



1 Comparison of high frequency, in-situ water quality analysers and sensors with conventional  
2 water sample collection and laboratory analyses: phosphorus and nitrogen species

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9

10 Abstract

11 The long-term collection of water samples for water quality analysis with high precision  
12 laboratory instrumentation is routine in monitoring programmes however, such sampling is  
13 labour intensive, expensive, and therefore undertaken at a low temporal frequency. Advances  
14 in environmental monitoring technology however, mean that it is now possible to collect high  
15 temporal frequency measurements for a wide range of water quality parameters without the  
16 need for the physical collection of a sample. The downside to this approach is that the data can  
17 be subject to more ‘noise’, due to environmental and instrument variables. This raises the  
18 question of whether high frequency, lower precision data are better than low frequency, higher  
19 precision data. This study reports the findings of an investigation of agricultural land drainage  
20 comparing measurements of total phosphorus (TP), total reactive phosphorus (TRP),  
21 ammonium (NH<sub>4</sub>-N) and total oxidised inorganic nitrogen (NO<sub>x</sub>-N) collected using both  
22 equipment in situ and concurrent water samples analysed in the laboratory. Results show that  
23 both in situ PHOSPHAX TP and NITRATAX NO<sub>x</sub>-N instruments can provide comparable



24 data to that measured using samples analysed in the laboratory; however, at high discharge and  
25 low NO<sub>x</sub>-N concentrations, the NITRATAX can under report the concentration. In contrast,  
26 PHOSPHAX TRP and YSI sonde NH<sub>4</sub>-N data were both found to be incomparable to the  
27 laboratory data. This was because concentrations of both parameters were well below the  
28 instruments accurately determinable level, and because the laboratory data at low  
29 concentrations were noisy.

30 Keywords: water quality; phosphorus, nitrogen, ammonium, sensors; in situ; runoff; field  
31 drainage

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### 33 1. Introduction

34 Long-term routine, but infrequent, water quality sampling used widely in strategic scale  
35 monitoring provides insight into longer-term trends (Howden et al., 2010). However, such  
36 sampling fails to capture higher resolution data necessary for insight into hydrological and  
37 biogeochemical processes and responses (Granger et al., 2010) including evidence of non-  
38 stationarity, self-organisation, and fractals (Harris and Heathwaite, 2005; Milne et al., 2009;  
39 Kirchner and Neal, 2013). Advances in environmental monitoring technology mean that it is  
40 now possible to collect high resolution measurements of a wide range of water quality  
41 parameters, providing detailed insight into hydrochemical temporal dynamics. Technologies  
42 vary depending on the parameters being measured, but typically include, automated wet  
43 chemistry apparatus in situ (e.g. for phosphorus (P) analysis) or ultra-violet optical sensors (e.g.  
44 for total oxidised nitrogen) (Palmer-Felgate et al., 2008; Donn et al., 2012; Carey et al., 2014;  
45 Skeffington et al., 2015; Bieroza and Heathwaite, 2015; Mellander et al., 2016). Frequency of  
46 measurements vary, ranging from every minute (or less) to hourly, depending upon the  
47 parameter, but are more typically undertaken at 15-minute intervals. Wet chemistry in situ



48 analysers and optical sensors have been shown to deliver important insights into nutrient  
49 fraction dynamics in response to runoff (Mellander et al., 2015) and catchment management  
50 (Perks et al., 2015). High resolution sampling and analysis in situ captures a broader range of  
51 pollutant concentrations than routine infrequent sampling and thereby elucidates hysteresis,  
52 diurnal patterns and non-storm dependent transfers (Heffernan and Cohen, 2010; Bende-Michl  
53 et al., 2013). Monitoring in situ can be used to identify pollutant transfer typologies. For  
54 example, Jordan et al. (2005) used in situ wet chemistry analysis to detect three types of total  
55 P (TP) transfer events: chronic storm-independent transfers reflecting on-farm slurry and  
56 fertiliser applications; acute storm-dependent transfers associated with agricultural diffuse  
57 pollution, and; acute storm-independent transfers reflecting specific incidental pollution  
58 events. In situ devices remove sample storage requirements and provide a means of avoiding  
59 water sample storage-associated chemical transformations (Bende-Michl and Hairsine, 2010).

60 Previously, studies were limited to the collection of water samples either manually or using  
61 automated water samplers, and then transfer of samples to laboratories for analysis by wet  
62 chemistry and colourimetric methods. However, despite transforming the hydrologic sciences  
63 over the past 50 years (Rode et al., 2016), questions remain about the precision of  
64 measurements made using these technologies relative to standard sample collection and  
65 laboratory analysis. The traditional auto-sampler approach followed by laboratory analyses of  
66 nutrient content can carry risks and uncertainties associated with a number of problems,  
67 including small sampling volume, preferential sampling effects, limited coverage of the stream  
68 cross-section and transformation risks during storage in conjunction with time delays between  
69 sample collection and subsequent laboratory analyses (Kotlash and Chessman, 1998; Harmel  
70 et al., 2006; McMillan et al., 2012). Storage-associated transformations are caused by a range  
71 of physical and biochemical processes including hydrolysis, sorption, precipitation, microbial  
72 uptake or release and complexation (Jarvie et al., 2002a; Harmel et al., 2006). Previous work



73 (McMillan et al., 2012) has suggested that biogeochemical effects during sample storage can  
74 contribute more to uncertainty than errors due to preferential sampling or lower extraction of  
75 sediment-bound nutrients. The greatest proportional losses of dissolved nutrients in stored  
76 water samples occur when concentrations are low, with losses up to 50 % for nitrate and 67 %  
77 for soluble reactive P after six days of storage with no refrigeration (Kotlash and Chessman,  
78 1998). Such uncertainties are also reported by Lloyd et al. (2016) who describe an almost 10-  
79 fold increase in uncertainty of both nitrate and TP loads measured over 2 years in a river in the  
80 U.K. when comparing laboratory and automated sensor data.

81 This raises the question of whether high frequency, low precision data is better than low  
82 frequency, high precision data. Rode et al. (2016) recognise that there are major issues related  
83 to calibration of automated sensing equipment and the need for regular servicing, along with  
84 a pressing need for the development of automated tools and standards for data quality  
85 assurance (Campbell et al., 2013). There is still much work to be done in quantifying the  
86 precision of automated water quality sensors, and accordingly, herein we report the findings  
87 of a study comparing measurements of TP, total reactive P (TRP), ammonium nitrogen (NH<sub>4</sub>-  
88 N) and total oxidised inorganic nitrogen (NO<sub>x</sub>-N) collected using both automated equipment  
89 and concurrent water samples analysed in the laboratory. We test the hypothesis that in situ  
90 measurements of these parameters can be as precise as those acquired by laboratory analysis  
91 of manually collected water samples analysed using standard laboratory techniques.

92

## 93 2. Material and Methods

### 94 2.1 Study Site

95 The study was undertaken on the North Wyke Farm Platform (NWFP), an instrumented  
96 research grassland farm of 63 ha, split into 15 hydrologically isolated sub-catchments, over



97 which three different 21 ha livestock and grassland management systems are imposed (Orr et  
98 al., 2016). From April 2013 to July 2015, all 15 NWFP sub-catchments were assigned to one  
99 of three treatments: (i) permanent pasture ('green' farmlet); (ii) increased use of legumes  
100 (blue farmlet), and; (iii) innovation (red farmlet), via a gradual and planned re-seeding  
101 campaign (Figure 1a). The soils of the NWFP are clay loams (Figure 1b). Within each sub-  
102 catchment a range of instrumentation takes measurements on water, air and soil parameters in  
103 situ, much of this data being at a high temporal resolution (15 mins).

## 104 2.2 Sampling Strategy

105 For this study, one sub-catchment from within each of the three management systems was  
106 chosen for investigation (numbered 2, 5 and 8; Figure 1b) of drainage caused by a rainfall  
107 event on the 3rd December 2015. A tipping bucket rain gauge (Adcon, Austria) located in the  
108 centre of each catchment measured the rainfall at a resolution of 0.2 mm per tip. During this  
109 event, measurements of discharge, TP, TRP, NH<sub>4</sub>-N and NO<sub>x</sub>-N were taken using the  
110 instrumentation in situ. Auto-samplers (Teledyne ISCO, New England, USA) were used to  
111 sample the discharge automatically at pre-determined flow thresholds. Manual grab samples  
112 were also collected throughout the discharge event and both these and the auto-sampler  
113 samples were analysed in the laboratory. Grab samples were taken to both sample the  
114 discharge before and after the main storm drainage, and a sub-set during the storm drainage  
115 were taken at exactly the same time as the automated in situ analysis so as to generate paired  
116 results for TP and TRP. Grab samples were kept cool, and a sub-sample filtered through a  
117 0.45 µm cellulose nitrate filter, before all samples were analysed in the laboratory within 48  
118 hours.

## 119 2.3 Measurement of Discharge



120 Each of the sub-catchments drain, via French drains, to a monitoring station where H flumes  
121 are located with a capacity designed for a 1 in 50-year storm event. The flumes intercept and  
122 channel drainage in such a way that discharge can be determined by a rating curve calculated  
123 based on the the height of the liquid in the flume. Drainage in this context is defined as all the  
124 water that moves from the sub-catchment through the flume irrespective of its hydrological  
125 pathway. Water heights within the flumes were measured by pressure level sensors (OTT  
126 hydromet, Loveland, CO., USA). These sensors measure the depth of water by means of an  
127 integrated controller and ceramic pressure-measuring cell. The level offset (to the flume bed)  
128 was checked fortnightly and updated, if required, in the logger software.

#### 129 2.4 Automated in situ Measurements

##### 130 2.4.1 Phosphorus

131 Total P and TRP are measured in a sample collected from a sump at the monitoring station by  
132 a separate device (SIGMATAX 2, Hach, Salford, UK) which homogenises an unfiltered  
133 sample using ultra-sound before passing it to a process photometer (PHOSPHAX sigma,  
134 Loveland, Colorado, USA). The analyser analyses ortho-phosphate colourimetrically using  
135 standard molybdenum blue chemistry. Total P samples are digested prior to colourimetric  
136 analysis by heating, under pressure, with sulphuric acid and sodium peroxydisulphate while  
137 TRP analysis occurs on an undigested sample. The PHOSPHAX was calibrated daily through  
138 the running of an internal standard; however, it was not possible to run further quality  
139 controls or references. The lowest concentration the instrument is reported to measure is 50  
140 ( $\pm 1$ )  $\mu\text{g PO}_4\text{-P l}^{-1}$ .

##### 141 2.4.2 Nitrogen

142 Drainage the sump in the conduit upstream of the flume is automatically pumped every 15  
143 mins into a purpose built stainless steel by-pass flow cell that houses the sensors. Water is



144 pumped into, and out of, the base of the flow cell and this, coupled with the V shape design,  
145 ensures that there is no retention of sediment or particulate matter either between samples or  
146 over time. Within the flow cell,  $\text{NH}_4\text{-N}$  ( $\text{NH}_4$  + ammonia) is measured using an ion selective  
147 electrode contained within a multi-parameter sonde (6600V2, YSI, Hampshire, UK). Total  
148 oxidised inorganic nitrogen is measured by a self-cleaning, optical UV absorption sensor  
149 (NITRATAX Plus SC, Loveland, Colorado, USA). There is no specified lowest working  
150 concentration for this sonde; however, as it has an accuracy of  $\pm 2 \text{ mg NH}_4\text{-N l}^{-1}$  at its lower  
151 range, anything below this is considered ‘not accurate’ (YSI, Ohio, USA. *pers. comm.*).  
152 Oxidised inorganic nitrogen dissolved in water absorbs UV light at wavelengths below 250  
153 nm, so by passing UV light through the medium and measuring the absorption using a 2-  
154 beam turbidity compensated photometer, the  $\text{NO}_x\text{-N}$  concentration is calculated. The lowest  
155 accurately determinable concentration for the instrument is  $0.503 \pm 0.5 \text{ mg NO}_x\text{-N l}^{-1}$ .  
156 Both probes were calibrated monthly and drift corrected, but no additional in situ quality  
157 controls were applied.

## 158 2.5 Laboratory Measurements

159 Unfiltered samples presented to the laboratory were analysed for both TP and RP thus giving  
160 equivalent data to that generated from the Phosphax instruments (i.e. TP and TRP).  
161 Samples requiring TP analysis were initially subject to an oxidation reaction using acidified  
162 potassium persulphate thus converting all P forms to RP. Both digested and undigested  
163 samples were then analysed for RP colourimetrically on an Aquachem 250 analyser using a  
164 molybdenum blue reaction (Murphy and Riley, 1962). The limits of quantification (LOQ: the  
165 lowest accurately determinable concentration) for TP and RP were  $10 (\pm 1.4)$  and  $2 (\pm 0.04)$   
166  $\mu\text{g PO}_4\text{-P l}^{-1}$ , respectively. The accuracy of TP digestions was checked using quality controls  
167 which were always within 8 % of the target value and with 78 % being within 5 %. Similarly,



168 quality controls were run during the analysis of RP which were always within 5 % of the  
169 target value.

170 Unfiltered samples were also analysed colourimetrically for  $\text{NH}_4\text{-N}$  and  $\text{NOx-N}$  on the  
171 Aquachem 250 analyser. Total oxidised inorganic nitrogen was determined through the  
172 reduction of nitrate to nitrite by hydrazine sulphate and total nitrite is diazotized with  
173 sulphanilamide and coupled with N-1-naphthylethylenediamine dihydrochloride to form an  
174 azo dye with an absorbance maximum at 540 nm. The LOQ for this method was  $0.1 (\pm 0.003)$   
175 mg  $\text{NOx-N l}^{-1}$  and quality controls were always within 3 % of their target.

176 Ammonia/ammonium was determined by the chlorination of ammonia with sodium  
177 dichloroisocyanurate to monochloramine, which reacts with salicylate to form a second  
178 intermediate, 5-aminosalicylate. Oxidative coupling of 5-aminosalicylate with salicylate  
179 forms an indophenol dye with an absorbance maximum at 660 nm. Nitroprusside stabilises  
180 the monochloramine intermediate and also promotes the final oxidative coupling stage. The  
181 LOQ for this method was  $0.4 (\pm 0.01)$  mg  $\text{NH}_4\text{-N l}^{-1}$  and quality controls were always within  
182 5 % of their target.

## 183 2.6 Data Pre-processing

184 For TP and TRP, the manual grab sampling and in situ flume measurements only occurred  
185 simultaneously on five out of 30 occasions for all three sub-catchments, thus only five paired  
186 samples could be compared directly, with the same time stamp. For the nitrogen species,  
187 measurements only occurred simultaneously on one of three occasions. Thus, to make efficient  
188 use of all the grab sampling data, the in situ flume chemistry data were infilled (or predicted)  
189 to provide an exact match to the grab sampling times. This was achieved using a splines fit (via  
190 the `na.spline()` function in the 'zoo' R package of Zeileis and Grothendieck (2005)). Outputs  
191 of prediction uncertainty for the infilled data were not sought, although future work could



192 transfer this uncertainty into the subsequent relationship analyses (e.g. via weighted correlation  
193 or regression analysis). In this respect, all infilled in situ data points are effectively viewed as  
194 measured in situ data for subsequent statistical analyses (this assumption is still checked  
195 visually, however). Constraints were also set in place to ensure the infilling did not provide  
196 values below zero or provide values at a higher level of precision than that measured.

## 197 2.7 Statistical Procedures

198 Once the infilling had been conducted, paired in situ flume and laboratory grab sampling data  
199 were graphically related using time series and scatterplots for all four water quality parameters.  
200 Time series plots are useful in that they can indicate systematic effects, such as sustained  
201 periods of over- or under-estimation, but where the general temporal pattern of the data is  
202 retained. The time series plots also provide an important visual assessment of the spline  
203 infilling procedure described above. For scatterplots, if the two methods of measurement were  
204 an exact match, then they should lie on the 45° line. Data points that lie below the 45° line  
205 indicate where the in situ data under-estimates the laboratory data, and vice-versa.

206 These visual summaries were complemented by a basic set of statistical goodness-of-fit  
207 diagnostics. The intercept and slope parameters from linear regression fits (between the in situ  
208 and laboratory data) are found, together with  $p$ -values for significance from zero and from one,  
209 respectively. Associated  $R^2$  values from the same regressions are also reported and should tend  
210 to 1. Mean error (ME), root mean squared error (RMSE) and Normalised RMSE (NRMSE)  
211 values are reported (via functions in the 'hydroGOF' R package), where all three diagnostics  
212 should tend to zero. In this case, the *errors* referred to *in situ minus laboratory* data, thus a  
213 negative ME value would indicate that the in situ data under-estimates the laboratory data, on  
214 average. RMSE reflects the variance of the errors, which ideally needs to be as small as  
215 possible. The NRMSE diagnostic is a relative measure of RMSE, and thus relays quite clearly



216 when the in situ data has a good or poor correspondence with the laboratory data, regardless of  
217 different scales of measurement from the different sub-catchments.

218 A final, but limited analysis was also conducted on the genuine paired samples found for TP  
219 and TRP only - i.e. only five pairs for each sub-catchment. This data was analysed using paired  
220 *t*-tests and analysis of variance (ANOVA), and was presented using Tukey mean-difference  
221 plots. Further analyses could have considered random sampling for five pairs from the infilled  
222 data of 30 pairs, and repeating the tests considered here, on each random sample. This would  
223 assess the sensitivity of the results to sample variation and to an extent, the infilling. However,  
224 this was considered beyond the scope of this study; and in any case, the outcomes would always  
225 be severely limited due to the very small sampling size.

226 All statistical analyses were conducted in R (<https://www.r-project.org/>), where in all cases,  
227 the in situ data were compared to the *unfiltered* laboratory data.

228

### 229 3. Results

#### 230 3.1 Data summaries

231 In the first instance, it is useful to summarise the measured data, where infilled data or paired  
232 data are not needed. In this respect, sample size and the ranges (minimums to maximums) for  
233 TP, TRP, NO<sub>x</sub>-N and NH<sub>4</sub>-N measured in the drainage from sub-catchments 2, 5 and 8,  
234 obtained by both the automated in situ analysers and laboratory analysed manually collected  
235 samples are presented in Table 1. Values of TP ranged between 40 to 770 µg P l<sup>-1</sup> and for TRP  
236 between 0 to 70 µg P l<sup>-1</sup> as measured by the in situ Phosphax analysers. For NO<sub>x</sub>-N and NH<sub>4</sub>-  
237 N, the values measured in situ ranged between 0.62 to 4.8 mg N l<sup>-1</sup> and 0.04 to 1.5 mg N l<sup>-1</sup>,  
238 respectively. The range of values measured in the manually collected samples analysed in the  
239 laboratory, in general, compare favourably to the in situ data. This is even though there are



240 much fewer data and that potential highs and lows in concentration could have been missed in  
241 the manually collected samples.

### 242 3.2 Chemistry Response to Discharge

243 Data on the discharge from the three sub-catchments and rainfall is presented in Figure 2. The  
244 results show three similar twin peaked hydrographs but with different magnitudes of peak  
245 discharge of 15, 22 and 28 l s<sup>-1</sup> for sub-catchments 2, 5 and 8, respectively. The different scales  
246 of the hydrographs reflect differences in, amongst other things, sub-catchment size, rainfall,  
247 slope, soil moisture and soil type. In all three sub-catchments, TP data from both in situ  
248 analysers and the laboratory analysed grab samples exhibited a positive relationship with  
249 discharge (Figure 3a-c). The highest values of TP were associated with the initial smaller peak  
250 in discharge, and a latter smaller peak in TP associated with the second, large, peak in  
251 discharge. In all cases, the chemographs generated by both analytical approaches appear similar  
252 and match the responses reported elsewhere (Heathwaite and Dils, 2000; Granger et al., 2010;  
253 Lloyd et al., 2016). Such relationships with discharge are less clear with the lower  
254 concentration TRP data (Figure 3d-f). In situ TRP concentration data again exhibit a positive  
255 relationship with discharge, and possibly even a two peaked chemograph, similar to that of the  
256 TP data. However, the low concentration range compared to that of the TP, means that when  
257 the data is rounded to the nearest 10 µg P l<sup>-1</sup>, the resolution of the chemograph is severely  
258 affected and detail is lost. The TRP data generated via laboratory analysis are not subject to  
259 this rounding effect; however, these data exhibit considerably more ‘noise’, and while it is  
260 possible to visualise some relationships with discharge, in all but the data from sub-catchment  
261 8, this is highly subjective.

262 The NO<sub>x</sub>-N chemographs generated by the in situ analysers and the laboratory analysed  
263 samples display the classic dilution effect reported elsewhere (Webb and Walling, 1985;  
264 Granger et al., 2010; Lloyd et al., 2016) with concentrations dropping rapidly with the onset of



265 increased discharge, and slowly recovering to pre-storm flow values over time on the falling  
266 limb of the hydrograph (Figure 4e-f). The data generated for  $\text{NH}_4\text{-N}$  from the in situ sensors  
267 clearly show a positive relationship with discharge from all sub-catchments and, interestingly,  
268 even a second  $\text{NH}_4\text{-N}$  peak on the chemograph of sub-catchment 8 associated with the main  
269 spike in discharge (Figure 4a-c). This positive relationship is not unusual (House and Warwick,  
270 1998a; Inamdar, 2007; Fucik et al., 2012), although it tends to be much lower in concentration  
271 compared to  $\text{NO}_x\text{-N}$  and often this is not very discernible as the  $\text{NH}_4\text{-N}$  is rapidly nitrified to  
272  $\text{NO}_x\text{-N}$  (House and Warwick, 1998b). Where high concentrations of  $\text{NH}_4\text{-N}$  occur as spikes  
273 associated with discharge, it is often more related to incidental losses of recently applied  $\text{NH}_4\text{-N}$   
274 N as a result of farmland management practices (Granger et al., 2010). Data generated from  
275 the laboratory analysed grab samples provide a slightly more mixed picture. Where  
276 concentrations were highest (sub-catchment 8), these data appear to confirm the positive  
277 relationship of  $\text{NH}_4\text{-N}$  with discharge, even reproducing the second  $\text{NH}_4\text{-N}$  peak. In sub-  
278 catchment 5, where  $\text{NH}_4\text{-N}$  concentrations were lowest, the laboratory data are noisier, but it  
279 is still possible to observe an increase in  $\text{NH}_4\text{-N}$  concentration with increased discharge. In sub-  
280 catchment 2, however, the laboratory  $\text{NH}_4\text{-N}$  data show no relationship with discharge (Figure  
281 4a).

282 In all chemographs (Figures 3-4), the outcomes of the in situ spline infilling described above,  
283 is shown. Here in filling never required a difficult extrapolation, but instead was always a  
284 simple interpolation that was richly informed by actual measured data that were temporally  
285 similar. Clearly, no unusual predictions result and the infilling should be considered reliable,  
286 and can be safely viewed as strongly comparable to the in situ data for subsequent statistical  
287 analyses.

288

289 3.3 Comparison of In Situ and Laboratory Analysis



290 The data obtained for genuine paired laboratory analysed manual grab samples and  
291 PHOSPHAX in situ TP and TRP are presented in Table 4 (it is of no value to do this for nitrogen  
292 species, as only one to three genuine pairs were available). The differences between the two  
293 sets of data are reported relative to the laboratory data which have been subject to full analytical  
294 quality control. Using this comparison, it can be seen that differences between TP values are  
295 lower than for TRP, with respective ranges between +56 to -30  $\mu\text{g P l}^{-1}$  (+29 % to -38 %) and  
296 +13 to -33  $\mu\text{g P l}^{-1}$  (+186 % to -57 %).

297 The difference between the two methods of measurement were assessed using paired *t*-tests.  
298 The average difference between laboratory and in situ values for TP was -3.933 (standard error  
299 of difference 4.947, 95 % CI -14.54, 6.677) and the standard deviation of differences was 19.16.  
300 The *t*-test for TP indicated that there was no evidence of a difference between laboratory and  
301 in situ measurements ( $t_{14} = -0.8$ ,  $p = 0.44$ ). However, the average difference between lab and  
302 in situ values for TRP was 8.933 (standard error of difference 3.534, 95 % CI 1.353, 16.51)  
303 with the standard deviation of differences being 13.69. This indicated that for TRP, that there  
304 was evidence of a statistically significant difference between laboratory and in situ  
305 measurements ( $t_{14} = 2.53$ ,  $p = 0.024$ ).

306 Differences between the two measurement methods was also assessed using ANOVA in order  
307 to take into account that the data came from three different sub-catchments. This did not suggest  
308 any influence of the sub-catchment difference on the size of measurement difference. Tukey  
309 mean-difference plots are presented in Figure 5 and plot the difference between the two values  
310 against the average of the two measured values. Limits of agreement (dashed lines) are plotted  
311 at  $\pm 2$  standard deviations from the mean difference and indicate the range that approximately  
312 95 % of the data is expected to fall in. From these plots the data suggest that, while there is no  
313 obvious trend in TRP data, differences in the TP values are greater at lower concentrations with



314 the laboratory generating higher values but that this difference reduces as the TP concentration  
315 increases.

### 316 3.4 Comparison of Modelled In Situ and Laboratory Analysis

317 Given the small sample number of actual in situ and laboratory analysed grab samples,  
318 assessing the differences between these two approaches is extremely limited. We therefore  
319 compare the modelled in situ and laboratory analysed samples. This is because we consider  
320 error in the data obtained from the laboratory to be relatively low (Madrid and Zayas, 2007),  
321 with these data subject to analytical quality controls and checks. Any difference between in  
322 situ values and the laboratory must therefore be explained via other processes and mechanisms.

#### 323 3.4.1 Phosphorus

324 For TP and TRP, the resultant paired data is presented using scatterplots in Figure 6. In all  
325 cases, the ideal 45° line is shown together with the actual linear fit. Results of the tests for  
326 whether or not the ideal and actual lines significantly deviate from each other are given in Table  
327 2, together with a general fit measure in  $R^2$ . At the 95 % level of significance, only the  
328 laboratory and in situ data for TP in sub-catchments 5 and 8 are in good agreement (as the  $p$ -  
329 values in Table 2 indicate the intercepts and slopes of their fitted lines are not significantly  
330 different to zero or one, respectively). Laboratory and in situ TP data in all three sub-  
331 catchments do however provide relatively high  $R^2$  values, where for sub-catchment 2, the in  
332 situ TP tends to under-estimate laboratory TP at high values, pivoting the fitted line downwards  
333 at these values. Table 3 provides the ME, RMSE and NRMSE results for TP and TRP, where  
334 the negative ME value for TP in sub-catchment 2, indicates an overall under-estimation of  
335 laboratory TP by in situ TP, whilst in the other two sub-catchments, the reverse is true.  
336 Although sub-catchment 2 does not indicate the strongest 1:1 relationship between the paired



337 TP data, its TP data are most alike in terms of variation - as seen by the least scatter around the  
338 fitted line, coupled with the lowest NRMSE value.

339 Corresponding results for TRP are not promising (Figure 5, Tables 2 and 3), where each  
340 scatterplot depicts a poor correspondence between the laboratory and in situ TRP data, and  
341 these poor relationships are statistically endorsed by the test results and the low  $R^2$  values  
342 presented in Table 2. Diagnostics presented in Table 3, provide little further insight into the  
343 behaviour of the paired TRP data, except that in situ TRP will tend to under-estimate laboratory  
344 TRP (as MEs are negative in two sub-catchments). Note however, that in situ TRP tends to be  
345 less variable than laboratory TP, as shown by the scatterplots.

#### 346 3.4.2 Nitrogen

347 Results for the differences between the in situ and laboratory  $\text{NH}_4\text{-N}$  data are quite complex.  
348 From the scatterplots in Figure 7, and the associated tests in Table 2, a 1:1 relationship between  
349 the paired  $\text{NH}_4\text{-N}$  data in sub-catchments 2 and 8 is clearly absent, although within sub-  
350 catchment 2 the data are influenced by an anomalously high laboratory  $\text{NH}_4\text{-N}$  result. However,  
351 the paired  $\text{NH}_4\text{-N}$  data do provide a high  $R^2$  value in catchment 8, indicating a reproducible  
352 relationship of sorts, albeit not one that is ideal. The most promising relationship for the paired  
353  $\text{NH}_4\text{-N}$  data is found in sub-catchment 5, where the  $R^2$  value is reasonable and the NRMSE  
354 value is much lower than that found in the other two sub-catchments.

355 Results for the differences between the paired  $\text{NO}_x\text{-N}$  data, in contrast, are quite promising.  
356 The scatterplots in Figure 7, overall, show a reasonable correspondence between the paired  
357  $\text{NO}_x\text{-N}$  data, for all three sub-catchments, which is endorsed by very high  $R^2$  values in Table  
358 2. Although in all cases, these relationships cannot be viewed as 1:1 as indicated by the test  
359 results in Table 2. For all cases, the in situ  $\text{NO}_x\text{-N}$  data tends to over-predict the laboratory  
360  $\text{NO}_x\text{-N}$  data.



361

362 4. Discussion

363 4.1 Phosphorus

364 Direct comparison of laboratory and the in situ TP data shows no evidence of a significant  
365 difference although, at lower concentrations, the in situ data would appear to be *lower* than the  
366 laboratory values. Here, however, it is important to bear in mind that no direct comparisons  
367 were made on samples that were taken on discharge at the higher end of the concentration  
368 range. The modelled data confirm that there is a good match, in general, between in situ data  
369 and laboratory values with fitted lines not being significantly different to zero or one in sub-  
370 catchments 5 and 8. In sub-catchment 2, conversely, it would appear that in situ data were  
371 lower than laboratory values at *higher* concentrations which is confirmed by the negative ME  
372 value for this sub-catchment. Irrespective of this, all data showed good correlations with  
373 relatively high  $R^2$  values, a fact that is confirmed by the good agreement shown by the  
374 chemographs in Figure 3 a-c. The data suggest that for TP, the PHOSPHAX in situ analysis  
375 provides reasonably good agreement with manual sampled laboratory analysed samples, and  
376 conversely that the manual samples do not suffer excessively from systematic, sampling or  
377 storage errors. However, it is noteworthy that the PHOSPHAX in situ data does produce a  
378 relatively 'smooth' chemograph which is in contrast to the laboratory data which is noticeably  
379 more 'noisy' and even contains a few anomalously high concentrations ('outliers') i.e. Figure  
380 3a. This is probably a result of one or a combination of, three important issues regarding TP:  
381 a) data generated in situ is 'smoothed' by the analyser by rounding values to the nearest  $10 \mu\text{g}$   
382  $\text{P l}^{-1}$ , b) sample container contamination at either the sampling stage or latterly during  
383 laboratory digestion, and c) laboratory analytical error. In the first case, the in situ values might  
384 actually be noisy, but this is not reflected in the smoothed data generated for download. In the  
385 second case, it is assumed that P of unknown origin (biological, tap water, chemical) could



386 have randomly, as opposed to systematically, contaminated some equipment leading to a high  
387 result. In this scenario it is hard to imagine how this sort of error could cause a lower result  
388 than expected. In the third, it could just be a random analytical artefact, which has resulted in  
389 an unusually high (but could also cause an unusually low) result.

390 In contrast, the comparison of the TRP data were far less conclusive. Direct sample pair  
391 comparison indicated a significantly lower concentration measured in situ than that measured  
392 in the laboratory. Further, the larger data set generated by comparing modelled in situ and  
393 manually sampled laboratory analysed TRP data shows very poor correspondence with  
394 significant differences in both slope and intercept being  $>0$  and  $<1$  in every case, respectively,  
395 indicating that the in situ data were consistently lower than that measured in the laboratory.  
396 The low  $R^2$  further confirms poor replication of data, a fact further confirmed by the  
397 chemographs presented in Figure 3 d-f. It can be seen from Figure 6 d-f that the main cause for  
398 poor correlation between the two data-sets is probably down to a combination of two factors;  
399 a) the low resolution of the PHOSPHAX in situ data, rounding all numbers to the nearest 10,  
400 and b) more importantly, that the vast majority of the PHOSPHAX in situ data is lower than  
401 the machine's analytical limit of  $50 \mu\text{g P l}^{-1}$ . That being said, in situ TRP concentrations have  
402 trends (Figure 3 d-f) which are not so clearly represented in laboratory TRP data which again  
403 although being above LOQ are extremely noisy.

404 One explanation is sample degradation between sampling and analysis. Ideally, samples should  
405 be analysed immediately after collection to minimise degradation effects, but sample storage  
406 is usually unavoidable prior to analysis. The concentrations of dissolved nutrient within water  
407 samples can vary during storage as the result of a wide range of physical, biological and  
408 chemical processes including sorption, hydrolysis, precipitation, complexation, and microbial  
409 uptake and release (Jarvie et al., 2002b). This is particularly relevant for the samples collected  
410 in this instance since they were unfiltered prior to analysis such that they were of the same



411 matrix as the sample collected by the PHOSPHAX. Rapid filtration of samples (typically >0.45  
412  $\mu\text{m}$ ) is usually recommended in order to exclude microbial cells and inorganic particulate  
413 material, which can result in changes in physical or chemical forms of P through processes  
414 such as microbial uptake or adsorption in unfiltered samples (Lambert et al., 1992; Jarvie et  
415 al., 2002b; Worsfold et al., 2005). Biological processes and sorption to particulate matter can  
416 be rapid; Lambert et al. (1992) reported that concentrations of ‘dissolved’ P decreased  
417 substantially over a four-hour period in unfiltered lake water samples. Another possible effect  
418 of the unfiltered matrix is that sample particulates could be causing noise in the laboratory TRP  
419 analysis, both through their physical presence in the flow cell and through biogeochemical  
420 alteration of the sample in reaction to analytical reagents (Jarvie et al., 2002b), although  
421 presumably this is also an effect that happens in the PHOSPHAX analyser.

#### 422 4.2 Nitrogen

423 Results for the differences between the in situ and laboratory  $\text{NH}_4\text{-N}$  data are quite complex  
424 but the 1:1 relationship between the paired  $\text{NH}_4\text{-N}$  data were in general poor. However, the  
425 paired  $\text{NH}_4\text{-N}$  data do provide reasonable  $R^2$  values in sub-catchments 5 and 8. This variation  
426 in the responses needs to be examined more carefully. From the chemographs in Figure 4 a-c,  
427 all three in situ sondes produced similar responses, with rising and falling concentrations  
428 matching rises and falls in discharge. This would seem to indicate that the sondes were  
429 detecting a genuine chemical response. However, all the in situ data reported are well below  
430 the accuracy of the sonde at this concentration (Figure 7 a-c). This could be as a result of other  
431 factors affecting the sonde other than  $\text{NH}_4\text{-N}$ . The ion selective electrode is subject to effects  
432 caused by changes in temperature and interference from ions, which are similar in nature to the  
433 analyte. While the changes in temperature, or ions such as sodium and chloride, might only be  
434 slight in response to field drainage, they could be enough to cause the small responses recorded  
435 here which have a maximum range of about  $1 \text{ mg } \text{NH}_4\text{-N } \text{l}^{-1}$  and which were always below the



436 recommended accurate working concentration for the instrument. That said, in sub-catchment  
437 8 which had the highest recorded  $\text{NH}_4\text{-N}$  values both by the sonde and the laboratory, the  
438 chemical response in  $\text{NH}_4\text{-N}$  is mirrored by the laboratory grab samples, giving the highest  $R^2$   
439 of 0.83. Laboratory concentration values were also in the main below LOQ, but were much  
440 closer to analytical limits than that of the sonde, and at their peak, slightly exceeded it. This  
441 would seem to indicate that the sonde response in sub-catchment 8 would appear to be genuine  
442 even if the absolute  $\text{NH}_4\text{-N}$  concentration is suspect. If this is the case, then we can maybe  
443 assume that the responses recorded in sub-catchments 2 and 5 are also genuine, even if their  
444 absolute values may not be. Laboratory values from these two sub-catchments were, however,  
445 well below LOQ so cannot be used to back up this conclusion. Interestingly, laboratory data  
446 from sub-catchment 5 (which recorded the lowest  $\text{NH}_4\text{-N}$  concentration from any of the three  
447 sub-catchments) does mirror the response of the sonde to a degree ( $R^2 = 0.67$ ), while the values  
448 from sub-catchment 2 show no similarity at all. This paradox is confusing as if the loss of  
449 response from sub-catchment 2 was due to sampling and unfiltered storage losses because of  
450 the low  $\text{NH}_4\text{-N}$  concentration (i.e. (Kotlash and Chessman, 1998; Lentz, 2013)) then that would  
451 surely have occurred in the even lower concentrations of sub-catchment 5. Further storm period  
452 analyses are required to help resolve this paradox.

453 The  $\text{NO}_x\text{-N}$  data, in contrast, show good similarity, which although not significantly similar,  
454 have a very high  $R^2$  (Table 2) and in all cases, the in situ modelled  $\text{NO}_x\text{-N}$  data tends to over-  
455 predict the laboratory data. The reason for this is clear from the chemographs in Figure 4 d-f,  
456 whereby the trends in both sets of data are virtually identical (leading to high  $R^2$ ), but whereby  
457 modelled in situ values and laboratory values differ at lower concentrations. In all three  
458 examples, pre- and post-discharge event values are near identical, but with the onset of  
459 increased discharge, the  $\text{NO}_x\text{-N}$  concentrations drop, with laboratory values dropping further  
460 than modelled in situ values. In all cases, none of the recorded values are below the instruments



461 working capabilities so should be considered reliable, and only a few values are below the  
462 laboratory LOQ. The reason for this discrepancy is unclear and is the result, or a combination  
463 of, either the sensor underestimating NO<sub>x</sub>-N at increased discharge, or the laboratory analysed  
464 grab samples having a lower measured NO<sub>x</sub>-N at high discharge.

465 Here it useful to bear in mind that as the measurement is based on the evaluation of (invisible)  
466 UV light, the colour of the medium has no effect. The sensor contains a two-beam absorption  
467 photometer with turbidity compensation. So perhaps this turbidity compensation is having a  
468 greater effect on reducing calculated NO<sub>x</sub>-N values in situ at times when turbidity is greatly  
469 increased (at high discharge).

470

## 471 5. Conclusions

472 An increasing number of studies are reporting the use of in situ analysers and sensors to collect  
473 high temporal resolution hydrochemical data. Whilst such data permit the use of exploratory  
474 data interpretation techniques such as hysteretic loops, much hydrochemical data are used to  
475 estimate time-variant or averaged concentrations in the context of environmental objectives or  
476 thresholds and to estimate nutrient loads. The comparison herein of nutrient species data  
477 collected using in situ analysers or sondes and manually collected laboratory analysed samples  
478 confirms the following:

- 479 • PHOSPHAX in situ TP data would appear to be reliable, most likely as the  
480 determined concentrations are nearly always more than the instrument's lower  
481 limits. Discrepancies between laboratory and in situ data appear to increase as the  
482 PHOSPHAX lower measurable limit is approached.
- 483 • PHOSPHAX TRP measurements, in the context of the field drainage described  
484 here, are unreliable as the concentrations were nearly always below the LOQ for



485 the instrument. This is reflected in the poor agreement between laboratory and  
486 instrument data. This poor agreement is largely due to the laboratory data being  
487 very ‘noisy’ despite being above laboratory LOQ and may reflect  
488 sampling/storage issues related to unfiltered sample matrix. Despite this, trends in  
489 the concentration were discernible using the in situ data, although validation of  
490 these trends requires more field work.

491 • The  $\text{NH}_4\text{-N}$  laboratory analysed data showed that concentrations were nearly  
492 always below LOQ for the laboratory and as such were well below measurable  
493 limits for the YSI sonde and electrode. This suggests that this analytical system is  
494 not appropriate for this type of environmental setting. Despite this trends in  $\text{NH}_4\text{-N}$   
495 concentration were discernible from the sonde, although whether these are  
496 analytical artefacts or genuine remains uncertain.

497 • Concentration of  $\text{NO}_x\text{-N}$  were always higher than LOQ for both the in situ  
498 NITRATAX sonde and the laboratory analysis. The two set of data show good  
499 agreement, and exhibit similar classical  $\text{NO}_x\text{-N}$  chemographs. However,  
500 differences in the  $\text{NO}_x\text{-N}$  are not linear and appear at lower concentration/higher  
501 Q, with the in situ data giving lower concentrations than the laboratory measured  
502 values. This may be an effect of the NITRATAX considering turbidity interference  
503 at higher Q.

504

## 505 6. Summary

506 PHOSPHAX TP and NITRATAX  $\text{NO}_x\text{-N}$  data show good agreement with laboratory data  
507 in this environmental setting. However, PHOSPHAX TRP and YSI  $\text{NH}_4\text{-N}$  data were less  
508 reliable as concentrations were below the instrumental limits. Both these instruments



509 generated data with repeatable trends in concentration, but trends that were not reflected in  
510 the laboratory data which, in turn, was noisier. It is unclear whether the instrument trends  
511 were genuine, or why they were not present in the laboratory data which is itself very  
512 variable.

513

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#### 521 8. References

- 522 Bende-Michl, U., and Hairsine, P. B.: A systematic approach to choosing an automated  
523 nutrient analyser for river monitoring, *J. Environ. Monit.*, 12, 127-134, 2010.
- 524 Bende-Michl, U., Verburg, K., and Cresswell, H. P.: High-frequency nutrient monitoring to  
525 infer seasonal patterns in catchment source availability, mobilisation and delivery, *Environ.*  
526 *Monit. Assess.*, 185, 9191-9219, 2013.
- 527 Bieroza, M. Z., and Heathwaite, A. L.: Seasonal variation in phosphorus concentration-  
528 discharge hysteresis inferred from high-frequency in situ monitoring, *J. Hydrol.*, 524, 333-  
529 347, 2015.
- 530 Campbell, J. L., Rustad, L. E., Porter, J. H., Taylor, J. R., Dereszynski, E. W., Shanley, J. B.,  
531 Gries, C., Henshaw, D. L., Martin, M. E., Sheldon, W. M., and Boose, E. R.: Quantity is  
532 Nothing without Quality: Automated QA/QC for Streaming Environmental Sensor Data,  
533 *Bioscience*, 63, 574-585, 2013.
- 534 Carey, R. O., Wollheim, W. M., Mulukutla, G. K., and Mineau, M. M.: Characterizing  
535 Storm-Event Nitrate Fluxes in a Fifth Order Suburbanizing Watershed Using In Situ Sensors,  
536 *Environ. Sci. Technol.*, 48, 7756-7765, 2014.



- 537 Donn, M. J., Barron, O. V., and Barr, A. D.: Identification of phosphorus export from low-  
538 runoff yielding areas using combined application of high frequency water quality data and  
539 MODHMS modelling, *Sci. Total Environ.*, 426, 264-271, 2012.
- 540 Fucik, P., Kaplicka, M., Kvittek, T., and Peterkova, J.: Dynamics of Stream Water Quality  
541 during Snowmelt and Rainfall - Runoff Events in a Small Agricultural Catchment, *Clean-Soil*  
542 *Air Water*, 40, 154-163, 2012.
- 543 Granger, S. J., Hawkins, J. M. B., Bol, R., White, S. M., Naden, P., Old, G., Bilotta, G. S.,  
544 Brazier, R. E., Macleod, C. J. A., and Haygarth, P. M.: High temporal resolution monitoring  
545 of multiple pollutant responses in drainage from an intensively managed grassland catchment  
546 caused by a summer storm, *Water Air Soil Pollut.*, 205, 377-393, 2010.
- 547 Harmel, R. D., Cooper, R. J., Slade, R. M., Haney, R. L., and Arnold, J. G.: Cumulative  
548 uncertainty in measured streamflow and water quality data for small watersheds, *Trans.*  
549 *ASABE*, 49, 689-701, 2006.
- 550 Harris, G., and Heathwaite, A. L.: Inadmissible evidence: knowledge and prediction in land  
551 and riverscapes, *J. Hydrol.*, 304, 3-19, 2005.
- 552 Heathwaite, A. L., and Dils, R. M.: Characterising phosphorus loss in surface and subsurface  
553 hydrological pathways, *Sci. Total Environ.*, 251, 523-538, 2000.
- 554 Heffernan, J. B., and Cohen, M. J.: Direct and indirect coupling of primary production and  
555 diel nitrate dynamics in a subtropical spring-fed river, *Limnol. Oceanogr.*, 55, 677-688, 2010.
- 556 House, W. A., and Warwick, M. S.: Hysteresis of the solute concentration/discharge  
557 relationship in rivers during storms, *Water Res.*, 32, 2279-2290, 1998a.
- 558 House, W. A., and Warwick, M. S.: Intensive measurements of nutrient dynamics in the  
559 River Swale, *Sci. Total Environ.*, 210, 111-137, 1998b.
- 560 Howden, N. J. K., Burt, T. P., Worrall, F., Whelan, M. J., and Bierozza, M.: Nitrate  
561 concentrations and fluxes in the River Thames over 140 years (1868-2008): are increases  
562 irreversible?, *Hydrol. Process.*, 24, 2657-2662, 2010.
- 563 Inamdar, S.: Exports of dissolved ammonium (NH<sub>4</sub><sup>+</sup>) during storm events across multiple  
564 catchments in a glaciated forested watershed, *Environ. Monit. Assess.*, 133, 347-363, 2007.
- 565 Jarvie, H. P., Neal, C., Williams, R. J., Neal, M., Wickham, H. D., Hill, L. K., Wade, A. J.,  
566 Warwick, A., and White, J.: Phosphorus sources, speciation and dynamics in the lowland  
567 eutrophic River Kennet, UK, *Sci. Total Environ.*, 282, 175-203, 2002a.
- 568 Jarvie, H. P., Withers, P. J. A., and Neal, C.: Review of robust measurement of phosphorus in  
569 river water: sampling, storage, fractionation and sensitivity, *Hydrology and Earth System*  
570 *Sciences*, 6, 113-131, 2002b.



- 571 Jordan, P., Arnscheidt, J., McGrogan, H., and McCormick, S.: High-resolution phosphorus  
572 transfers at the catchment scale: the hidden importance of non-storm transfers, *Hydrology  
573 and Earth System Sciences*, 9, 685-691, 2005.
- 574 Kirchner, J. W., and Neal, C.: Universal fractal scaling in stream chemistry and its  
575 implications for solute transport and water quality trend detection, *Proc. Natl. Acad. Sci. U.  
576 S. A.*, 110, 12213-12218, 2013.
- 577 Kotlash, A. R., and Chessman, B. C.: Effects of water sample preservation and storage on  
578 nitrogen and phosphorus determinations: Implications for the use of automated sampling  
579 equipment, *Water Res.*, 32, 3731-3737, 1998.
- 580 Lambert, D., Maher, W., and Hogg, I.: Changes in phosphorus fractions during storage of  
581 lake water, *Water Res.*, 26, 645-648, 1992.
- 582 Lentz, R. D.: Delayed Sample Filtration and Storage Effects on Dissolved Nutrients  
583 Measured in Agricultural Runoff, *Commun. Soil Sci. Plant Anal.*, 44, 2952-2960, 2013.
- 584 Lloyd, C. E. M., Freer, J. E., Johnes, P. J., and Collins, A. L.: Using hysteresis analysis of  
585 high-resolution water quality monitoring data, including uncertainty, to infer controls on  
586 nutrient and sediment transfer in catchments, *Sci. Total Environ.*, 543, 388-404, 2016.
- 587 Madrid, Y., and Zayas, Z. P.: Water sampling: Traditional methods and new approaches in  
588 water sampling strategy, *Trac-Trends Anal. Chem.*, 26, 293-299, 2007.
- 589 McMillan, H., Krueger, T., and Freer, J.: Benchmarking observational uncertainties for  
590 hydrology: rainfall, river discharge and water quality, *Hydrol. Process.*, 26, 4078-4111, 2012.
- 591 Mellander, P. E., Jordan, P., Shore, M., Melland, A. R., and Shortle, G.: Flow paths and  
592 phosphorus transfer pathways in two agricultural streams with contrasting flow controls,  
593 *Hydrol. Process.*, 29, 3504-3518, 2015.
- 594 Mellander, P. E., Jordan, P., Shore, M., McDonald, N. T., Wall, D. P., Shortle, G., and Daly,  
595 K.: Identifying contrasting influences and surface water signals for specific groundwater  
596 phosphorus vulnerability, *Sci. Total Environ.*, 541, 292-302, 2016.
- 597 Milne, A. E., Macleod, C. J. A., Haygarth, P. M., Hawkins, J. M. B., and Lark, R. M.: The  
598 wavelet packet transform: A technique for investigating temporal variation of river water  
599 solutes, *J. Hydrol.*, 379, 1-19, 2009.
- 600 Murphy, J., and Riley, J. P.: A modified single solution method for determination of  
601 phosphate in natural waters, *Anal. Chim. Acta*, 26, 31-&, 1962.
- 602 Orr, R. J., Murray, P. J., Eyles, C. J., Blackwell, M. S. A., Cardenas, L. M., Collins, A. L.,  
603 Dungait, J. A. J., Goulding, K. W. T., Griffith, B. A., Gurr, S. J., Harris, P., Hawkins, J. M.  
604 B., Misselbrook, T. H., Rawlings, C., Shepherd, A., Sint, H., Takahashi, T., Tozer, K. N.,  
605 Whitmore, A. P., Wu, L., and Lee, M. R. F.: The North Wyke Farm Platform: effect of



- 606 temperate grassland farming systems on soil moisture contents, runoff and associated water  
607 quality dynamics, *Eur. J. Soil Sci.*, 67, 374-385, 2016.
- 608 Palmer-Felgate, E. J., Jarvie, H. P., Williams, R. J., Mortimer, R. J. G., Loewenthal, M., and  
609 Neal, C.: Phosphorus dynamics and productivity in a sewage-impacted lowland chalk stream,  
610 *J. Hydrol.*, 351, 87-97, 2008.
- 611 Perks, M. T., Owen, G. J., Benskin, C. M. H., Jonczyk, J., Deasy, C., Burke, S., Reaney, S.  
612 M., and Haygarth, P. M.: Dominant mechanisms for the delivery of fine sediment and  
613 phosphorus to fluvial networks draining grassland dominated headwater catchments, *Sci.*  
614 *Total Environ.*, 523, 178-190, 2015.
- 615 Rode, M., Wade, A. J., Cohen, M. J., Hensley, R. T., Bowes, M. J., Kirchner, J. W.,  
616 Arhonditsis, G. B., Jordan, P., Kronvang, B., Halliday, S. J., Skeffington, R. A., Rozemeijer,  
617 J. C., Aubert, A. H., Rinke, K., and Jomaa, S.: Sensors in the Stream: The High-Frequency  
618 Wave of the Present, *Environ. Sci. Technol.*, 50, 10297-10307, 2016.
- 619 Skeffington, R. A., Halliday, S. J., Wade, A. J., Bowes, M. J., and Loewenthal, M.: Using  
620 high-frequency water quality data to assess sampling strategies for the EU Water Framework  
621 Directive, *Hydrology and Earth System Sciences*, 19, 2491-2504, 2015.
- 622 Webb, B. W., and Walling, D. E.: Nitrate behaviour in streamflow from a grassland  
623 catchment in Devon, U.K., *Water Res.*, 19, 1005-1016, 1985.
- 624 Worsfold, P. J., Gimbert, L. J., Mankasingh, U., Omaka, O. N., Hanrahan, G., Gardolinski,  
625 P., Haygarth, P. M., Turner, B. L., Keith-Roach, M. J., and McKelvie, I. D.: Sampling,  
626 sample treatment and quality assurance issues for the determination of phosphorus species in  
627 natural waters and soils, *Talanta*, 66, 273-293, 2005.
- 628 Zeileis, A., and Grothendieck, G.: zoo: S3 infrastructure for regular and irregular time series,  
629 *Journal of Statistical Software*, 14, 27, 2005.
- 630
- 631



632

633

		Sub-catchment 2		Sub-catchment 5		Sub-catchment 8	
		n	range	n	range	n	range
TP $\mu\text{g P l}^{-1}$	In situ	64	40 - 300	62	50 - 380	44	50 - 770
	Lab	30	46 - 365	31	68 - 428	38	48 - 706
TRP $\mu\text{g P l}^{-1}$	In situ	65	20 - 70	63	0 - 60	45	0 - 60
	Lab	30	8 - 77	31	7 - 111	38	6 - 76
NO <sub>x</sub> -N $\text{mg N l}^{-1}$	In situ	130	0.62 - 1.7	126	1.6 - 4.8	119	0.72 - 2.1
	Lab	30	0.11 - 1.7	31	0.66 - 5.1	38	0.24 - 2.1
NH <sub>4</sub> -N $\text{mg N l}^{-1}$	In situ	130	0.12 - 0.32	126	0.04 - 0.14	119	0.55 - 1.5
	Lab	30	0 - 0.19	31	0.01 - 0.13	38	0.09 - 0.48

634

635 **Table 1.** Summary of values (min – max) measured in drainage from the three NWFP sub-  
636 catchments using both the in situ automated analysers and laboratory analysis of manually  
637 collected samples.

638

639

	Catchment	Intercept	<i>p</i> -value	Slope	<i>p</i> -value	<i>R</i> <sup>2</sup>
TP	2	22.40	0.033	0.79	0.002	0.85
	5	-2.34	<b>0.881</b>	1.09	<b>0.293</b>	0.84
	8	20.45	<b>0.259</b>	1.04	<b>0.482</b>	0.86
TRP	2	23.48	0.003	0.46	0.006	0.18
	5	22.48	0.001	0.22	0.000	0.09
	8	19.08	0.003	0.41	0.000	0.21
NH <sub>4</sub> -N	2	0.24	0.000	-0.15	0.003	0.01
	5	0.05	0.000	0.67	0.001	0.67
	8	0.46	0.000	2.17	0.000	0.83
NO <sub>x</sub> -N	2	0.63	0.000	0.54	0.000	0.98
	5	1.20	0.000	0.70	0.000	0.96
	8	0.80	0.000	0.60	0.000	0.98

640

641 **Table 2:** Summary of linear regression outputs for in situ versus laboratory data. The  
642 *p*-values that are bolded indicate intercepts or slopes that are not significantly different  
643 to zero or one, respectively, at the 95% level.

644

645



646

Parameter	Sub-catchment	ME	RMSE	NRMSE (%)
TP	2	-8.07	32.13	39.4
	5	12.05	43.47	50.1
	8	30.98	67.62	42.2
TRP	2	3.27	18.19	110.6
	5	-12.68	26.13	117.1
	8	-4.83	20.55	101.1
NH <sub>4</sub> -N	2	0.20	0.21	638.7
	5	0.03	0.04	88.5
	8	0.82	0.84	626.6
NO <sub>x</sub> -N	2	0.30	0.40	68.3
	5	0.47	0.72	43.8
	8	0.47	0.54	80.1

647

648 **Table 3:** Summary of goodness of fit statistics for in situ versus laboratory data.

649

650

651

Sub-catchment 2				Sub-catchment 5				Sub-catchment 8			
TP ( $\mu\text{g P l}^{-1}$ )		TRP ( $\mu\text{g P l}^{-1}$ )		TP ( $\mu\text{g P l}^{-1}$ )		TRP ( $\mu\text{g P l}^{-1}$ )		TP ( $\mu\text{g P l}^{-1}$ )		TRP ( $\mu\text{g P l}^{-1}$ )	
In situ	Lab	In situ	Lab	In situ	Lab	In situ	Lab	In situ	Lab	In situ	Lab
50	45	20	36	50	80	20	42	50	48	20	7
60	77	20	40	70	72	20	36	160	161	20	27
170	174	60	61	210	197	50	83	390	371	50	61
140	138	50	52	150	133	40	47	250	194	60	56
70	73	20	25	60	68	20	46	70	60	30	15

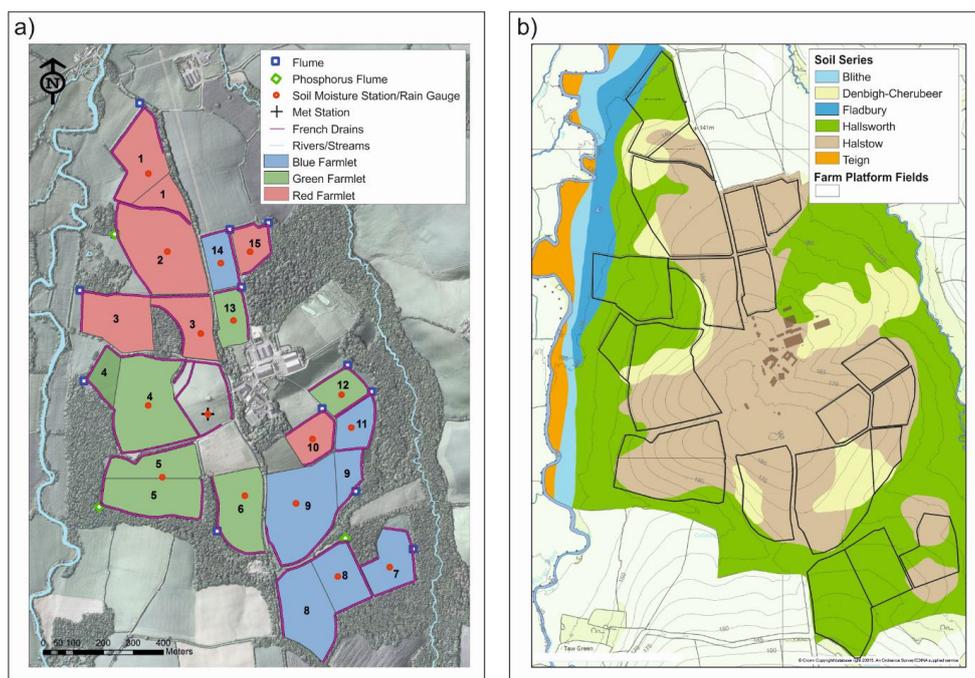
652

653 **Table 4.** Comparison of TP and TRP data obtained from in situ analysers and laboratory  
 654 analysed manual samples of discharge sampled at the same time (i.e. genuine temporal pairs).

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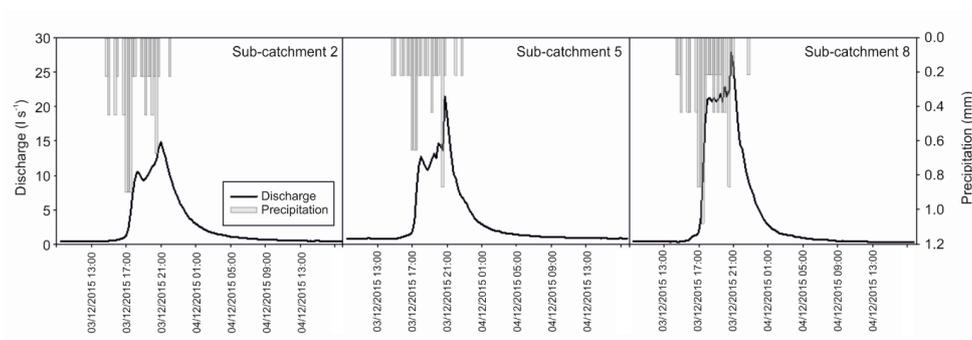
658

659 **Figure 1:** Maps of a) the North Wyke Farm Platform including infrastructure and b) soil

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series distribution.

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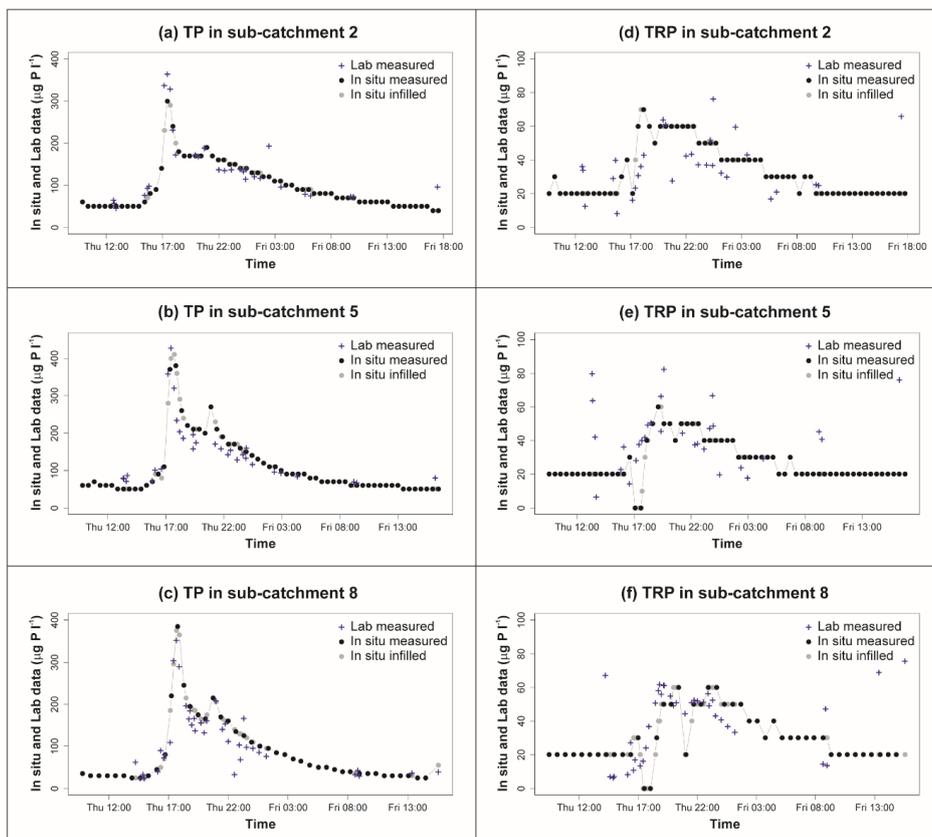


662

663 **Figure 2:** Discharge and precipitation for each sub-catchment.

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667 **Figure 3:** Time series plots for Total Phosphorus and Total Reactive Phosphorus showing the

668 data measured in situ relative to that measured in the laboratory physical sample, and the

669 modelled 'in filled' data.

670

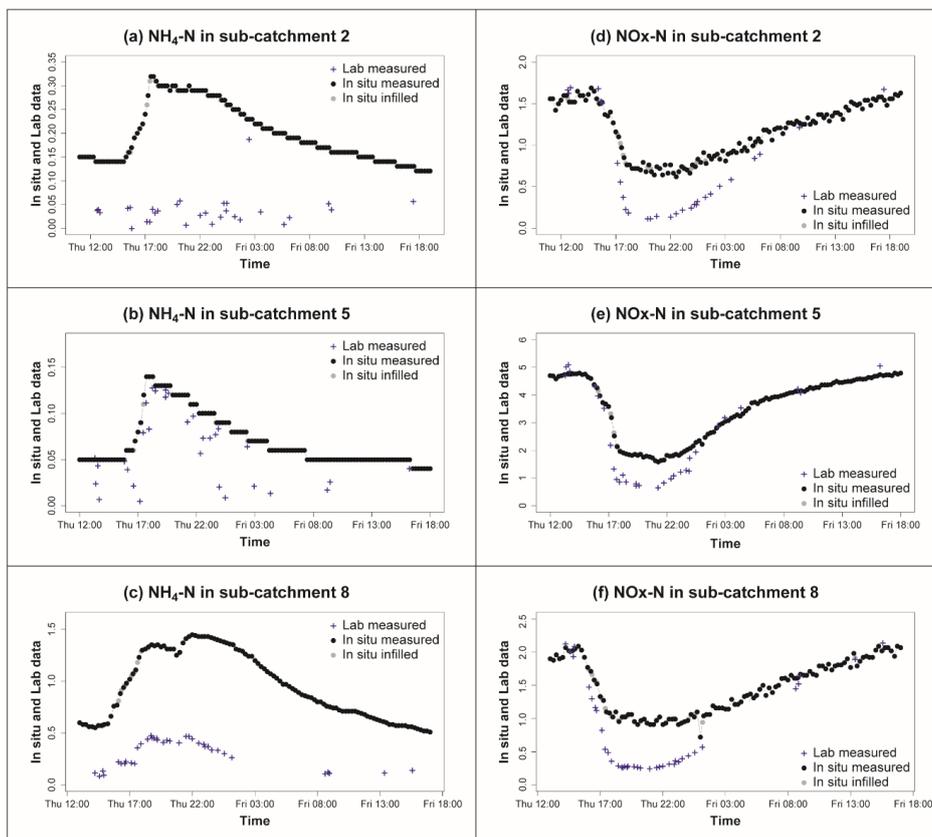
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677 **Figure 4:** Time series plots for  $\text{NH}_4\text{-N}$  and  $\text{NOx-N}$  showing the data measured in situ relative

678 to that measured in the laboratory physical sample, and the modelled 'in filled' data.

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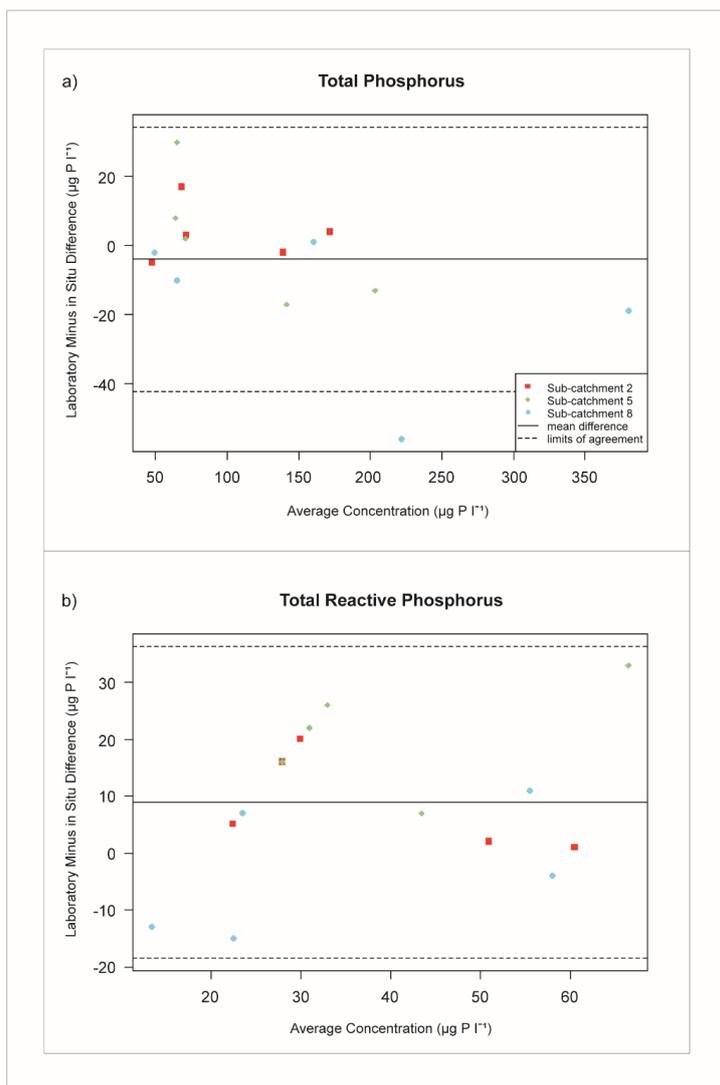
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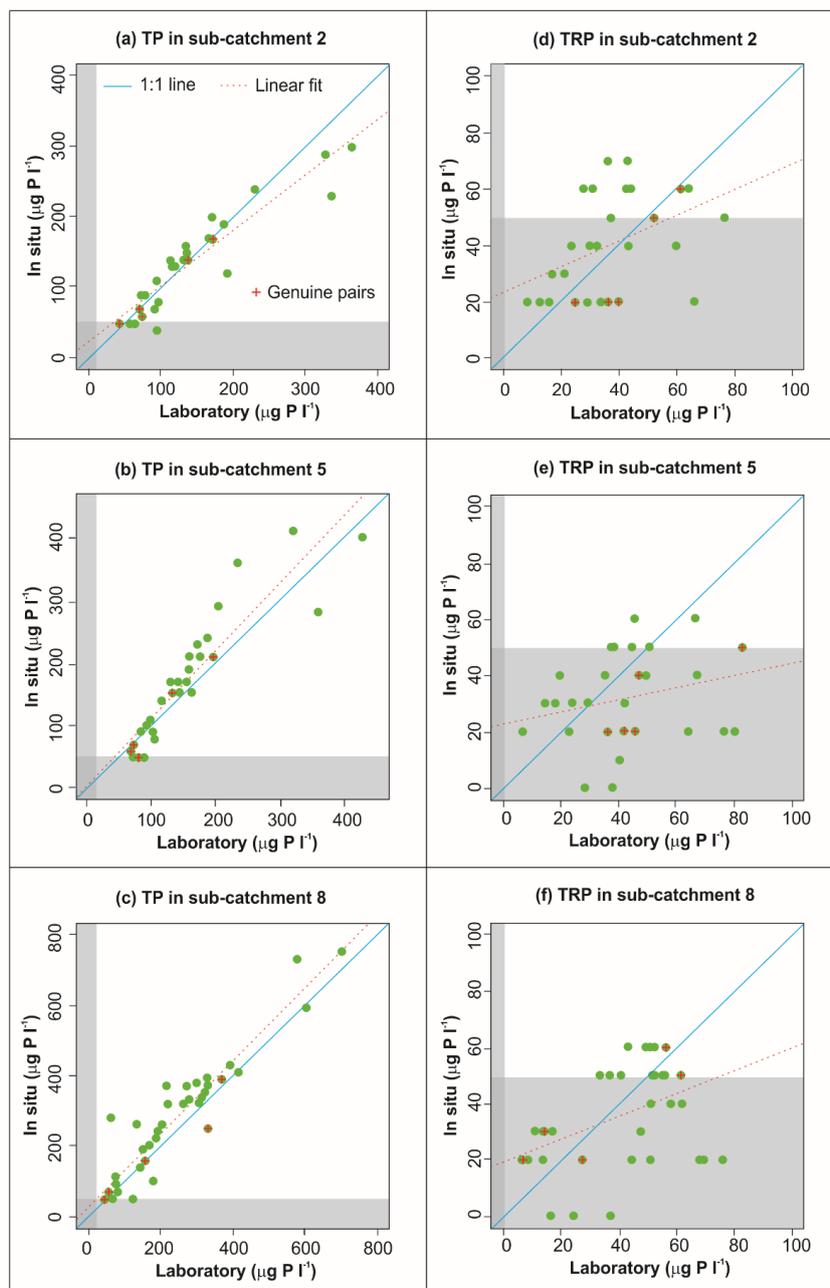
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**Figure 5.** Tukey mean-difference plots showing the average concentration of the two measurement methods against the difference between the two values (for TP and TRP only).

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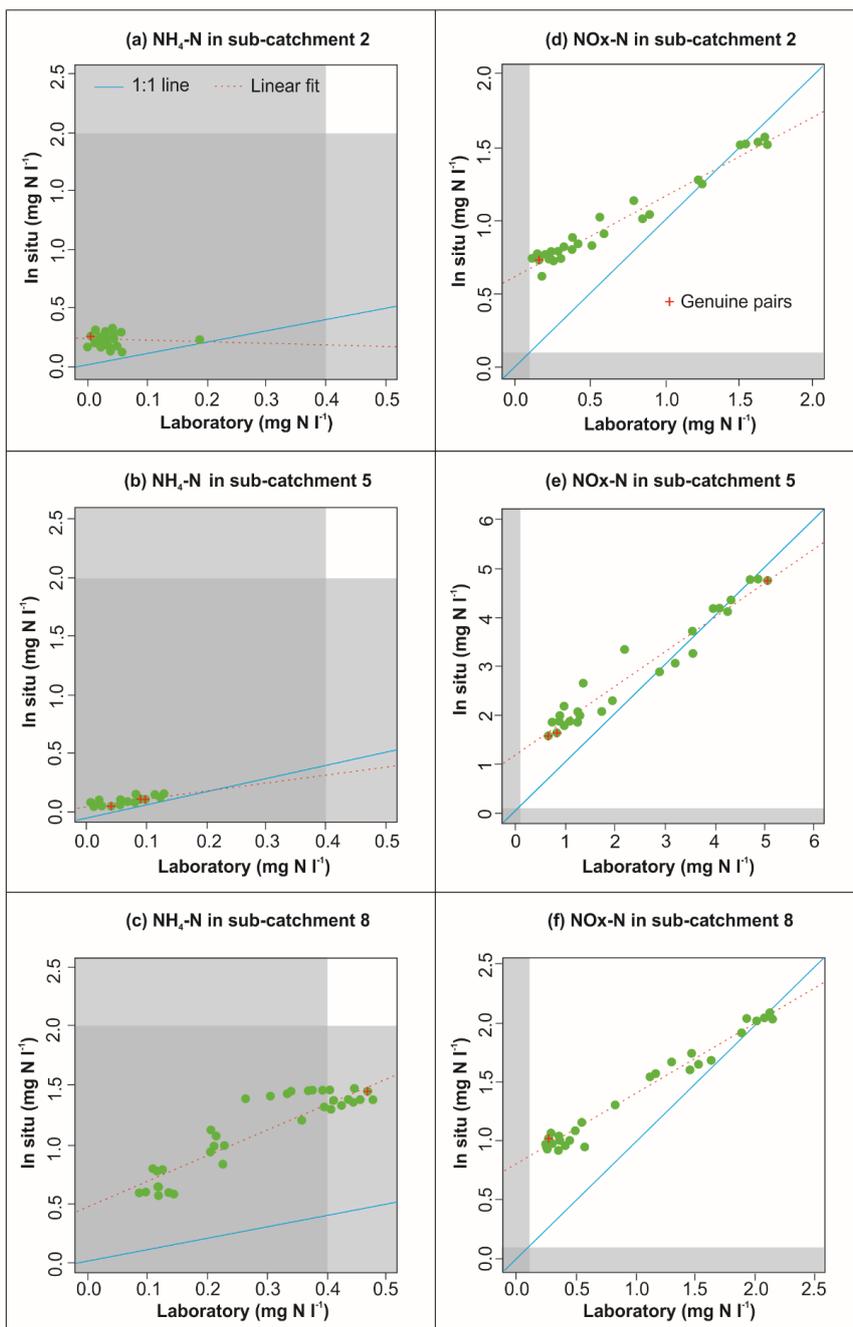


692

693 **Figure 6:** Scatterplots for paired TP and TRP data. Shaded grey areas indicate areas below

694

limits of analysis for accurate determination.



695

696 **Figure 7:** Scatterplots for paired NH<sub>4</sub>-N and NO<sub>x</sub>-N data. Shaded grey areas indicate areas

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below limits of analysis for accurate determination.