin): Occurrence in Surface Water, Sediments
 972 and Biota Samples. *Sci. Total Environ.* **2019**, *696*, 133933.
 973 <https://doi.org/10.1016/j.scitotenv.2019.133933>.
- 974 (112) Mai, L.; You, S.-N.; He, H.; Bao, L.-J.; Liu, L.-Y.; Zeng, E. Y. Riverine Microplastic
 975 Pollution in the Pearl River Delta, China: Are Modeled Estimates Accurate? *Environ. Sci.*
 976 *Technol.* **2019**, *53* (20), 11810–11817. <https://doi.org/10.1021/acs.est.9b04838>.
- 977 (113) Park, T.-J.; Lee, S.-H.; Lee, M.-S.; Lee, J.-K.; Lee, S.-H.; Zoh, K.-D. Occurrence of
 978 Microplastics in the Han River and Riverine Fish in South Korea. *Sci. Total Environ.*
 979 **2020**, *708*, 134535. <https://doi.org/10.1016/j.scitotenv.2019.134535>.
- 980 (114) Rasta, M.; Sattari, M.; Taleshi, M. S.; Namin, J. I. Identification and Distribution of
 981 Microplastics in the Sediments and Surface Waters of Anzali Wetland in the Southwest
 982 Caspian Sea, Northern Iran. *Mar. Pollut. Bull.* **2020**, *160*, 111541.
 983 <https://doi.org/10.1016/j.marpolbul.2020.111541>.
- 984 (115) Egessa, R.; Nankabirwa, A.; Ocaya, H.; Pabire, W. G. Microplastic Pollution in Surface
 985 Water of Lake Victoria. *Sci. Total Environ.* **2020**, *741*, 140201.
 986 <https://doi.org/10.1016/j.scitotenv.2020.140201>.
- 987 (116) Scherer, C.; Weber, A.; Stock, F.; Vurusic, S.; Egerci, H.; Kochleus, C.; Arendt, N.;
 988 Foeldi, C.; Dierkes, G.; Wagner, M.; Brennholt, N.; Reifferscheid, G. Comparative
 989 Assessment of Microplastics in Water and Sediment of a Large European River. *Sci. Total*
 990 *Environ.* **2020**, *738*, 139866. <https://doi.org/10.1016/j.scitotenv.2020.139866>.
- 991 (117) Di Mauro, R.; Kupchik, M. J.; Benfield, M. C. Abundant Plankton-Sized Microplastic
 992 Particles in Shelf Waters of the Northern Gulf of Mexico. *Environ. Pollut.* **2017**, *230*,
 993 798–809. <https://doi.org/10.1016/j.envpol.2017.07.030>.
- 994 (118) Jiang, C.; Yin, L.; Li, Z.; Wen, X.; Luo, X.; Hu, S.; Yang, H.; Long, Y.; Deng, B.; Huang,
 995 L.; Liu, Y. Microplastic Pollution in the Rivers of the Tibet Plateau. *Environ. Pollut.* **2019**,
 996 *249*, 91–98. <https://doi.org/10.1016/j.envpol.2019.03.022>.
- 997 (119) Wang, W.; Ndungu, A. W.; Li, Z.; Wang, J. Microplastics Pollution in Inland Freshwaters
 998 of China: A Case Study in Urban Surface Waters of Wuhan, China. *Sci. Total Environ.*
 999 **2017**, *575*, 1369–1374. <https://doi.org/10.1016/j.scitotenv.2016.09.213>.
- 1000 (120) Bordós, G.; Urbányi, B.; Micsinai, A.; Kriszt, B.; Palotai, Z.; Szabó, I.; Hantosi, Z.;
 1001 Szoboszlaj, S. Identification of Microplastics in Fish Ponds and Natural Freshwater

- 1002 Environments of the Carpathian Basin, Europe. *Chemosphere* **2019**, *216*, 110–116.
1003 <https://doi.org/10.1016/j.chemosphere.2018.10.110>.
- 1004 (121) Ding, L.; Mao, R. fan; Guo, X.; Yang, X.; Zhang, Q.; Yang, C. Microplastics in Surface
1005 Waters and Sediments of the Wei River, in the Northwest of China. *Sci. Total Environ.*
1006 **2019**, *667*, 427–434. <https://doi.org/10.1016/j.scitotenv.2019.02.332>.
- 1007 (122) Karlsson, T. M.; Kärrman, A.; Rotander, A.; Hassellöv, M. Comparison between Manta
1008 Trawl and in Situ Pump Filtration Methods, and Guidance for Visual Identification of
1009 Microplastics in Surface Waters. *Environ. Sci. Pollut. Res.* **2020**, *27* (5), 5559–5571.
1010 <https://doi.org/10.1007/s11356-019-07274-5>.
- 1011 (123) Jiang, Y.; Yang, F.; Zhao, Y.; Wang, J. Greenland Sea Gyre Increases Microplastic
1012 Pollution in the Surface Waters of the Nordic Seas. *Sci. Total Environ.* **2020**, *712*, 136484.
1013 <https://doi.org/10.1016/j.scitotenv.2019.136484>.
- 1014 (124) Jiang, C.; Yin, L.; Wen, X.; Du, C.; Wu, L.; Long, Y.; Liu, Y.; Ma, Y.; Yin, Q.; Zhou, Z.;
1015 Pan, H. Microplastics in Sediment and Surface Water of West Dongting Lake and South
1016 Dongting Lake: Abundance, Source and Composition. *Int. J. Environ. Res. Public Health*
1017 **2018**, *15* (10), 2164. <https://doi.org/10.3390/ijerph15102164>.
- 1018 (125) Xu, P.; Peng, G.; Su, L.; Gao, Y.; Gao, L.; Li, D. Microplastic Risk Assessment in Surface
1019 Waters: A Case Study in the Changjiang Estuary, China. *Mar. Pollut. Bull.* **2018**, *133*,
1020 647–654. <https://doi.org/10.1016/j.marpolbul.2018.06.020>.
- 1021 (126) Li, Y.; Lu, Z.; Zheng, H.; Wang, J.; Chen, C. Microplastics in Surface Water and
1022 Sediments of Chongming Island in the Yangtze Estuary, China. *Environ. Sci. Eur.* **2020**,
1023 *32* (1), 15. <https://doi.org/10.1186/s12302-020-0297-7>.
- 1024 (127) Cai, M.; He, H.; Liu, M.; Li, S.; Tang, G.; Wang, W.; Huang, P.; Wei, G.; Lin, Y.; Chen,
1025 B.; Hu, J.; Cen, Z. Lost but Can't Be Neglected: Huge Quantities of Small Microplastics
1026 Hide in the South China Sea. *Sci. Total Environ.* **2018**, *633*, 1206–1216.
1027 <https://doi.org/10.1016/j.scitotenv.2018.03.197>.
- 1028 (128) Wakkaf, T.; El Zrelli, R.; Kedzierski, M.; Balti, R.; Shaiek, M.; Mansour, L.; Tlig-Zouari,
1029 S.; Bruzard, S.; Rabaoui, L. Characterization of Microplastics in the Surface Waters of an
1030 Urban Lagoon (Bizerte Lagoon, Southern Mediterranean Sea): Composition, Density,
1031 Distribution, and Influence of Environmental Factors. *Mar. Pollut. Bull.* **2020**, *160*,
1032 111625. <https://doi.org/10.1016/j.marpolbul.2020.111625>.
- 1033 (129) Chen, B.; Fan, Y.; Huang, W.; Rayhan, A. B. M. S.; Chen, K.; Cai, M. Observation of
1034 Microplastics in Mariculture Water of Longjiao Bay, Southeast China: Influence by
1035 Human Activities. *Mar. Pollut. Bull.* **2020**, *160*, 111655.
1036 <https://doi.org/10.1016/j.marpolbul.2020.111655>.
- 1037 (130) Wiggin, K. J.; Holland, E. B. Validation and Application of Cost and Time Effective
1038 Methods for the Detection of 3–500 µm Sized Microplastics in the Urban Marine and
1039 Estuarine Environments Surrounding Long Beach, California. *Mar. Pollut. Bull.* **2019**,
1040 *143*, 152–162. <https://doi.org/10.1016/j.marpolbul.2019.03.060>.
- 1041 (131) South, A. *Rnaturalearth: World Map Data from Natural Earth*; R package.
- 1042 (132) Koelmans, A. A.; Mohamed Nor, N. H.; Hermesen, E.; Kooi, M.; Mintenig, S. M.; De
1043 France, J. Microplastics in Freshwaters and Drinking Water: Critical Review and
1044 Assessment of Data Quality. *Water Res.* **2019**, *155*, 410–422.
1045 <https://doi.org/10.1016/j.watres.2019.02.054>.
- 1046 (133) Prata, J. C.; Manana, M. J.; da Costa, J. P.; Duarte, A. C.; Rocha-Santos, T. What Is the
1047 Minimum Volume of Sample to Find Small Microplastics: Laboratory Experiments and

- 1048 Sampling of Aveiro Lagoon and Vouga River, Portugal. *Water* **2020**, *12* (4), 1219.
1049 <https://doi.org/10.3390/w12041219>.
- 1050 (134) Lindeque, P. K.; Cole, M.; Coppock, R. L.; Lewis, C. N.; Miller, R. Z.; Watts, A. J. R.;
1051 Wilson-McNeal, A.; Wright, S. L.; Galloway, T. S. Are We Underestimating Microplastic
1052 Abundance in the Marine Environment? A Comparison of Microplastic Capture with Nets
1053 of Different Mesh-Size. *Environ. Pollut.* **2020**, 114721.
1054 <https://doi.org/10.1016/j.envpol.2020.114721>.
- 1055 (135) Garcia, T. M.; Campos, C. C.; Mota, E. M. T.; Santos, N. M. O.; Campelo, R. P. de S.;
1056 Prado, L. C. G.; Melo Junior, M.; Soares, M. de O. Microplastics in Subsurface Waters of
1057 the Western Equatorial Atlantic (Brazil). *Mar. Pollut. Bull.* **2020**, *150*, 110705.
1058 <https://doi.org/10.1016/j.marpolbul.2019.110705>.
- 1059 (136) Koelmans, A. A.; Redondo-Hasselerharm, P. E.; Mohamed Nor, N. H.; Kooi, M. Solving
1060 the Nonalignment of Methods and Approaches Used in Microplastic Research to
1061 Consistently Characterize Risk. *Environ. Sci. Technol.* **2020**, *54* (19), 12307–12315.
1062 <https://doi.org/10.1021/acs.est.0c02982>.
- 1063 (137) Lusher, A. L.; Burke, A.; O'Connor, I.; Officer, R. Microplastic Pollution in the Northeast
1064 Atlantic Ocean: Validated and Opportunistic Sampling. *Mar. Pollut. Bull.* **2014**, *88* (1),
1065 325–333. <https://doi.org/10.1016/j.marpolbul.2014.08.023>.
- 1066 (138) Hanvey, J. S.; Lewis, P. J.; Lavers, J. L.; Crosbie, N. D.; Pozo, K.; Clarke, B. O. A
1067 Review of Analytical Techniques for Quantifying Microplastics in Sediments. *Anal.*
1068 *Methods* **2017**, *9* (9), 1369–1383. <https://doi.org/10.1039/C6AY02707E>.
- 1069 (139) Ryan, P. G.; Suaria, G.; Perold, V.; Pierucci, A.; Bornman, T. G.; Aliani, S. Sampling
1070 Microfibres at the Sea Surface: The Effects of Mesh Size, Sample Volume and Water
1071 Depth. *Environ. Pollut.* **2020**, 258, 113413. <https://doi.org/10.1016/j.envpol.2019.113413>.
- 1072 (140) Witzig, C. S.; Földi, C.; Wörle, K.; Habermehl, P.; Pittroff, M.; Müller, Y. K.; Lauschke,
1073 T.; Fiener, P.; Dierkes, G.; Freier, K. P.; Zumbülte, N. When Good Intentions Go Bad—
1074 False Positive Microplastic Detection Caused by Disposable Gloves. *Environ. Sci.*
1075 *Technol.* **2020**, *54* (19), 12164–12172. <https://doi.org/10.1021/acs.est.0c03742>.
- 1076 (141) Lenz, R.; Enders, K.; Stedmon, C. A.; Mackenzie, D. M. A.; Nielsen, T. G. A Critical
1077 Assessment of Visual Identification of Marine Microplastic Using Raman Spectroscopy
1078 for Analysis Improvement. *Mar. Pollut. Bull.* **2015**, *100* (1), 82–91.
1079 <https://doi.org/10.1016/j.marpolbul.2015.09.026>.
- 1080 (142) Cada, G. F.; Loar, J. M. Relative Effectiveness of Two Ichthyoplankton Sampling
1081 Techniques. *Can. J. Fish. Aquat. Sci.* **1982**, *39* (6), 811–814. <https://doi.org/10.1139/f82-110>.
- 1082
- 1083 (143) Masson, S.; Pinel-Alloul, B.; Méthot, G.; Richard, N. Comparison of Nets and Pump
1084 Sampling Gears to Assess Zooplankton Vertical Distribution in Stratified Lakes. *J.*
1085 *Plankton Res.* **2004**, *26* (10), 1199–1206. <https://doi.org/10.1093/plankt/fbh109>.
- 1086 (144) Appel, D. S. Zooplankton Sampling in Riverine Systems: A Gear Comparison in the
1087 Upper Mississippi River. Masters Thesis, University of Wisconsin - La Crosse, 2019.
1088