Assessment of sample frequency bias and precision in fluvial flux calculations – an
 improved low bias estimation method

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11 Abstract

12 Despite the importance of calculating the flux of solutes and particulates through the global fluvial network the number of studies that have considered the bias and precision of any 13 method is limited. Furthermore, no study has, on the basis of the bias of the method, proposed 14 new methods with a lower bias nor considered the implications of the bias estimation for 15 existing published studies. Using 3 years of high frequency data (hourly) for dissolved 16 organic carbon (DOC) this study systematically degraded the data and recalculated the flux 17 for varying sample frequencies and considered a range of interpolation, ratio and 18 extrapolation methods. The results show that: 19

i) Interpolation and ratio methods showed a consistent, small bias for sampling frequencies
up to every 14 days, but bias rapidly increased for lower sample frequencies with the flux
estimates being between 40 and 45% of the "true" flux at 31 day (monthly) sampling.

23 ii) The best ratio method was based upon correction against an unrealistic assumption that24 river flow was normally distributed.

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25 iii) Extrapolation methods based on fixed sampling period monitoring proved to be erratic but26 no better than interpolation methods.

Based upon the nature of the sources of variation within the flow and solute datasets wepropose the following method for calculating the fluvial flux (F) of a solute:

$$F = KE(C_i)Q_{total}$$

Where: Q_{total} = the total flow in a year (m³/yr); E(C_i) = the expected value of the sampled concentrations (mg/l); and K= a conversion factor. This new method preserved all the available flow information and had a bias of as low as 8% for monthly sampling. When the method was applied to DOC flux from Great Britain bias correction meant a 97% increase in the national flux over previous estimates.

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35 Keywords: dissolved organic carbon; DOC; DOC budget

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37 Introduction

There are many methods that have been proposed for the calculation of dissolved or particulate fluxes from rivers. Methods used to estimate fluvial fluxes of a given component can be broadly classified into either interpolation methods (e.g. Webb et al., 1997) or extrapolation methods (e.g. De Vries and Klavers, 1994). Some studies (OSPAR, 1998) also consider ratio methods alongside interpolation and extrapolation methods, however, ratio methods are usually based upon a correction factor being applied to an interpolation method. A summary of current methods is given in Table 1.

Given the importance of flux estimates for studies of solute and solid material transport through the fluvial network, many studies have proposed flux calculation methods and some studies have also assessed the accuracy and precision of different approaches for different determinands and over different sampling regimes. A number of these have been 49 used with correction factors in order to remove sampling or calculation bias (e.g. Ferguson, 1986). These flux correction factors have been considered from a number of perspectives. 50 Cohn (2005) considered the problem of solute flux estimation when measured concentrations 51 are close to detection limits but most studies have considered the differences between 52 sampling frequencies and methods. The quality of methods and sampling frequencies need to 53 54 be discussed in two ways. Firstly, the accuracy can be considered as the difference between the true load and estimated load and represents the systematic bias. Secondly, the precision of 55 the method or sampling frequency which represents the spread of the load estimates about a 56 certain value and represents the consistency of the load estimates. In many studies that 57 discuss uncertainty in flux estimation due to changing method or sampling frequency it is the 58 precision that is described and not the bias or accuracy. An example of this is, Webb et al. 59 (1997) who considered 5 interpolation and 2 extrapolation methods and found that for 60 suspended sediment flux estimation extrapolation methods gave the least biased results and 61 that bias increased with decreased sample frequency, but they could give only precision 62 estimates for solute fluxes because there was no "true" value for each determinand against 63 which to compare the bias. Similarly, Littlewood et al., (1998) could only trace precision with 64 changing sampling frequency with "indicative" curves. Littlewood (1995) and Webb et al. 65 (2000) generated synthetic concentration time series, tested a number of flux estimation 66 methods and suggested that interpolation methods were generally more reliable and less 67 prone to errors than the more complex extrapolation methods. They suggested that 68 extrapolation methods work best where a good rating curve between concentration and flow 69 70 can be found, but this is not normally the case for determinands that exhibit a strong seasonal 71 component. However, neither study could comment on bias and both studies relied on the error structure that was in the synthetic time series, i.e. if the synthetic time series did not 72 include all the components of the variation in the data then the error estimation in the flux 73

74 calculation would have itself been biased. Johnes (2007) considered 17 catchments where there was daily measurement of phosphorus but had no sub-daily data and had to assume that 75 "method 5" (Littlewood, 1995) was the true value and only considered precision but not bias. 76 Skarbøvik et al. (2012) considered a record of daily of suspended sediment and although they 77 were able to suggest that the least biased were extrapolation methods compared to 78 interpolation methods this result was not true across all sampling frequencies and years. Burt 79 et al. (2011) using daily flow and nitrate concentration data did consider both bias and 80 precision to show that seven-day sampling gave a high level of precision with 95% of flux 81 estimates within 2.7% of the "true" flux (based on a complete set of daily samples). Even 28-82 day ("monthly") sampling had 95% of values within 6.5% of the mean: a surprisingly high 83 84 level of precision given the infrequency of the sampling. In terms of accuracy, they showed 85 that the mean flux derived from 7-day sampling was very close to the "true" flux, but the 14day and 28-day means underestimated the "true" load. Kulasova et al. (2012) did have sub-86 daily measurement of nitrate and total phosphorus and although they discuss precision 87 relative to extrapolation methods they did not consider bias of the methods they used. 88 Cassidy and Jordan (2011), with sub-daily measurement of phosphorus, considered both bias 89 and precision in their approach and thus showed bias with decreasing sampling frequency 90 with bias of up to 60% upon weekly sampling (ie. 60% lower than the true value) and high 91 uncertainty for all sampling frequencies except for near continuous monitoring. 92

93 Several studies have recommended or considered adaptive strategies. Kronvang and 94 Bruhn (1996) suggested taking samples hydrologically rather than on a calendar basis and a 95 number of studies (Cooper and Watts, 2002; Skarbøvik et al., 2012) have suggested including 96 flood samples alongside regular sampling. Without a "true" load estimate the authors of 97 studies of adaptive strategies cannot comment on their bias or precision, and indeed, Cassidy 98 and Jordan (2011) found an over-estimation, or positive bias, when flood samples were

included in flux estimation. Roberts (1997) considered the sampling frequency within the 99 UK's Harmonised Monitoring Scheme (Simpson, 1980) and compared calculation of annual 100 fluvial phosphate flux based on 52 weekly samples, 4 weeks of daily samples and 4 days of 101 hourly samples, but concluded that for pragmatic and financial reasons sampling was 102 103 degraded to monthly. Flux estimation methods based upon adaptive strategies will not be appropriate for many monitoring programmes designed to estimate flux simply because most 104 national monitoring programmes are based upon regular sampling. Moater et al. (2012) 105 considered the precision and bias of differing sampling frequencies given daily sampling at 106 125 sites, the study did not consider differing methods of flux estimation although it did 107 propose an empirical approach to flux estimation based on measures of the flow duration 108 109 curve.

110 Although the studies above have considered particulates and a range of solutes none of the studies above have considered the precision and bias of flux estimation methods 111 relative to dissolved organic carbon (DOC). Recent interest in DOC has been based upon the 112 113 observation that DOC concentrations in the surface waters across the northern hemisphere have been increasing over the last few decades (eg. Monteith et al., 2007). Studies of flux 114 have been used to consider the causes of this rise (eg. Worrall et al., 2008) but the flux of 115 DOC is also a vital component of the carbon cycle of the terrestrial biosphere (eg. Worrall et 116 al., 2009a) and changes in the DOC flux across a watershed is component in the estimation of 117 the flux of greenhouse gases to the atmosphere (Worrall et al., 2012). The aims of this study, 118 therefore, are multiple. Firstly, to assess the bias and precision of flux calculation methods 119 relative to DOC: a solute not previously considered. Secondly, to not only provide a measure 120 121 of bias but to assess how the change in the bias of the method with changing sampling frequency could be used to direct the method of flux estimation. 122

124 Approach & Methodology

This study was able to consider sub-daily monitoring of the DOC concentrations and river 125 flow over a 3 year period. From hourly data a "true" value of load was calculated for each of 126 the available years. The time series of DOC concentration and river flow were then 127 systematically degraded so that combinations of data were selected based upon a single 128 sample being collected each day with frequency of sampling from 1 day to 31 days, i.e. from 129 daily to monthly. Within each day, pairs of data were selected at random and 100 sets of flux 130 calculations for each of the three years were made for each sampling frequency for each year 131 and for each flux estimation method. Results from each sampling frequency were compared 132 to the "true" load flux for each year for each sampling frequency for each flux estimation 133 method in order to give an estimate of the bias, and precision, with increasing sample 134 135 frequency.

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137 Study site

Data were collected for the River Dee just upstream of the city of Chester where data could 138 be twinned with flow records (Figure 1). The river Dee at Chester has a catchment area of 139 1674 km², with annual average rainfall (1961 – 1990) of 1143 mm with 10% of the 140 catchment being classed as mountain, heath and bog which can be considered as the major 141 DOC considered in this study (National Riverflow 142 source of the Archive http://www.ceh.ac.uk/data/nrfa/). The concentration data were collected hourly between 1st 143 January 2009 and 31st December 2011 and the flow data every 15 minutes over the same 144 period. Over the 3 year period the median river discharge = $15.5 \text{ m}^3/\text{s}$ with 95% exceedence 145 flow = 5.8 m^3 /s and 5% exceedence flow = 101.4 m^3 /s. For the DOC concentration over the 146 3 year period the median concentration = 11.2 mg C/l with 95^{th} percentile = 21.3 mg C/l and 147 the 5^{th} percentile = 4.6 mg C/l. The DOC concentration data were collected using an UV 148

149 absorbance probe (ABB AV400) calibrated for DOC concentration using potassium hydrogen phthalate on a regular basis. The calibration between UV absorbance and DOC concentration 150 is not necessarily stationary and would be a source of uncertainty above and beyond that due 151 to flux estimation method, however further discussion is beyond the scope of this study as 152 this study concerns the error in the flux estimation method and not the error in the 153 measurements. The error in the relationship between UV absorbance and DOC used within 154 the measuring probe would only be of concern to this study if it had a systematic bias that 155 was true whenever, and wherever, it was used for DOC measurement and in the UK DOC 156 157 concentration is not normally made using this approach (Simpson, 1980).

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159 Flux estimation methods

160 The study did not choose to compare all possible interpolation and extrapolation methods 161 rather selected methods for contrast. We choose "method 2" (Littlewood et al., 1998) which 162 is a simple interpolation method:

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$$F = K \sum_{1}^{n} \frac{C_i Q_i}{n}$$
 (i)

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166 Where: C_i = the instantaneous concentration (mg/l); Q_i = the instantaneous discharge (m³/s); 167 n = number of samples; and K= a unit conversion factor.

More advanced interpolation methods are interpolation methods that attempt to correct for the conditions at the time of sampling – sometimes referred to as ratio methods = and the most used or recommended is "method 5" (eg. Johnes, 2007):

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$$F = K(\sum_{i=1}^{n} C_{i} Q_{i}) \left(\frac{Q}{\sum_{i=1}^{n} Q_{i}}\right)$$
(ii)

Where: \overline{Q} = the average discharge for the period (m³/s); n = the number of samples; and K = 174 conversion factor. The interpolation "method 5" is "method 2" with the ratio factor 175 correction, the correction factor compares the flows at times of sampling to the other 176 measured flows. However, "method 5" assumes a normal distribution of flow data which 177 would be unusual for most known rivers. Therefore the general case would be: 178

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$$F = K(\sum_{i=1}^{n} C_{i}Q_{i})\left(\frac{E(\bar{Q})}{E(Q_{i})}\right)$$
(iii)

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Where: E(x) = expected value of x. The form of the expected value would differ depending 182 on the type of distribution that best described the distribution of the sampled and total flows. 183 184 For the purpose of this study normal; log normal and gamma distributions were considered. Note that for a normal distribution the expected value would be the arithmetic mean; for the 185 log-normal distribution the expected value would be the geometric mean; and the gamma 186 distribution the expected value $-k\theta$, where: k = shape factor and $\theta = scale$ factor. Equation 187 (iii) is "method 5" when a normal distribution is considered and henceforward this study will 188 refer to "method 6" when E(x) is based upon a log normal distribution and "method 7" when 189 E(x) is based upon a gamma distribution. The normal and log normal distributions were fitted 190 to sampled and total measured flows by method of moments and the gamma distribution was 191 fitted by maximum likelihood. 192

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A log-log rating curve approach was used as an extrapolation method for calculating flux estimates: 194

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$$logC_i = AlogQ_i + logB$$
 (iv)

Where A and B are constants. Equation (iv) was fitted to the data sampled for each samplingfrequency and then applied to the entire flow record for each year of the available data.

A comparison of the concentration, flow and flux cumulative distributions (Figure 2) 200 suggests that the flux distribution was really dominated by the flow and not the distribution of 201 the concentration and therefore, methods that do not preserve the distribution of the flow or 202 do not maximise the information available in situations where flow is more extensively 203 sampled than the concentration will show considerable bias. For example, when calculating 204 the DOC flux across the UK, Worrall and Burt (2007) used the data collected as by the 205 Harmonised Monitoring Scheme (Simpson, 1980) and for most catchments for most years 206 only monthly samples were available while in most cases daily flows were available. 207 Therefore, this study proposes the following method: 208

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210
$$F = KE(C_i)Q_{total}$$
 (v)

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Where: Q_{total} = the total flow in a year (m³/yr); E(C_i) = the expected value of the sampled 212 concentrations (mg/l). As above the form of the expected value depends upon the distribution 213 chosen to describe the sampled concentrations and as above for the purpose of this study 214 normal, log normal and gamma distributions were chosen. Equally, if the flux is dominated 215 by variation in the flow and sample sizes in any one year are small then taking data from 216 217 previous years sampling could improve the estimation of the concentration distribution. The latter would be true as long as the concentration time series was stationary or could be made 218 stationary. For the purposes of this study it was assumed that the DOC time series was 219 220 stationary for the 3 years of the study. So $E(C_i)$ was calculated not only for normal, log normal and gamma distributions but also for each of these based upon sampling from all 221 three years of available data or from each year of available data. To assess the best 222

combination this was considered as a factorial experimental of four factors and their 223 interactions. The four factors were: the year the data were collected (with three levels - 2009, 224 2010 & 2011); the averaging method (with three levels - normal, log-normal and gamma 225 distributions); assessment period (with two levels based upon averaging over 1 or 3 years of 226 data); and the frequency of sampling (with levels from 1 to 31 days). By considering as a 227 factorial experiment we can identify not only significant effects, by analysis of variance 228 (ANOVA), but also the size of the effect and so identify which combination of approaches 229 would lead to lowest possible bias. 230

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232 Application of results

The findings of the above studies will be applied to consideration of DOC flux from the UK. The flux of DOC from the UK has been calculated from 1975 to 2007 by Worrall et al. (2009b). However, the approach was based upon "method 5" and performed only an analysis of precision but not of bias in the method. In Worrall et al. (2009b) the flux estimates were based upon data from the harmonised monitoring network (Simpson, 1980) and only sites where sampling was at least monthly in any one year were included, although sampling frequencies were as regular as sub-weekly for some sites for some years.

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241 **Results**

The comparison between the "true" load and the results from interpolation methods shows a time course in the systematic bias of the results with overestimation on daily sampling with systematic underestimation occurring between day 9 and 14, but by day 31 the methods were between 40 and 45% of the "true" load (Figure 3) an underestimate of up to 60%. Of the interpolation methods, "method 5" was the least biased overall and was always the highest estimate of the load. The reason for the better performance of "method 5" is not because of 248 inherently better method of estimation rather because it assumes a normal distribution for flow and so it systematically overestimates the expected value of the flow distribution (Figure 249 4). So by using a normal distribution the calculation fortuitously skews itself towards higher 250 values and so preserves higher estimates of the annual flux. The imprecision of the 251 interpolation methods, as expressed as the 5th to 95th interpercentile range, increases with 252 decreasing sampling frequency and by a sampling frequency of 15 days the imprecision is 253 between ± 15 and 20% which is very similar to the precision of the interpolation methods 254 estimated by Webb et al. (1997), for a sampling frequency of 31 days the imprecision reaches 255 256 ±12% (Figure 5).

For the extrapolation method the change across the sampling frequencies shows that 257 the pattern of bias was erratic but for monthly sampling the bias of an extrapolation method is 258 no better than interpolation with median estimates between 19 and 55% of the "true" load 259 (Figure 6). The precision of the extrapolation shows that for all the extrapolation methods 260 considered that imprecision rose rapidly as sample frequency dropped to every 5 days after 261 which imprecision varied between ± 20 and 40% of the median (Figure 7). The erratic nature 262 of the response of using equation (iv) for regularly monitored data must be ascribed to the 263 sensitivity of the method to the inclusion of high flow events within the sample. The result 264 would appear less erratic if the study had perhaps used 500 rather than 100 sets of samples, 265 but such an increase would not have changed the result with respect to the bias of the method. 266

Estimates based upon equation (v) showed that in general the method provided a less bias result for sampling frequencies of greater than 14 days (Figure 8) and indeed the analysis of variance (ANOVA) from the factorial analysis of the results found no significant differences between sampling frequencies up to 14 days for all approaches based upon equation (v). It is obvious that one of the versions of equations (v) based upon the annual estimate of the expected value from fitting a gamma distribution to the sample data was very

biased with estimates being 39% of the "true" load with sample frequencies less than 2 days 273 (Figure 8). This large bias is perhaps for a very practical reason, i.e. the difficulty of 274 estimating a gamma distribution with limited data. Conversely, the best overall method was 275 equation (v) based upon the expected value of gamma distribution based on the data sampled 276 from all years of the available data with flux estimates being 92% of the "true" load even at 277 sampling frequencies as low as every 31 days. However, when only one year of data was 278 available the best method was Equation (v) based upon a normal distribution, again the 279 reason for this is most likely that the normal distribution systemically overestimates the 280 expected value compared to that of the population and thus keeping the flux estimates higher 281 and so closer to the "true" load. The ANOVA of the factorial results showed that all factors 282 had a significant effect upon outcome and degree of bias. Over all sampling frequencies the 283 284 most accurate, least biased method, was confirmed as equation (v) based on the expected value calculated from the gamma distribution of all 3 years of available data and when the 285 interaction between sampling frequency; averaging method; and assessment period shows 286 287 that at the lowest sample frequencies the best method is still averaging over all 3 years of available data based upon a gamma distribution. However, the ANOVA showed that there 288 was a significant difference in the bias of methods based upon equation (v) between the years 289 of available data (i.e. between 2009, 2010 and 2011). But, even in the year which had the 290 significantly highest bias (2009) the best method was still that based on equation (v) using the 291 expected value from the fit of a gamma distribution to all 3 years of available data methods 292 based on equation (v) – in this case the bias was 19% and the value would be 81% of the true 293 value. To summarise, the true load based upon hourly data in 2009 was 16.8 ktonnes C/yr, 294 295 based on equation (v) with the expected value based upon the fitting of a gamma distribution to all 3 years of data gives a value of 14.4 ktonnes C/yr whereas the best performing ratio 296 method gave a value of 7.8 ktonnes C/yr. 297

The methods based upon Equation (v) were dramatically more precise than those for 298 the previous methods with imprecision rising almost linearly with declining sample 299 frequency but to only $\pm 2\%$ after 31 days (Figure 9). The reason for the high precision of the 300 method expressed in Equation (v) is that the method was deliberately chosen to be dominated 301 by the major source of variation, i.e. the flow. The total annual flow is constant between 302 sampling frequencies and so even though the estimation of the expected value of the 303 concentration changes and becomes less precise its effect on the overall flux estimation is 304 minimal. 305

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307 Application of the results

Given that previously "method 5" had been used to calculate the DOC flux and so the result 308 in Figure 3 can be used to correct the bias for each site for each year where there was at least 309 monthly sampling, i.e. for a site in one year where there was weekly sampling the above 310 result would suggest the flux would be overestimated by "method 5" but if in a subsequent 311 year there was only fortnightly sampling then "method 5" would underestimate the flux of 312 DOC from that site. Thus bias correcting each year of data from each site and then 313 amalgamating to the national level suggests that were previously the DOC flux varied from 314 812 to 1920 ktonnes C/yr after bias correction the DOC flux varied from 797 to 3090 ktonnes 315 C/yr (Figure 10) – all values are given to 3 significant figures for comparison purposes. 316 Firstly, the average correction over the course of the entire time series was 1.965 (97%) thus 317 showing by how much the flux of DOC has been underestimated in previous studies. 318 Secondly, the range of DOC flux has a lower minimum value than the uncorrected values and 319 320 this can be seen to occur in 1975 when the average sampling frequency was 23 per year, i.e. once every 17 days. As for many sites the sampling frequency was greater than every 14 days 321 and so the flux was being overestimated by "method 5". The annual average sample 322

frequency decreases across the period of the record with an average sample frequency of 323 every 17 days in 1975 to once every 29 days in 2005 with a significant trend and average 324 decrease in sampling frequency of 1 day less frequent every 3 years. Given the result in 325 Figure 3 a significant shift in sample frequency represents a significant increase in the bias 326 correction required over the course of the time series. Therefore, when the uncorrected series 327 shows no significant trend over time and this may simply be because of decline in the sample 328 frequency. In the case of the national annual DOC flux time series this was not the case: there 329 is no significant trend in either the uncorrected or corrected series. Given the result above on 330 the changing precision of time series it should be possible to better constrain the precision on 331 the flux time series, however, as pointed out by Worrall and Burt (2007) the dominant error 332 in the time series was the upscaling of the individual flux records for individual sites for each 333 334 year to the scale of the country and that estimation error cannot be improved by any of the 335 above results.

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337 Discussion

The implications of this study are clear, that many flux estimates of DOC and other solutes or 338 particulates are serious underestimates. Previous studies that considered the bias or 339 imprecision of flux estimates (eg. Cassidy and Jordan, 2011) have shown similar inaccuracy 340 and bias in low frequency sampling as demonstrated in this study but have not gone further to 341 demonstrate the implications of what they found for any of the studies that used the biased or 342 imprecise methods. The implications for many studies is obvious and has already been 343 illustrated with reference to the authors study of DOC flux from the UK: any study that used 344 either interpolation, or indeed extrapolation methods, on data with a sampling frequency of 345 less than once every 2 weeks must be considered to have severely underestimated the flux. 346 For example, the data used in the estimate of flux came from the harmonised monitoring 347

348 scheme (HMS - Simpson, 1980) which was established in order to assess fluxes from Europe to its surrounding oceans. The HMS network use "method 5" and what has been observed for 349 the UK data collection is that the sampling frequency has declined, perhaps for practical and 350 economic reasons, to monthly from an average of fortnightly over the period of the scheme. 351 Worrall and Burt (2007) have already pointed out that the HMS reported fluxes are an 352 underestimate of fluxes because they do not allow for the unsampled catchments and in the 353 UK the sampled catchments of the HMS only represent just over 60% of the UK land area 354 and no correction for river flows from the other 40% was made. Now it would be possible to 355 suggest that many of the fluxes calculated under these schemes are less than 50% of the true 356 value because of sampling bias at such low frequency sampling. The length of the HMS 357 records (back to 1974) could represent one solution to the problem of low sample frequency 358 359 as this study has shown that in stationary data then the longer the period over which an expected value of the concentration distribution was calculated then the better the estimate of 360 the flux. Of course that requires stationarity over several years if not decades of data but it is 361 362 relatively trivial to make a time series stationary.

In some cases the implications of the above study will have little impact. Worrall et 363 al., (2009a) calculated the DOC flux from a peat-covered catchment as part of estimating the 364 carbon budget of the catchment. Worrall et al. (2009a) used "method 5" on samples taken 365 weekly and suggested a standard error on the estimation of $\pm 11\%$. The results of this study do 366 suggest that interpolation was probably the best method for calculating a flux given such a 367 sampling regime but at 7 days "method 5" would overestimate the flux by 10.8% with an 368 imprecision of $\pm 8\%$, i.e. the imprecision and indeed the flux had been overestimated. An over 369 estimation on DOC flux in that study would mean that DOC varied from 11.3 to 77.3 tonnes 370 C/km²/yr rather than 12.5 to 85.9 tonnes C/km²/yr as reported – again 3 significant figures are 371 shown for comparative purposes. Since DOC flux is a loss pathway within the C budget then 372

its overestimation means that the size of the C sink represented by the peats of this catchment has been underestimated. It should also be noted that in the study the dissolved CO_2 was also calculated by "method 5" but its flux peaked at 15.1 tonnes C/km²/yr giving only a maximum bias of 1.5 tonnes C/km²/yr.

There are cases where flux estimation will not have the same inaccuracy as discussed above. For example, Howden et al. (2010) examined the longest water quality record in the world and assessed the changes in nitrate flux over 120 years based upon monthly data, however, the monthly data in that case were a true average and not a monthly spot sample, i.e. the real sampling frequency was actually sub-weekly and so the bias would be on the order of +5% rather than the -60% which might be assumed if the detail of the sampling strategy were not understood.

384 What is not clear from this study is the transferability of this result, i.e. could the pattern of bias and precision found here be indicative, and therefore applicable in other 385 catchments or for other determinands? Firstly, the result found here is similar in magnitude 386 and direction to the results of Cassidy and Jordan (2011) and Moatar et al. (2012). The 387 important result of this study is that the present problem with most flux calculation methods 388 is not the approximation of the type of distribution, i.e. there was little improvement by 389 changing between estimation based upon normal, log normal or gamma distributions rather 390 that the greatest improvement was achieved when all the available flow data were used. The 391 reason for the success of the approach outlined in Equation (v) is that the approach preserved 392 that maximum amount of information from the variable with the greatest variation. Any flux 393 estimation method is in essence the multiplication of a concentration and a flow variable and 394 395 Goodman (1960) shows that the variance of any product of two variables will be dominated by the variable with the greater variance and it is easy to demonstrate that the variance of 396 flow is greater than that of DOC concentration. Therefore, the result presented here would 397

have general applicability wherever the variance in the flow dominates the estimation whichfor most water constituents would be the case.

400 One way of demonstrating the advantage of Equation (v) is to calculate how well each 401 method represents the true variance in the original time series. Goodman (1960) proposed 402 methods for the calculation of the variance of a product, where for independent variables the 403 variance in the product of independent variables is estimated by:

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$$V(xy) = \bar{x}^2 V(y) + \bar{y}^2 V(x) + V(x) V(y)$$
 (vi)

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407 And where the variables are not independent by:

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$$V(xy) = \bar{x}^2 V(y) + \bar{y}^2 V(x) + 2\bar{x}\bar{y}E_{11} + 2\bar{x}E_{12} + 2\bar{y}E_{21} + E_{22} - E_{11}^2$$
 (vii)

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Where: V(x) = variance in x; and $E_{ij} = E\{(\Delta x)^i (\Delta y)^j\}$ given $\Delta x = x - \bar{x}$. In 2009 the variance in 411 the river flow was 1589 while that for the DOC concentration was only 29 while that for the 412 flux series was 548400. Applying Equation (vii) for flux calculation by Equation (i) to 100 413 random samples of concentration and river flow for a 30 day sampling period gave an 414 415 average estimate of the variance in the flux time series that was $75 \pm 12\%$ of the true value of the variance in the original flux series. Alternatively, considering flux calculation by 416 Equation (v) then it is appropriate to use Equation (vi) because the data are no longer paired 417 and so independent of each other then with 100 sets of randomly sampled concentration data 418 and the river flow series for 2009 the average variance was $106 \pm 9\%$ of the true value of the 419 variance. Therefore, Equation (v) better preserves the variance in the flux time series. 420

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423 Conclusions

The study has shown that interpolation, ratio and extrapolation methods for calculating fluvial flux of solutes develop very considerable bias for sampling frequencies greater than every 14 days, with underestimation by 60% observed for dissolved organic carbon (DOC) flux in monthly samples. On the basis of an assessment of the source of variation within the flux calculation we can show that a simpler method based upon all the available flow data is less biased than existing methods with as little as 8% underestimation even on monthly spot sampling.

431

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517 Figure 1. Location of the monitoring and gauging station for the high frequency data used in 518 this study.

519

- Figure 2. the comparison of distribution of the hourly values of the riverflow, concentrationand fluxes for the study rescaled relative to the maximum observed value of each.
- 522
- Figure 3. Comparison of the relative flux estimation for interpolation methods over samplingfrequencies from daily to one sample every 31 days.

525

526 Figure 4. Comparison of fitted distributions to the observed flow data.

527

528 Figure 5. Comparison of the relative precision for interpolation methods over sampling 529 frequencies from daily to one sample every 31 days.

530

Figure 6. Comparison of the relative flux estimation for extrapolation methods over samplingfrequencies from daily to one sample every 31 days.

533

Figure 7. Comparison of the relative precision for extrapolation methods over samplingfrequencies from daily to one sample every 31 days.

536

Figure 8. Comparison of the relative flux estimation for methods based upon equation (v)over sampling frequencies from daily to one sample every 31 days.

539

540 Figure 9. Comparison of the relative precision for methods based upon equation (v) over 541 sampling frequencies from daily to one sample every 31 days.

- 542 Figure 10. Comparison between the flux of DOC from Great Britain between 1975 and 2005
- 543 calculated with and without bias correction.