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13 Abstract

Many studies have been conducted to quantify microplastic contamination, but only a few of 14 them have actually the sampling methodology and associated uncertainties. This study seeks 15 to examine the influence of sampling strategy on the confidence interval of river microplastic 16 estimates. 16 samples are collected in the Gave de Pau River (southwestern France) during a 17 three-hour window with a 330- μ m mesh size net. Three different exposure times (3, 5 and 7 18 minutes) allow for a respective filtration rate by the net of 35.6 m³ (3 samples), 59.4 m³ (10 19 samples), and 83.2 m³ (3 samples) of water. Organic matter contained in samples is removed 20 by hydrogen peroxide oxidation. The plastic particles are then counted and classified under a 21 binocular microscope. The microplastic concentrations vary between 2.64 and 4.24 22 microplastics/m³, with a median value of 3.26 microplastics/m³. Statistical analysis does not 23 show differences in microplastic concentrations for the three exposure times. This result 24

seems to demonstrate that a filtration of approx. 35 m³ of water is sufficient under similar 25 26 conditions (similar flow condition and degree of microplastic contamination) and can help reduce sampling and sample processing time. Other analyses, based on 10 filtrations of 59.4 27 m^3 , show that the higher the number of samples, the lower the confidence interval. For 28 triplicates, the mean confidence interval reaches 15% of the median value. Thus, collecting 29 triplicates would seem to offer a reasonable optimum, in combining an acceptable error 30 percentage and time efficiency. These results might depend on the microplastic load of the 31 river, therefore making it necessary to conduct similar analyses on other rivers. This study 32 reports for the first time uncertainties related to microplastic sampling in rivers. Such findings 33 34 will serve to set up long term monitoring, highlight spatial differences between sites and 35 improve the accuracy of annual microplastic fluxes in rivers.

36

37 Keywords: microplastics, uncertainties, rivers, pollution, sampling methodology

38

39 Introduction

40 Over the last decade, considerable attention has been paid to plastic pollution in the marine environment (Li et al., 2018; Li et al., 2016). Yet rivers, which are one of the major pathways 41 for plastics entering the ocean, have not been investigated to the same extent despite the fact 42 43 that in the past five years, an increasing number of studies have focused on river plastic contamination. Data are still scarce and sampling methods need to be improved and 44 harmonised (Dris et al., 2018; Eerkes-Medrano et al., 2015; Wagner et al., 2014). Moreover, 45 there is a major need to set up networks for measuring microplastics in inland waters to 46 monitor the evolution of their contamination over time and to set up databases at regional 47 scales, in line with regional regulations, like the Water Framework Directive in Europe 48

49 (although microplastics are not included as an indicator of good environmental status in its50 current version).

Only easy to implement methodologies can be considered. However, they have to be 51 52 representative of the water bodies. The 200-year long experience in river monitoring suggests that fluctuations in usual water quality parameters occur at much larger time scales than those 53 of sampling, and as a consequence, for instance, sampling 1 L of water is often enough to 54 measure both dissolved and particulate water quality parameters. Such an experience is 55 missing for microplastic contamination, in particular, for low concentration contamination, 56 taking into account that the dynamics of such particles in the water column is up to now badly 57 known. To the best of our knowledge, only two articles address small-scale temporal and 58 spatial variability of microplastics found in rivers (Dris et al., 2018; Liedermann et al., 2018). 59 Dris et al. (2018) analysed the temporal and spatial variability of fibre concentrations in the 60 Seine and Marne Rivers (France), using a 80 µm mesh net. They showed that the longer the 61 net deployment time, the lower the variability between consecutive samples. They also 62 assessed fibre distribution variability throughout the river cross-section and observed that 63 concentrations are similar across the water column and tend to increase near the banks. 64 Liedermann et al. (2018) studied microplastics distribution, using a 500 µm mesh net, within 65 the Danube River and also detected a slight tendency towards higher concentrations nearer the 66 banks. 67

On the basis of the sampling of 16 successive replicates and 3 sampling durations, in a river in Southwestern France, we investigate the microplastic concentration fluctuations and assess the corresponding uncertainties, and their variation with both the number of replicates and the time exposure.

72 Materials and Methods

73 Study site and sampling methodology

This study was conducted on April 6th, 2018 between 11 am and 2 pm in the Gave de Pau River, downstream of the Pau city centre yet still within the conurbation (Southwestern France, lat.: 43.304828°, long.: -0.436492°). At this location, the river is 87 m wide with a maximum depth of about 2 m. The flow is torrential with an annual mean discharge of 69.1 m³/s (average over the period February 2000 - July 2018 at station Q5231010 - Gave de Pau at Artiguelouve - Pont de Lescar; *Banque hydro* database, 2018).

The river flow surface velocity was measured three times in a row before sampling using a flowmeter (Flow Probe FR211) and ranged from 1.0 to 1.2 m/s. The average value (1.1 m/s) was used to estimate the volume of water being filtered by the net. The mean river flow at station Q5231010 was 75.8 m³/s (*Banque hydro* database, 2018).

Microplastic particles were sampled using a 330-µm mesh size net with a rectangular opening of 30 cm by 60 cm. This net was attached to a bridge roughly 6 meters from the river's left bank. A second rope, fastened to the frame, was used to pull the net out of the water from the bank. Two buoys were assembled on top of the frame to hold it just over the water surface, and weights were used to keep it straight in the water column (Picture S1, Video S1).

A total of 16 samples were collected within a three-hour window. The maximum number of samples that can be collected in this time frame ranges from 15 to 20 depending on the net exposition time chosen. Beyond this time window, the hypothesis of steady state of the river flow and pollution could be wrong. It was decided to use low net exposition times to avoid clogging and reduce sampling and sample processing time. This is of high importance considering that the method should be operational for large scale monitoring. The various immersion times of the net chosen were 3 min (3 replicates), 5 min (10 replicates) and 7 min 96 (3 replicates), corresponding to 35.6, 59.4 and 83.2 m³ of filtered water, respectively. The 597 minute net exposition time (59.4 m³) was thought to be the best option. This time was
98 therefore investigated further. Other values (i.e. 3 and 7 minutes, 35.6 and 83.2 m³
99 respectively) were also tested, although to a lower extent. Samples were stored in glass
100 containers with metal lids away from sunlight and at room temperature.

101

102 Microplastic extraction and identification

103 The samples were run through sieves with mesh sizes of 5 mm and 0.3 mm. The macroplastic fraction (> 5 mm) was observed with the naked eye. The microplastic fraction, with a particle 104 size lying between 0.3 mm and 5 mm, was treated in order to remove organic matter (Masura 105 et al., 2015; Hurley et al., 2018). 20 mL of aqueous 0.05 M Fe(II) solution and 20 mL of 30% 106 107 hydrogen peroxide were added to a glass beaker containing the 0.3 to 5 mm fraction of the 108 sample. The resulting mixture was then placed on a lab bench at room temperature for 5 109 minutes before being heated to 75°C and held at that temperature until gas bubbles could be observed. As the first bubbles cracked the surface, the beaker was removed from the hot plate 110 to avoid a violent reaction. When the solution had cooled slightly, the beaker was returned to 111 the hot plate and heated to 75°C for an additional 30 minutes. This operation was repeated 112 four times per sample due to the high quantity of organic matter. 113

The plastic particles were then counted and classified under a binocular microscope (Leica EZ4) by colour (blue, red, transparent/white, black, green and other) and type (round, fragments, angular and other shapes) (MERI, 2015). The fibres were not considered in this study. The same operator handled all the samples. Tweezers were used to poke at individual items whenever doubts arose.

Given the size range target (> $330 \ \mu m$), and as only fragments were considered, the risk of under or over estimation remains very low. In order to keep in mind the objective of a simple,

rapid and efficient method, it was decided not to proceed to a chemical characterisation, as
global methods based on Pyr-GC-MS are up to now high skilled and not quantitative, and
spectroscopic techniques (µFTIR or µRaman) are also high skilled and time consuming.

124

125 Statistical analyses

The number of plastic particles per sample was presented in terms of number of microplastics per cubic meter (MPs/m³). The median microplastic concentrations for the three volumes of filtered water, corresponding to the three exposure times, were compared using the nonparametrical Kruskal-Wallis test and the R software (R Core Team, 2018). Since the number of samples is small, non-parametric statistics, including medians and quantiles, are used.

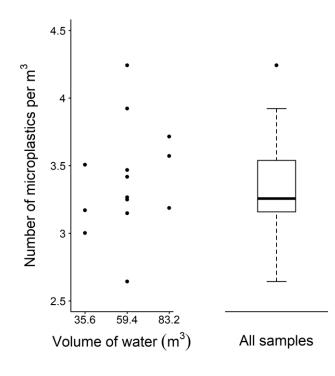
Sampling uncertainties were assessed based on the 10 filtrations of 59.4 m³ (5 minutes). A 131 resampling technique was applied to a number of samples ranging from 3 to 10. The number 132 of existing combinations varied from 560 (for 3 samples) to 8,008 (for 10 samples), with a 133 134 maximum of 12,870 (for 8 samples). For each combination of samples, the mean medians and mean standard deviations of microplastic densities per cubic meter were computed. As the 135 total number of combinations remains reasonable, all the combinations were tested. The 136 standard errors of the means and confidence intervals were then computed using a 95% 137 confidence level. 138

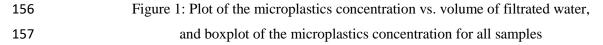
139 The delta of the standard errors of the means with respect to the median mean values has been 140 plotted vs. the number of samples in order to determine how the number of samples 141 influences the related uncertainties.

143 Results

In total, 3,191 microplastics were found within the 950 m³ of filtered water. 11.88 m³ of water were filtered per minute of exposure. Exposure times of 3, 5 and 7 minutes allowed filtering 35.6, 59.4 and 83.2 m³ of water, respectively. No macroplastic was caught in the net.

- The microplastic concentrations varied between 2.64 and 4.24 MPs/m³, with a median value
 of 3.26 MPs/m³.
- The shapes and colours of microplastic particles are presented in Table S1. The microplastic
 shape and colour variability between samples is low (the standard error ranges from 3% to 8%
 for shape variability and from 0.3% to 9% for colour variability).
- 152
- 153 Influence of the net exposure time
- 154 The distribution of values recorded is shown in Figure 1.





158

The microplastic concentration varies between 3.00 and 3.51 MPs/m³, 2.64 and 4.24 MPs/m³, 159 and 3.19 and 3.72 MPs/m³ for the three exposure times (3, 5 and 7 minutes corresponding to 160 35.6, 59.4 and 83.2 m³), respectively. Despite the exposure time, the concentrations remain 161 close although the 5-minute exposure covers a wider range of values. 162

A Kruskal-Wallis test does not highlight any statistical differences between microplastics 163 concentration for each exposure time ($\alpha = 0.05$, p = 0.5). The median value of microplastic 164 density per cubic meter for the 16 samples equals: 3.26 ± 0.21 MPs/m³. 165

166

Uncertainties 167

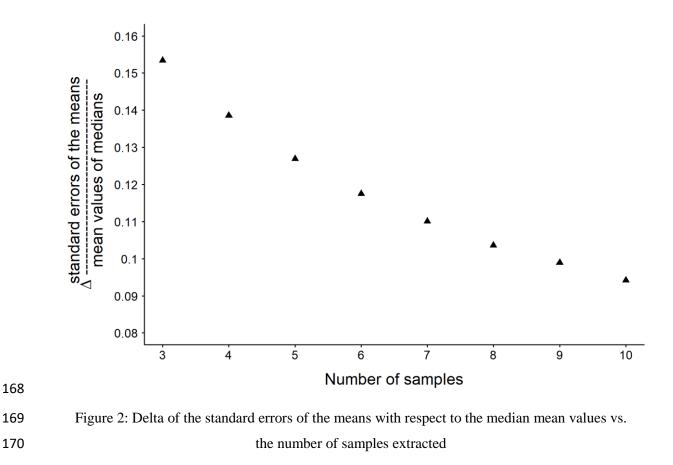


Figure 2 presents the delta of the standard errors of the means with respect to the median 172 mean values according to the number of samples extracted. This figure demonstrates that the 173

higher the number of samples are, the lower the confidence interval. All values are given inTable S2.

Analyses conducted on the 10 filtrations of 59.4 m³ (5 minutes) have shown that for triplicates, the mean confidence interval for the 560 combinations reaches 15% of the median value, while the maximum confidence interval for the 560 combinations reaches 40% of the median value. The confidence interval for all 10 samples is the lowest, i.e. reaching 9% of the median value (95% confidence level). Values for mean and maximum confidence intervals for each combination of samples are given in Table S2.

182

183 Discussion and conclusion

The median concentration of microplastics in the Gave de Pau River $(3.26 \pm 0.21 \text{ MPs/m}^3)$ 184 was ten times higher than concentrations measured in the Danube River between Vienna and 185 Bratislava (0.32 MPs/m³; Lechner *et al.*, 2014; exposition time and volume of water filtered 186 not mentioned) and in the Seine River downstream of the heavily-populated Paris Basin (0.35 187 MPs/m³, mesh size: 330 μ m, net towed behind a motor boat for 15 minutes at about 2 m/s for 188 a volume of filtered water ranging from 182 to 200 m³, Dris et al., 2015). Contrary to Dris et 189 al. (2018), who observed a decrease in variability between samples for longer exposure times, 190 our results do not present such a correlation. A comparison with other studies proves to be 191 difficult due to the significant differences in methods, extraction protocols and units (Li et al., 192 2018). The work presented in this study is based on 16 samples, all extracted on the same day, 193 194 at the same location and within a time window for which the river flow and pollution can be considered in a steady state. This value does not therefore reflect temporal variability in 195 microplastics concentration (van Emmerik et al., 2018) and hence does not reflect the mean 196 197 annual load of the river either.

The large quantity of microplastics in our samples can be explained by the fact that two 198 199 former landfills, located near the riverbanks upstream of our site, were swept by the river 200 during several flooding events, the most recent occurring in February 2018. No chemical characterisation was conducted. Visual identification of microplastics can sometimes lead to 201 substantial errors; however, given the size range target (> 330 µm), and because only 202 fragments were considered, the risk of under or over estimation remains very low. Moreover, 203 204 this bias is similar for all samples as the same operator handled all manipulations and it is therefore possible to compare them. Microplastic concentrations for the three net exposition 205 times could not be shown to be statistically different. This result should be handled with 206 207 caution due to the low number in some groups of samples. This point in mind, it seems that the filtration of approx. 35 m³ of water is appropriate for future samplings under similar 208 209 conditions (similar flow condition and similar degree of microplastic contamination of the 210 river). This lower volume of water could reduce the time devoted to sampling, sample treatment and identification. To verify that the conditions are similar to the ones describe is 211 this study, researchers are urged to use this method to investigate whether those results apply 212 to their rivers or not before starting a monitoring and setting up a monitoring station at a 213 214 specific river location. The number of samples taken for each volume of filtrated water should 215 ideally be the same to ensure statistics robustness. When choosing the different volumes of 216 water to be filtered as a first test, mind that a long exposure of the net might allow the catch of rare particles, the setback being that the net could clog rapidly in case of high concentration of 217 218 suspended matter.

The 10 samples of 59.4 m^3 (5 minutes) extracted gave us the opportunity to study the correlation between the number of samples collected and associated uncertainties. For triplicates, the mean confidence interval reached 15% of the median value. The confidence interval decreased quickly the higher the number of samples: it can drop to 9% of the median value for all 10 samples. Thus, collecting triplicates seems to be a reasonable optimum that combines an acceptable error percentage and time efficiency. These results might depend on the microplastic load of the river, therefore prompting the need to conduct similar analyses on other rivers. The order of magnitude of uncertainties reported in this paper can potentially be affected by: i) time exposure, ii) river hydrodynamic conditions (low vs. high water levels), and iii) the amount of microplastics sampled. As such, similar work must be performed under other conditions.

This study has reported, for the very first time, uncertainties related to microplastic sampling 230 in rivers. The median concentration of microplastics in the Gave de Pau River recorded in this 231 232 study $(3.26 \pm 0.21 \text{ MPs/m}^3)$ is day-specific and do not allow for any extrapolation. Under similar conditions, a 3-minute exposure time for the net (filtration of approx. 35 m³) and the 233 collection of triplicates seem to offer a reasonable optimum, by virtue of combining an 234 acceptable error percentage and time efficiency. Even if these results have to be confirmed by 235 increasing the number of samples per time exposure, this paper share a methodology that 236 237 helps design better microplastics studies. These results will help to set up long term monitoring, determine microplastic fluxes and highlight the spatial difference between sites. 238

239

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245

246 Conflicts of Interest

247 The authors hereby declare the absence of any conflicts of interest.

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