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# Accepted Manuscript

Temporal and spatial variation in pharmaceutical concentrations in an urban river system

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	ACCEPTED MANUSCRIPT
1 2	Temporal and spatial variation in pharmaceutical concentrations in an urban river system
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15	
16	
17	Abstract
18	Many studies have quantified pharmaceuticals in the environment, few however,
19	have incorporated detailed temporal and spatial variability due to associated costs in
20	terms of time and materials. Here, we target 33 physico-chemically diverse
21	pharmaceuticals in a spatiotemporal exposure study into the occurrence of
22	pharmaceuticals in the wastewater system and the Rivers Ouse and Foss (two diverse
23	river systems) in the city of York, UK. Removal rates in two of the WWTPs sampled (a
24	carbon activated sludge (CAS) and trickling filter plant) ranged from not eliminated
25	(carbamazepine) to >99% (paracetamol). Data comparisons indicate that
26	pharmaceutical exposures in river systems are highly variable regionally, in part due to
27	variability in prescribing practices, hydrology, wastewater management, and urbanisation
28	and that select annual median pharmaceutical concentrations observed in this study
29	were higher than those previously observed in the European Union and Asia thus far.

30 Significant spatial variability was found between all sites in both river systems, while 31 seasonal variability was significant for 86% and 50% of compounds in the River Foss 32 and Ouse, respectively. Seasonal variations in flow, in-stream attenuation, usage and 33 septic effluent releases are suspected drivers behind some of the observed temporal 34 exposure variability. When the data were used to evaluate a simple environmental exposure model for pharmaceuticals, mean ratios of predicted environmental 35 concentrations (PECs), obtained using the model, to measured environmental 36 37 concentrations (MECs) were 0.51 and 0.04 for the River Foss and River Ouse, respectively. Such PEC/MEC ratios indicate that the model underestimates actual 38 concentrations in both river systems, but to a much greater extent in the larger River 39 40 Ouse.

Keywords: LC-MS/MS; surface water; wastewater; seasonal; exposure; predicted
environmental concentration

43

### 44 **1.0 Introduction**

Determining pharmaceutical exposures in environmental matrices has become a 45 substantial area of research since the 1990s (Daughton, 2016). The presence of 46 47 pharmaceuticals in freshwater systems has now been documented globally, with research especially focused in Europe and North America (aus der Beek et al., 2016). 48 Pharmaceuticals primarily enter the environment through patient use when an 49 unmetabolised fraction is excreted and subsequently passes through wastewater 50 treatment plants (WWTPs), which are typically not designed to remove such organic 51 52 contaminants (Luo et al., 2014). Consequently, WWTPs are significant sources of pharmaceuticals to the environment (Lindholm-Lehto et al., 2016). A recent study of 53 54 United Kingdom (UK) WWTPs estimated that 13% of effluent discharges could pose 55 risks to the receiving environment regarding pharmaceutical exposures (Comber et al.,

56 2018). Removal rates are highly variable between treatment types (Kasprzyk-Hordern et al., 2009; Luo et al., 2014), seasons (Golovko et al., 2014), and even within treatment 57 58 plants themselves (Verlicchi et al., 2012). Moreover, removal rates have only been 59 estimated for a small fraction of the total number of pharmaceuticals in use (Boxall et al., 2014) and only a few studies have reported WWTP removals in the UK specifically 60 (Comber et al., 2018; Kasprzyk-Hordern et al., 2009, 2008). WWTP removal rates are 61 valuable parameters, and their inclusion in occurrence modelling substantially improves 62 the accuracy of pharmaceutical exposure predictions (Burns et al., 2017; Verlicchi et al., 63 64 2014).

The potential for, and extent of, effects posed by pharmaceutical exposure to non-65 target organisms, such as fish or invertebrates, is largely unknown (Vasquez et al., 66 2014). However, there is mounting evidence that select pharmaceuticals are having 67 68 deleterious effects at environmentally relevant (i.e. real-world) concentrations. Examples documented environmentally relevant 69 of effects at concentrations include antidepressants causing behavioural changes in fish (fluoxetine) (Mccallum et al., 2017). 70 disruption during early development (venlafaxine) (Thompson et al., 2017), the 71 equivalent of human side effects from exposure to the anti-diabetic drug metformin 72 73 (Niemuth et al., 2015) or the feminization of wild fish populations downstream of a 74 pharmaceutical manufacturing facility in France (Sanchez et al., 2011). It is therefore 75 important to characterise the source and fate of pharmaceuticals in the aquatic environment to aid in risk assessment as approaches evaluating potential adverse effect 76 concentrations emerge. 77

To adequately characterise the fate of pharmaceuticals in the environment, robust monitoring campaigns which include seasonal or year-long sampling covering a range of compounds at a reasonable spatial resolution are required. However, only a small number of spatiotemporal exposure studies have been performed that meet these

82 criteria (Baker and Kasprzyk-Hordern, 2013; Daneshvar et al., 2010; Kasprzyk-Hordern 83 et al., 2008; Paíga et al., 2016). These exposure studies are extremely valuable as they 84 provide detailed information which can be related back to the myriad of factors (many 85 varying both seasonally and temporally) that influence environmental concentrations of 86 pharmaceuticals including hydrology (Kasprzyk-Hordern et al., 2008), WWTP removal efficiency (Silva et al., 2014), pharmaceutical usage (Sun et al., 2014), and in-stream 87 removal processes (e.g. biodegradation and sorption to sediment) (Daneshvar et al., 88 89 2010; Camacho-Munoz et al., 2010; Moreno-González et al., 2014). In combination, the impact of these processes on pharmaceutical exposure and fate is largely unknown but, 90 if better defined, could improve exposure prediction approaches and offer greater 91 92 confidence, in terms of exposure, when evaluating risks that pharmaceuticals may pose 93 to the environment.

Recently, a handful of aqueous rapid pharmaceutical determination high-94 performance liquid chromatography tandem mass spectrometry (HPLC-MS/MS) 95 96 methods have been developed that achieve comparable limits of detection (LODs) to 97 those including sample pre-concentration or clean-up (Anumol et al., 2015; Boix et al., 2015; Campos-Manas et al., 2017; Furlong et al., 2014; Oliveira et al., 2015). Such 98 99 methods involve utilising larger than normal injection volumes (~100 µL) to increase the 100 likelihood of detection (Petrie et al., 2016). Removal of the extraction step reduces 101 sample preparation time and can increase the number of samples that can be processed 102 (highly beneficial to large spatiotemporal exposure campaigns). A significant analytical 103 problem arising during pharmaceutical quantification is matrix effects (typically mass 104 spectrometric ionisation enhancement or suppression). The presence of background 105 interferences in "dirty" matrices (e.g. streams, WWTP effluent, etc.) can co-elute with 106 target analytes and impair quantification past the point of suitability (Petrović et al., 107 2005). Several approaches have been attempted to reduce matrix effects including

108 sample pre-concentration and clean-up to help isolate target pharmaceuticals (Van De 109 Steene et al., 2006). Such pre-concentration, however, is difficult to optimise, time 110 consuming, costly, and may also concentrate interfering analytes, thus unintentionally 111 increasing matrix effects (Yu et al., 2012). Matrix interferences have been reported to be 112 comparatively lower for rapid determination methods than more costly and laborious 113 sample pre-concentration/clean-up methods (Anumol et al., 2015).

114 In this study, which was performed in the frame of the Innovative Medicines Initiative iPiE project on intelligent assessment of pharmaceuticals in the environment, 115 116 we validate and apply a rapid determination aqueous HPLC-MS/MS method for the quantification of 33 physico-chemically diverse pharmaceuticals to a year-long surface-117 water exposure campaign. Monitoring was conducted during 2016 at 11 sites along the 118 119 urbanised and larger River Ouse and smaller, more rural River Foss which converge within the city of York, UK (Figure 1). The monthly sampling design provided good 120 temporal resolution while unparalleled spatial resolution was achieved in the two 121 122 contrasting river systems. In addition, influent and effluent samples from two of the 123 WWTPs that serve the city were collected when possible and removal efficiencies estimated. Predicted exposure concentrations (PECs) were calculated for both rivers 124 125 using a simple model and the model was then evaluated against annually averaged 126 measured environmental concentrations (MECs) calculated from the monthly sampling 127 data.

#### 128 **2.0 Methods**

129 2.1 Study area and sample collection

130 2.1.1 Study Compounds

Study compounds were selected based on those previously detected in the York
 river system during an initial scoping study in which 95 pharmaceutical and degradation

products were surveyed (Burns et al., 2017). From these results, 32 pharmaceuticals
were selected due to either their known or expected presence. An additional compound,
gabapentin, was also included in the study due to its high usage, resistance to
environmental degradation, and ecotoxic potential (Herrmann et al., 2015).

137 2.1.2 Study Area

The River Ouse and River Foss were chosen for the study, as they flow through 138 139 the city of York, UK, and converge downstream of the city centre (Figure 1). The two rivers represent differing levels of urbanisation and size. To minimise potential variability, 140 141 grab water samples were collected from the network of 11 sampling sites in the same order and on approximately the same day and time each month from January to 142 December 2016. Site locations were strategically chosen based on their ease of access 143 and position in relation to WWTP outfalls. Both rivers were sampled with sufficient 144 spatial resolution to build concentration profiles and increase the probability of detecting 145 transient pharmaceuticals in the absence of composite sampling techniques. Three 146 WWTPs serve the city within the sampling network (Figure 1). WWTP A is a trickling 147 filter plant and serves a population of 18 600, WWTP B is a conventional activated 148 sludge (CAS) facility serving a population of 27 900, while WWTP C is a surplus 149 150 activated sludge (SAS) plant serving a population of 180 500. Sampling site and WWTP 151 characteristics along with dates of sampling are detailed in Supplemental Material, 152 Tables S1 and S2.

153 2.1.3 Sample Collection

All samples collected were subject to the same sampling protocol. At each site, three 1-L field replicates were collected from the centroid of flow (when possible); sampling sites had been previously determined to be well-mixed, therefore sampling in a single location was deemed appropriate (Supplemental Material, Figure S1). For each field replicate, a 10-mL aliquot was drawn into a 24-mL disposable syringe and filtered

159 through a primed 0.7-µm glass-fibre filter (GF/F) (Whatman Inc.) into an amber glass vial and immediately frozen in the field using dry ice. To demonstrate that field filtration and 160 161 collection did not contaminate samples, three field blanks per sampling visit were 162 collected. HPLC-grade water was brought to the field, filtered and prepared identically to 163 field samples. Samples were then returned to the laboratory and stored at -18°C until analysis which occurred within seven days. The concentration reported for each sample 164 165 per site is the median of the three field replicates collected. The filtering of samples in the field is beneficial as it removes particulates which can extend HPLC column life, 166 reduce instrument maintenance, as well as remove bacteria associated with particulates 167 that could facilitate analyte degradation. There is a formal possibility that analytes could 168 169 be retained on the filter; however pharmaceutical filtration studies including 26 170 compounds (acids, bases and amphoteres) ranging in hydrophobicity (logKow -2.3 to 171 6.3) suggest these losses will be insignificant (<5%) (Mompelat et al., 2013), thus an 172 assessment of filter losses has not been repeated here.

173 2.2 High performance liquid chromatography-tandem mass spectrometry

174 A Thermo Scientific<sup>™</sup> TSQ Endura MS operating in multiple reaction monitoring 175 mode interfaced with an EASY-Max NG<sup>™</sup> heated electrospray source operating in positive mode was used for pharmaceutical detection. Two transitions were monitored 176 177 for each analyte and the m/z and collision energy optimised using the Thermo<sup>TM</sup> Tune 178 2.0 software, summarised in Supplemental Material, Table S3. Chromatographic separation was achieved with a Dionex Ultimate 3000 HPLC (Thermo Scientific™) 179 180 equipped with a 100-uL sample injection loop and autosampler maintained at 4°C. Mobile phase A consisted of HPLC-grade water amended with 12-mL of 1 M formic acid 181 182 and 10-mL of 1 M ammonium hydroxide for a total volume of 1-L, and mobile phase B was 100% methanol (Furlong et al., 2014). The chromatographic conditions and 183 184 program are reported in the Supplemental Material Table S4.

185 Internal standard (IS) calibration was used to quantify the pharmaceuticals in the method described. For reasons of expense and availability, not all pharmaceuticals had 186 187 a corresponding isotopically labelled internal standard (ILIS) (Supplemental Material, 188 Table S3). In these cases, atrazine-d<sub>5</sub> was used and has been previously determined 189 suitable for this role (Furlong et al., 2014). Samples were fully thawed and a 995-uL 190 aliquot pipetted into a 1.5-mL LC vial and a 5-µL spike of IS solution (80 ng/L) added. 191 Samples were immediately analysed after preparation. Peak detection criteria were in 192 accordance with Commission Decision (2002/657/EC). Due to analytical complications, fexofenadine could not be quantified in the April surface-water samples. Further details 193 194 of peak gualification and guantitation are provided in the Supplementary Material.

195 The use of ILIS is a good strategy to compensate for matrix effects (Stüber and 196 Reemtsma, 2004). This is not a perfect solution as matrix effects can still influence 197 quantification, possibly due to a slight difference in retention time ( $t_{\rm B}$ ) between the ILIS 198 and target analyte resulting in differing ionisation efficiencies (Wang et al., 2007). 199 Therefore, sample matrix spikes were routinely prepared and analysed with all sample 200 batches to provide an indication of the presence of interferences which cause signal 201 suppression/enhancement and could impact quantification. In this study, acceptable 202 matrix recovery was considered to be 70% to 120% in accordance with previously 203 published methods (Boix et al., 2015; USEPA, 2016; Furlong et al., 2014). Matrix 204 'recovery' falling outside this range indicates signal suppression/enhancement could be 205 occurring and samples should quantitatively be interpreted with caution. At least three 206 matrix spike samples from different sampling sites were prepared per analytical batch to 207 monitor for matrix effects throughout the sampling campaign as the sample matrices are 208 heterogenous and likely to vary temporally. Surface-water matrix spikes were prepared 209 by spiking 20 µL of 80 ng/L or 200 ng/L calibration solution into a sample replicate with 5 µL of IS solution. The much higher ambient concentration of pharmaceuticals in WWTP 210

influent and effluent required the matrix spike samples to be prepared at a higher
concentration, 4000 ng/L. Matrix recovery was calculated by subtracting the ambient
sample concentration and dividing by the concentration spiked.

With each sample batch at least three calibration check samples (CCSs) were prepared to monitor accuracy throughout the analytical batch (injected every 10 samples). These CCSs were prepared to a concentration of 80 ng/L by pipetting 20  $\mu$ L of the relevant calibration solution into 975  $\mu$ L of HPLC grade water and spiked with 5  $\mu$ L of IS solution. At the end of each batch a 4 ng/L calibration solution spike, prepared similarly, was also injected. The accuracy of these CCSs was required to be within 20% or affected samples were re-analysed (Furlong et al., 2014; USEPA, 2016).

This formed part of a rigorous quality control plan which was followed during environmental sample analysis using a series of sample matrix spikes, calibration solution spikes, field blanks, and laboratory blanks randomly dispersed throughout analytical batches. Further detail of quality control, how these samples were prepared and results are reported in the Supplementary Material.

226 2.3 Analytical method validation

Method validation included an assessment of precision (inter- and intra-day), limits of detection, limits of quantification, and recovery from all studied matrices. The methods and results with which each of these parameters were assessed are reported in the Supplemental Material.

231 2.4 WWTP removal efficiency

Due to access restrictions, 24 h composite samples for influent and effluent could only be collected once from WWTP A and B during summer 2016 (Supplementary Material, Table S2). Only grab samples unsuitable for estimating removals could be collected from WWTP C. WWTP removal efficiency was estimated, when appropriate,

236 for WWTP A and B based on mean influent and effluent concentrations according to 237 Equation 1. In this context 'removal' is the change in concentration between influent and 238 effluent which does not represent true removal, but rather partitioning to the solid phase 239 and/or the formation of transformation products. Negative removals can occur, 240 potentially due to sampling limitations (e.g. longer than 24 h hydraulic/sludge retention time) (Ort et al., 2010), from the conversion of conjugated metabolites back to the parent 241 242 compound during treatment (Verlicchi et al. 2012), or desorption from sludge during 243 secondary treament (Blair et al., 2015).

244 % Removal= 
$$\left(1 - \frac{\text{Effluent}}{\text{Influent}}\right) \times 100$$

#### 245 2.5 Statistical Analysis

246 Data analysis was performed using Graphpad Prism (Graphpad Software, 2017). 247 To use statistical tests when non-detects were present, data substitution according to 248 Equation 2 was undertaken. This approach was suggested to be appropriate for left 249 censoring of up to 40% of a dataset (Antweiler, 2015). If the non-detect frequency for a compound was greater than 40%, it was not included in statistical testing. To determine 250 251 whether significant spatial differences existed between sites, pairwise t-tests were conducted based on the monthly concentrations (Furlong et al., 2017). To determine 252 whether any analytes were seasonally variable in each river, concentrations from sites 253 254 F3-F4 and O3-O4 were grouped by season and a Friedman's Test followed by a Dunn's multiple comparisons post hoc test was undertaken. These sites were used in the 255 256 seasonality test due to their downstream location in relation to WWTP A and B, as well 257 as their location in relation to Environment Agency flow gauges (Figure 1) as the flow 258 recorded at these gauges was not representative of flow conditions at the remaining study sites (Center for Ecology & Hydrology, 2016). 259

260 Substitution = 
$$\frac{\sqrt{2}}{2}$$
\*LOD [2]

10

[1]

261 2.6 Predicted environmental concentrations

262 Annual average MECs were compared to PECs to gauge the accuracy of simple 263 exposure algorithms commonly used for the prioritisation of pharmaceuticals and risk 264 assessment (Burns et al., 2017). Local annual pharmaceutical usage data were obtained from the National Health Service Business Authority (National Health Service, 2016), 265 while wastewater generation was assumed to be 200 L/person day (European 266 267 Medicines Agency, 2006). Experimental WWTP removal rates (Eqn. 1) were used with river specific dilution factors based on the average flow from sampling days to generate 268 269 a PEC for both rivers. PEC calculations were based on the approach suggested by the 270 European Medicines Agency (2006). Parameters and equations used to predict the 271 PECs are provided in the Supplemental Material Table S6.

# 272 3.0 Results & Discussion

#### 273 3.1 Method performance and quality control

274 The method was determined to be sufficiently reproducible as assessed by the relative standard deviation of multiple injections (n=8) during (5.5 %RSD) and across (7.5 275 %RSD) analysis days according to USEPA (2016) guidelines and Boix et al. (2015) 276 277 where an RSD $\leq$  20% above the LOQ (i.e. 80 ng/L) is desirable. The limits of detections 278 (LOD) ranged from 0.9 ng/L (carbamazepine) to 12.4 (gabapentin) and an LOD <10 ng/L 279 was achieved for 91% of analytes (Table S5). There were no guantifiable concentrations 280 of any of the target pharmaceuticals in field blanks collected routinely throughout the 281 monitoring campaign. Routine matrix spikes in surface water fell within the acceptable 282 70 – 120% recovery range for concentrations of 80 and 200 ng/L, indicating that 283 throughout the sample analysis guantification was not unacceptably impaired due to 284 matrix effects (Figure 2). Matrix effects were observed in WWTP effluent and influent, a phenomenon also reported by others, and suggested to be due to the presence of a 285 286 greater proportion of chemical species that can affect consistent ionisation in

287 comparison to surface water (Boix et al., 2015; Oliveira et al., 2015). In effluent 13% and 288 in influent 19% of analytes fell outside the acceptable matrix signal response, identified 289 in Figure 2 and 3. Signal enhancement was most prominent for diphenhydramine in both 290 influent and effluent (442% and 375%, respectively), while metformin (214%) and 291 tramadol (156%) also exhibited significant signal enhancement in influent. In this study, a slight shift in relative  $t_{\rm B}$  of the analyte with respect to its ILIS, was observed in WWTP 292 293 influent and effluent in comparison to surface water, which, in addition to it containing a 294 larger number of chemical constituents, could help explain why matrix effects were not well compensated for all analytes using isotopically labelled internal standards. WWTP 295 296 influent and effluent matrix spikes indicate that caution is needed when interpreting 297 quantitative results and removal efficiencies due to significant matrix effects, while matrix 298 spikes in surface water indicate that matrix effects are sufficiently compensated for by 299 the internal standards.

300 3.2 Pharmaceuticals in WWTPs

301 The highest summed pharmaceutical concentrations in influent were observed in 302 samples from WWTP B, while highest summed concentrations in effluent were observed 303 in samples taken at WWTP A. Paracetamol had the highest concentration in all WWTP 304 influents, 282, 186 and 117 µg/L at WWTP B, A and C, respectively. In effluent, gabapentin had the highest concentration (8541 ng/L) at WWTP C followed by 305 306 metformin (6111 ng/L) at WWTP A and fexofenadine (2094 ng/L) in effluent at WWTP C. Seven pharmaceuticals (diphenhydramine, norethisterone, oseltamivir, raloxifene, 307 308 sertraline, triamterene and verapamil) were not detected in any WWTP sample. Average 309 concentration and standard deviation (SD) of WWTP influent and effluent samples are 310 reported in the Supplemental Data Table S10.

In a global review of pharmaceuticals in WWTPs, Verlicchi et al. (2012) reported
 influent concentrations for many compounds also observed in the WWTP samples in this

313 study. Codeine, paracetamol, gabapentin, hydrocodone, tramadol, erythromycin, 314 trimethoprim, diltiazem, atenolol, propranolol, carbamazepine, gabapentin, cimetidine, 315 and ranitidine influent concentrations all fell within the ranges reported by Verlicchi et al. 316 (2012), while concentrations of amitriptyline were an order of magnitude lower. A study 317 of effluents in the European Union (EU) reported average concentrations an order of magnitude lower than those determined here for tramadol, codeine, citalopram, 318 319 fexofenadine, diltiazem, ranitidine, and amitriptyline, while effluent concentrations were 320 similar for venlafaxine, trimethoprim, carbamazepine, and sulfamethoxazole in the York 321 samples (Loos et al., 2013).

322 The estimated removal efficiency in each WWTP is presented for all detected analytes in Figure 3. The median removal efficiency was estimated to be 75% in WWTP 323 A and 38% in WWTP B. Paracetamol was the analyte most efficiently removed at both 324 treatment plants (>99%), while removals greater than 75% were reported for gabapentin, 325 326 ranitidine, atenolol, sulfamethoxazole, metformin, and codeine. Despite being a trickling 327 filter plant which might be expected to have poorer pharmaceutical removal than CAS systems (Kasprzyk-Hordern et al., 2009), WWTP A had similar and even greater 328 removals for select compounds (i.e. carbamazepine, diltiazem, citalopram, erythromycin, 329 cimetidine, and ranitidine). In the UK specifically, similar removals were reported 330 331 previously (Kasprzyk-Hordern et al., 2009) for trimethoprim, amitriptyline, diltiazem, 332 cimetidine, gabapentin, and paracetamol, while sulfamethoxazole, erythromycin, 333 codeine, tramadol, carbamazepine, propranolol and ranitidine were, in general, more 334 efficiently removed for this study. WWTPs with similar treatment capabilities were also 335 studied previously in the UK (Kasprzyk-Hordern et al., 2009). In comparison with results 336 reported here, WWTP removal rates were highly variable despite operating in the same 337 region and employing similar treatments, a conclusion also observed in other regions 338 (Verlicchi et al., 2012). The single sampling event in the WWTPs is limited, however

339 these estimates are still useful for comparative purposes. For example, sitagliptin removal efficiency (25 - 40%) has not been previously reported to the authors' 340 341 knowledge. Therefore, while WWTPs are significant sources of pharmaceuticals entering 342 the environment, analysis of WWTP removal efficiencies (i.e. reduction in parent 343 pharmaceutical concentration from influent to effluent) as documented in this and previously published studies, demonstrate that WWTPs are generally decreasing the 344 345 aguatic environmental burden by significantly reducing certain parent pharmaceutical 346 concentrations (not considering degradates or transformation products) for many of the 347 compounds studied.

348 3.3 Pharmaceuticals in Surface Water

Of the 33 pharmaceuticals monitored, 21 were detected in all 12 months in 349 samples from the River Foss. Three compounds, oxazepam, verapamil, and triamterene, 350 351 were not detected in any Foss sample. The remaining nine study compounds, norethisterone, 352 diazepam, diphenhydramine, loratadine, oseltamivir, raloxifene, 353 sulfamethoxazole, sertraline, and temazepam, were sporadically detected from month to month in this river. In comparison, ten compounds (carbamazepine, codeine, 354 fexofenadine, gabapentin, hydrocodone, lidocaine, metformin, paracetamol, tramadol, 355 and trimethoprim) were detected in all 12 months in the River Ouse samples. Eight 356 357 compounds were not detected in any Ouse sample: diazepam, loratadine, oseltamivir, 358 oxazepam, raloxifene, sulfamethoxazole, triamterene, and verapamil. The highest five 359 annual median concentrations followed the same trend in both rivers: 360 metformin>gabapentin>paracetamol>fexofenadine>tramadol, indicating that usage patterns, WWTP removal and environmental fate for the most prevalent pharmaceuticals 361 362 are similar in these two systems. The range, detection frequency and annual median for each pharmaceutical in both river systems is reported in Tables 1 and 2. 363

364 Monthly total pharmaceutical concentrations at each sampling site are presented in Figures 4 and 5. These concentration figures provide a spatiotemporal overview of the 365 366 relationship between sampling sites, rivers, and WWTPs serving the city. Monthly 367 summed concentrations are higher in the River Foss (e.g. above 2000 ng/L) at sites 368 downstream of the WWTP in comparison to the River Ouse, where most monthly 369 summed concentrations are below 1000 ng/L despite the WWTPs on the River Ouse 370 serving a larger population. This is due to greater dilution of discharged effluent in the 371 Ouse; for example, flow ranged from 9.2 to 233 m<sup>3</sup>/s in the Ouse, compared with 0.0096 to 1.68 m<sup>3</sup>/s in the Foss on sampling days (Figure 1). For the sites immediately 372 373 downstream of the WWTPs (O3, O6, and F2), the months with the lowest flows, July and 374 June, yielded both the most analytes and the highest concentrations. Thus, 375 concentrations appear to be inversely proportional to flow at site F2, similarly to observations reported previously (Kolpin et al., 2004). The trend is not continued moving 376 377 downstream in the River Foss (sites F3-F5), potentially due to pharmaceutical losses 378 stemming from dilution or in-stream removal processes such as biodegradation or 379 sorption to sediment (Moreno-González et al., 2014), or due to pharmaceutical 380 contributions from domestic septic systems (Carmona et al., 2014), and/or inputs from 381 combined sewer overflows (CSO) (Phillips et al., 2012). In the Foss, a substantial spike 382 downstream of F2 in paracetamol (9822 ng/L) was detected in the March sampling along 383 with less intense spikes from other pharmaceuticals, such as metformin (2592 ng/L). 384 These observations may be explained by local septic tank effluent entering the river 385 downstream of the F2 site, captured during the March sampling period. Paracetamol can 386 be >99% removed and metformin >93%, in conventional water treatment (Figure 3), 387 therefore the spike in March concentrations might be explained by releases of septic 388 effluent (James et al., 2016). James et al. (2012) reported paracetamol concentrations of 389 5000 ng/L at a septic effluent impacted site and identified it as a possible tracer of septic

390 system contamination. Combined sewer overflow (CSO) releases could provide an alternative explanation for the concentration spike (Philips et al. 2012), as a CSO is 391 392 located just upstream of the F3 site. Low rainfall (University of York, 2018) prior to 393 sampling suggest CSOs would not likely be in operation, therefore septic effluent 394 releases provide a plausible explanation. Concentrations in the River Ouse varied less 395 month to month than in the Foss, and a relationship with flow was less clear, with March 396 and May in general having slightly greater total concentrations. March has also been 397 reported to have the highest monthly concentration in recent temporal studies (Padhye et al., 2014; Sun et al., 2014). Sun et al. (2014) suggested March coincided with a spike 398 399 in pharmaceutical usage and reduced WWTP removal capacity. This may explain the slightly higher concentrations observed in the River Ouse at sites upstream of the Foss 400 401 confluence (O1-O4), while the spike in May (River Ouse) coincides with decreased river 402 flow (Figure 1).

403 Metformin, a type II diabetes drug, had the highest annual median concentration 404 (1117 and 237 ng/L in the Foss and Ouse, respectively), followed by gabapentin (anticonvulsant) (843 and 230 ng/L, Foss and Ouse, respectively) and paracetamol 405 406 (analgesic) (209 and 77.6 ng/L, Foss and Ouse, respectively). This trend is different from 407 those observed in previous temporal exposure campaigns studying similar compounds 408 throughout the world. For example in China, Zhang et al. (2015) studied urbanized rivers 409 and found antibiotics the most frequently detected pharmaceuticals. They did, however, 410 report atenolol as having one of the highest annual median concentrations (53 ng/L), 411 which is similar to the median concentration for this compound reported at site F2 (55.4 412 ng/L) in the current study. In Spain, Camacho-Munoz et al. (2010) reported propranolol most frequently detected in surface water, with a higher average concentration (80 ng/L) 413 than observed in this study (20.1 ng/L). In Portugal, Paíga et al. (2016) reported 414 415 carbamazepine the most frequently detected pharmaceutical with an annual median of

416 31.7 ng/L, while other similarly studied compounds, citalopram and venlafaxine had annual median concentrations of 0.86 and 40.1 ng/L, respectively and trimethoprim was 417 418 not detected. In the River Foss, the highest annual median concentrations for 419 carbamazepine, citalopram and venlafaxine was 66 ng/L, 15.4 and 21 ng/L, respectively 420 while trimethoprim was detected in 100% of samples with an annual median of 30 ng/L. 421 In Sweden, carbamazepine was also most frequently detected and at a higher annual 422 mean than observed in York, 204 ng/L versus 66 ng/L in the River Foss, while atenolol 423 concentration was similar to that reported here (60.2 ng/L, compared to 55.4 ng/L) (Daneshvar et al., 2010). In a similar temporal study in Wales, tramadol and gabapentin 424 425 had the highest annual median concentrations (968 ng/L and 227 ng/L, respectively) 426 (Kasprzyk-Hordern et al., 2008). Median concentrations of: gabapentin, tramadol, 427 trimethoprim, paracetamol, carbamazepine, cimetidine and atenolol, in Wales were higher than we saw in York, while concentrations of diltiazem, atenolol, 428 429 sulfamethoxazole, and erythromycin concentrations in the River Foss were lower than observed in Wales (Kasprzyk-Hordern et al., 2008). These comparisons suggest that 430 431 annual pharmaceutical exposures in river systems are highly variable regionally, in part due to variability in prescribing practices, hydrology, wastewater management, and the 432 degree of urbanisation. In addition, certain annual median concentrations of 433 434 pharmaceuticals observed in this study are higher than those previously observed in the 435 European Union and Asia.

436 3.3.1 Spatial Trends

The spatial trends for both rivers are presented in Figure 6; significant differences between a site and the adjacent downstream site are also noted. Spatial trends are apparent in both rivers, the greatest number of significant differences (p<0.05) were found between the sites upstream and downstream of the WWTPs (i.e. F1-F2, O3-O4 and O5-O6) (Figure 6). In addition, significance increases were found when comparing

442 to sites further downstream. WWTPs make a significant contribution to pharmaceutical 443 concentrations in both river systems, however upstream sources of certain 444 pharmaceuticals exist in both rivers as significance was not achieved for cimetidine in 445 the Foss and paracetamol, codeine, trimethoprim, and atenolol in the Ouse. There are 446 WWTPs along the River Nidd (Figure 5) and upstream of sites O1 and F1 (>10 km) demonstrating that pharmaceuticals from upstream sources are transported into the city. 447 448 Concentrations are generally highest immediately downstream of the WWTPs and decrease moving to downstream sites, evidenced by difference in height (i.e. 449 concentration) between the bars from each site (Figure 6), similarly to observations in 450 previous studies (Kasprzyk-Hordern et al., 2008). The decrease in concentrations 451 452 moving downstream is variable between compounds indicating that in-stream 453 attenuation is compound specific. For example, carbamazepine concentrations are 454 similar between sites downstream of the WWTP in the River Foss (i.e. F2-F5), while 455 over the same stretch of river concentrations of hydrocodone and citalopram decreases by 51% and 38%, respectively (Figure 6). In the Ouse, all concentrations decrease 456 457 slightly from O3 to O4, however there is a slight increase occurring at O5, likely due to the confluence with the River Foss and again at O6, which is downstream of WWTP C. 458

459 In the River Foss, carbamazepine had only a single significant spatial difference 460 between the site upstream of WWTP A discharge (site F1) and the sites downstream of 461 the discharge. Carbamazepine has been reported to be resistant to biodegradation and stable in the environment (Moreno-González et al., 2014). In the River Ouse, all 462 463 pharmaceuticals exhibited spatially significant trends. Carbamazepine was significantly 464 different between each site downstream of WWTP B tested (i.e. O3 to O6). Since this did not occur in the River Foss over a similar distance. 13.3 km between sites F2 and F5 465 versus 11 km between sites O3 and O6, and the literature agrees that carbamazepine is 466 467 resistant to biotransformation, a combination of dilution (e.g. urban drainage/runoff) and

468 other pharmaceutical sources (i.e. River Foss) moving downstream could be a plausible469 explanation.

470 Overall, these results indicate that a wide variety of environmental processes such 471 as dilution and in-stream degradation are operating to differing extents in neighbouring 472 rivers leading to different spatial patterns in pharmaceutical concentrations between 473 sampling sites. For example, the reduction in concentrations moving downstream in the 474 River Foss may be symptomatic of in-stream removal processes such as photolysis or microbial degradation (Daneshvar et al., 2010), while fluctuating concentrations in the 475 476 River Ouse could be due to a complex dynamic between dilution and other pharmaceutical sources (i.e. tributaries, urban drainage) while natural removal 477 processes potentially operating in the Foss may be masked or occur to a lesser extent in 478 479 the larger Ouse system.

#### 480 3.3.2 Seasonal Variability

481 Temporal variability between the seasons (Figure 7) is presented similarly to the 482 approach for displaying spatial variability between sampling sites (Figure 6). Seasonal 483 differences in pharmaceutical concentrations exist in the two river systems, especially in 484 the River Foss. In both rivers, the lowest concentrations correspond with winter, the 485 season which had the highest average flow (2.7 times higher than the next highest season, autumn). Conversely, the highest mass loads occur in winter, 1.4 times higher 486 487 than the next highest season, spring. Lower concentrations in winter have also been 488 reported previously (Baker and Kasprzyk-Hordern, 2013; Kasprzyk-Hordern et al., 2008), 489 however several studies report higher concentrations in winter (Kot-Wasik et al., 2016; 490 Lindholm-Lehto et al., 2016; Zhang et al., 2015). In addition, the extent of concentration 491 variability between seasons differs between compounds, which could be due to 492 seasonal patterns in usage (Sun et al., 2014) or seasonal variability in photodegradation 493 or biodegradation, of which both processes can peak in summer, thus having a greater

494 impact on more readily biodegradable compounds (Lindholm-Lehto et al., 2016). In 495 general, autumn was the season with the second highest median concentrations, except 496 for paracetamol, where highest median values were observed during spring in both 497 rivers. This could be due to increased usage coinciding with symptomatic treatment of 498 illnesses more common in spring such as colds (Vatovec et al., 2016) in conjunction with 499 lower flows than winter. To determine whether concentrations between seasons were 500 significant, Friedman's test was used for pharmaceuticals with sufficient detections. 501 Concentrations of 17 compounds (86%) were found to vary significantly by season in the 502 River Foss, while amitriptyline, codeine, cimetidine, metformin, and ranitidine did not 503 vary seasonally. Nine compounds (50%) had significant seasonal differences in the 504 River Ouse, atenolol, carbamazepine, codeine, desvenlafaxine, gabapentin, lidocaine, 505 ranitidine, sitagliptin, and trimethoprim.

506 The reasons for temporal variations in pharmaceutical concentrations have varied 507 between studies with several reporting flow as the major driver, observing higher 508 concentrations during times of low flow (Kasprzyk-Hordern et al., 2008; Kolpin et al., 2004). Others suggest higher pharmaceutical concentrations in winter months coincide 509 with higher winter usage patterns (Sun et al., 2014) or decreased biodegradation in 510 511 winter (Moreno-González et al., 2014), while others found no significant differences 512 between sampled seasons (Camacho-Munoz et al., 2010). Due to higher concentrations 513 coinciding with low-flow months in this study, we also suggest that flow appears to be a major driver behind the observed seasonal variability in pharmaceutical concentrations 514 515 in the current study. The lack of significant seasonal differences found in the River Ouse 516 could be explained by a lower annual variability in flow on sampling days than the River 517 Foss (i.e. two orders of magnitude versus three). Further detailed investigation into the drivers behind the pharmaceutical concentrations observed both temporally and spatially 518 519 is required to differentiate between the possible explanations, and could include

520 comparing results with prescription data and flow and estimating the impact of in-stream 521 losses seasonally in different climates, and in river hydrological properties (e.g. depth 522 and flow). Such analyses will be facilitated by the detailed pharmaceutical monitoring 523 data reported in this study.

524

#### 525 *3.4 Comparisons of PECs and MECs*

526 The PEC/MEC ratios for each compound for which it was possible to calculate an annual average MEC are reported in Figure 8. A ratio greater than 1 indicates PECs 527 were higher than MECs and lower when less than 1. The PECs are severely 528 underestimated in the Ouse; this may be due to pharmaceuticals being transported from 529 530 upstream or problems with sewer connectivity within the sampling network not being 531 accounted for in the simplistic PEC calculation. Several studies have attempted to gauge 532 the accuracy of PECs by calculating a ratio with MECs, however the criterion for what 533 constitutes accurate is variable across studies (Burns et al., 2017). This assessment has 534 been previously limited to a small number of compounds and based on a limited number 535 of sampling events not representative of the annual average MEC which the PEC was designed to predict. In this way, we present novel findings that indicate when an annual 536 537 average MECs is calculated, less hydrologically complex river systems where 538 pharmaceutical sources are limited, PECs characterise annual exposure within a factor 2 for 41% of compounds in this study (average factor 2.8), with no factor greater than 11. 539 However paracetamol is an exception (underestimated by a factor of 73); the usage 540 estimate did not incorporate over-the-counter contributions therefore underestimates 541 were not unexpected (Burns et al., 2017). Conversely, the results from the River Ouse 542 543 indicate that major limitations are associated with this predictive approach. All ratios

544 were off by a factor of at least 7 (average 27) and up to 139, which according to studies 545 characterising the PEC/MEC, is outside an acceptable range (Verlicchi et al., 2014).

546 As the simple exposure model is routinely used for regulatory environmental risk 547 assessment (ERA) of new pharmaceuticals, our findings have important regulatory 548 implications. The predictions of exposure, currently being used to assess new 549 compounds, are likely under- or over-estimating concentrations, depending on the type 550 of compound. The use of a spatially referenced 'down the drain' hydrological model such 551 as LF2000-WQX (Williams et al., 2012) or GREAT-ER (Feijtel et al., 1997) would likely 552 result in improved predictions, as these models have the capacity to incorporate inputs from upstream sources; this is appropriate, as many rivers in the region pass through 553 554 multiple urbanised areas and thus are subject to multiple WWTP inputs. In addition, the 555 hydrological aspect can incorporate contributions or dilutions from the confluence with other river systems. Work currently being performed in the iPiE project involves the 556 development of a spatially resolved model for European surface waters. The high-quality 557 558 monitoring data presented in this study will be used to help evaluate this model. Our work also shows that inputs from other sources, potentially septic effluent, can be very 559 important for some compounds at certain time of year. The consideration of these direct 560 561 inputs in the risk assessment process may therefore be warranted.

#### 562 **4.0 Conclusion**

A rapid determination HPLC-MS/MS method for 33 pharmaceuticals was validated and applied in a 12-month spatiotemporal pharmaceutical exposure campaign. WWTP removal efficiency was found to be similar between CAS and trickling filter technology for the target pharmaceuticals. Pharmaceutical concentrations in two contrasting river systems that run through the city of York, UK were found to vary significantly spatially and temporally, with the greatest variation observed for paracetamol in the River Foss, ranging from not detected to over 9822 ng/L. Temporal variations in concentration were

570 less frequently observed in the larger River Ouse, potentially due to the lower variability 571 in flow which could be an important driver behind pharmaceutical concentrations in the 572 study system. PEC/MEC ratios indicated that compounds in both rivers were generally 573 underestimated by commonly used simple predictive exposure algorithms. In total, 41% 574 of PEC/MEC ratios for the River Foss data were within a factor of 2, while for the River Ouse average ratios indicated predictions were off by a factor of 27. This analytical 575 576 method and extensive monitoring results will be instrumental in improving the 577 understanding of temporal pharmaceutical fate and occurrence in river systems.

578

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Compound	F1		F2	F2			F4		F5		
Compound	Range (Med)	%	Range (Med)	%	Range (Med)	%	Range (Med)	%	Range (Med)	%	
Amitriptyline	n.d.	0	n.d. – 25.7 (10.3)	92	1.2* – 12.2 (5.7)	100	n.d. – 11.2 (2.6)	83	n.d. – 6.4 (2.0*)	75	
Atenolol	n.d.	0	18.9 – 100 (55.4)	100	12.3* – 98.2 (43.6)	100	13.7* – 97.8 (34.8)	100	10.1* – 67.0 (21.8)	100	
Carbamazepine	n.d. – 11.8 (4.5)	67	19.0 –195 (45.2)	100	8.7 – 194 (66.0)	100	12.5 – 175 (61.6)	100	11.4 – 193 (36.8)	100	
Cimetidine	n.d. – 49.6 (19.8)	83	n.d. – 44.0 (19.9)	92	3.0* - 40.5 (10.6)	100	2.1* - 16.9 (7.3*)	100	n.d. – 11.8 (7.2*)	67	
Citalopram	n.d.	0	5.0 – 71.4 (15.4)	100	3.8* - 31.0 (15.3)	100	3.1* - 13.5 (7.8)	100	n.d. – 11.4 (5.9)	83	
Codeine	n.d. – 10.8 (5.9*)	83	8.0 – 101 (59.2)	100	11.5 – 84.2 (57.3)	100	12.9 – 97.7 (44.0)	100	12.0 – 64.7 (29.1)	100	
Desvenlafaxine	n.d. – 55.8 (16.8)	83	25.8 – 268 (70.0)	100	4.6* - 195 (86.2)	100	11.7 – 170 (77.3)	100	8.5* - 96.4 (44.5)	100	
Diazepam	n.d.	0	n.d. – 1.6* (n.d.)	8.3	n.d 1.6* (n.d.)	8.3	n.d 1.8* (n.d.)	8.3	n.d 2.3* (n.d.)	8.3	
Diltiazem	n.d. – 4.1 (1.2*)	75	4.7 – 48.7 (16.4)	100	4.7 – 36.0 (14.5)	100	4.4 – 25.0 (10.6)	100	n.d. – 12.7 (5.8)	92	
Diphenhydramine	n.d.	0	n.d12.7 (9.5)	67	n.d. – 3.8 (n.d.)	25	n.d. – 1.6* (n.d.)	17	n.d. – 3.4 (n.d.)	8.3	
Erythromycin	n.d. – 34.5 (20.2*)	58	26.8 – 242 (90.0)	100	15.0* - 263 (88.8)	100	18.8* - 142 (80.5)	100	14.4 – 116 (45.9)	100	
Fexofenadine <sup>1</sup>	n.d. – 104 (24.9)	83	43.8 – 1144 (177)	100	17.2 – 956 (253)	100	27.5 – 638 (166)	100	26.4 – 268 (92.5)	100	
Gabapentin	17.4* – 229 (82.7)	100	476 – 1429 (789)	100	260 – 1445 (843)	100	404 – 1183 (768)	100	223 – 1341 (544)	100	
Hydrocodone	n.d. – 5.7 (n.d.)	43	11.2 – 91.8 (21.6)	100	6.4 – 60.3 (25.0)	100	6.8 – 43.5 (20.6)	100	5.2 – 22.2 (11.1)	100	
Lidocaine	n.d. – 3.9 (2.6*)	58	4.6 - 40.4 (8.2)	100	1.7* - 39.7 (11.8)	100	3.1 – 36.9 (10.4)	100	n.d. – 16.0 (6.1)	92	
Loratadine	n.d.	0	n.d.	0	n.d. – 6.46 (n.d.)	8.3	n.d.	0	n.d.	0	
Metformin	45.2 – 291 (121)	100	246 -1783 (856)	100	266 – 2339 (1117)	100	340 – 2595 (888)	100	263 – 1750 (664)	100	

**Table 1**. Summary results (ng/L) for the River Foss from the January to December 2016 monitoring campaign. The annual median, concentration range and frequency of detection for each sampling site are reported.

Compound	F1		F2	F2			F4		F5	
Compound	Range (Med)	%	Range (Med)	%	Range (Med)	%	Range (Med)	%	Range (Med)	%
Norethisterone	n.d.	0	n.d. – 7.4* (n.d.)	8.3	n.d.	0	n.d.	0	n.d.	0
Oseltamivir	n.d.	0	n.d. – 8.8* (n.d)	8.3	n.d.	0	n.d.	0	n.d.	0
Oxazepam	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0
Paracetamol	n.d. – 119 (60.0)	67	14.3* - 749 (74.4)	100	n.d. – 9822 (97.2)	92	32.0 – 9676 (209)	100	25.0 – 5445 (180)	100
Propranolol	n.d.	0	n.d. – 64.9 (17.8)	92	n.d. – 29.9 (20.1)	92	n.d. – 20.6 (10.0*)	92	n.d. – 18.3 (10.4*)	50
Raloxifene	n.d.	0	n.d.	0	n.d7.2*	8.3	n.d. – 7.2*	8.3	n.d.	0
Ranitidine	n.d. – 10.8* (n.d.)	17	n.d. – 69.6 (53.4)	83	6.6* – 74.0 (27.9)	100	n.d. – 60.6 (22.2)	92	n.d. – 30.0 (13.6*)	92
Sertraline	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d 21.2 (n.d)	8.3
Sitagliptin	n.d.	0	16.5 – 121 (35.2)	100	9.3* - 103 (46.5)	100	15.2 – 85.7 (36.9)	100	12.2 <sup>*</sup> – 33.9 (19.5)	100
Sulfamethoxazole	n.d.	0	n.d. – 10.2* (n.d.)	33	n.d. – 33.0 (n.d.)	50	n.d. – 27.5 (n.d.	42	n.d. – 18.1* (n.d.)	17
Temazepam	n.d.	0	n.d. – 38.2 (12.1)	67	n.d. – 25.0 (16.7)	75	n.d. – 27.8 (15.9)	67	n.d. – 12.6 (7.1*)	58
Tramadol	n.d. – 48.1 (31.2)	75	54.4 – 650 (117)	100	21.0 – 456 (177)	100	34.0 - 368 (169)	100	29.2 – 201 (84.7)	100
Triamterene	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0
Trimethoprim	n.d. – 9.8 (2.5*)	75	13.2 - 76.0 (30.3)	100	10.1- 60.3 (26.4)	100	15.2 – 49.4 (19.8)	100	5.3 – 38.0 (13.8)	100
Venlafaxine	n.d. – 4.3 (2.2*)	42	9.2 – 102 (16.2)	100	2.4* - 82.6 (20.6)	100	5.9 – 37.9 (17.6)	100	2.3* -17.8 (9.2)	100
Verapamil	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0

**Table 1**. Summary results for the River Foss from the January to December 2016 monitoring campaign. The annual median, concentration range and frequency of detection for each sampling site are reported.

<sup>\*</sup>Below LOQ

<sup>1</sup> data for 11 months only available (April 2016 missing).

n.d. No detect

(Med) Median

% Detection frequency (100% = 12 months)

**Table 2.** Summary results (ng/L) for the River Ouse from the January to December 2016 monitoring campaign. The annual median, concentration range and frequency detection for each sampling site are reported.

	01		O2		O3	O3			O5		O6	O6	
Compound	Range (med)	%	Range (med)	%	Range (med)	%	Range (med)	%	Range (med)	%	Range (med)	%	
Amitriptyline	n.d.	0	n.d.	0	n.d. – 2.7 (n.d.)	17	n.d1.2* (n.d.)	17	n.d. – 1.5* (n.d.)	8	n.d2.5 (n.d.)	17	
Atenolol	n.d.	0	n.d. – 22.0 (11.1*)	58	n.d. – 19.5 (10.7*)	67	n.d. – 16.9* (10.2*)	75	n.d. – 20.4 (10.4*)	67	n.d. – 18.8 (13.6*)	92	
Carbamazepine	1.0* – 14.0 (5.8)	100	1.1* - 34.8 (9.2)	100	1.4* - 54.4 (19.2)	100	1.1* - 31.4 (12.1)	100	1.7* - 33.9 (15.0)	100	7.9 – 48.0 (23.4)	100	
Cimetidine	n.d. – 2.3* (n.d.)	8	n.d. – 2.4* (n.d.)	8	n.d 5.7* (n.d.)	33	n.d. – 2.9* (n.d.)	17	n.d.	0	n.d. – 3.7 (n.d.)	42	
Citalopram	n.d 3.3* (n.d.)	8	n.d. – 3.7* (n.d.)	33	n.d. – 7.0 (4.0*)	75	n.d. – 3.2* (n.d.)	50	n.d. – 4.0* (2.2*)	67	n.d. – 7.2 (4.8)	83	
Codeine	n.d. – 13.5 (10.5*)	92	3.3 – 17.1 (10.7)	100	3.0* – 20.5 (14.3)	100	3.5* – 17.5 (13.8)	100	4.5* – 17.4 (14.9)	100	6.4* - 17.8 (8.8)	100	
Desvenlafaxine	n.d. – 14.8 (n.d.)	50	n.d. – 27.5 (11.3)	75	n.d. – 46.8 (21.5)	83	n.d31.0 (14.2)	83	n.d. – 28.8 (15.2)	75	12.3 – 40.1 (26.8)	100	
Diazepam	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	
Diltiazem	n.d. – 1.6* (n.d.)	25	n.d. – 2.5 (n.d.)	50	n.d. – 8.0 (3.6)	92	n.d. – 6.4 (1.8*)	67	n.d. – 3.7 (1.8*)	75	n.d. – 4.3 (3.7)	92	
Diphenhydramine	n.d.	0	n.d. – 1.7* (n.d.)	8	n.d 2.9 (n.d.)	25	n.d.	0	n.d 4.8 (n.d.)	8	n.d 2.2* (n.d.)	8	
Erythromycin	n.d.	0	n.d. – 17.3* (n.d.)	33	n.d. – 31.1 (21.3*)	92	n.d. – 20.3* (15.3*)	67	n.d. – 21.7* (n.d.)	50	n.d. – 33.9 (21.3*)	83	
Fexofenadine <sup>1</sup>	n.d. – 41.7 (17.9)	83	n.d. – 48.7 (24.1)	83	n.d. – 77.8 (46.1)	92	n.d. – 68.2 (25.8)	83	n.d. – 44.0 (29.2)	92	7.4* – 98.5 (33.4)	100	
Gabapentin	28.1* -242 (130)	100	39.4 – 351 (191)	100	24.5* - 429 (230)	100	30.0* - 369 (202)	100	33.8* - 364 (192)	100	39.5 – 450 (208)	100	
Hydrocodone	n.d. – 2.9 (n.d.)	50	n.d. – 5.7 (3.6)	83	n.d. – 14.9 (7.8)	92	n.d. – 8.0 (4.0)	92	n.d. – 6.9 (4.0)	92	2.2 – 10.7 (6.0)	100	
Lidocaine	n.d. – 4.1 (n.d.)	50	n.d. – 5.0 (2.7*)	83	n.d. – 6.5 (3.7)	92	n.d. – 5.4 (2.8)	83	n.d. – 5.6 (3.1)	83	1.6* - 8.8 (4.1)	100	
Loratadine	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	
Metformin	52.5 – 323 (180)	100	63.4 – 431 (223)	100	60.6 – 422 (237)	100	60.2 – 422 (237)	100	73.6 – 445 (233)	100	142 – 483 (276)	100	
Norethisterone	n.d.	0	n.d7.7 (n.d.)	8	n.d.	0	n.d.	0	n.d.	0	n.d.	0	
Oseltamivir	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	

Table 2. Summary results (ng/L) for the River Ouse from the January to December 2016 monitoring campaign. The annual median, concentration range and frequency detection for each sampling site are reported.

	O1		O2	O2			O4		O5		O6	
Compound	Range (med)	%	Range (med)	%	Range (med)	%	Range (med)	%	Range (med)	%	Range (med)	%
Oxazepam	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0
Paracetamol	22.3* – 191 (46.4)	100	15.4* - 202 (51.7)	100	16.8* – 186 (54.5)	100	20.1* – 186 (54.3)	100	22.7 – 369 (77.6)	100	21.2 – 226 (66.9)	100
Propranolol	n.d.	0	n.d.	0	n.d. – 8.3* (n.d.)	33	n.d.	0	n.d.	0	n.d. – 7.6* (n.d.)	8
Raloxifene	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0
Ranitidine	n.d10.3* (n.d.)	25	n.d. – 10.5* (n.d.)	25	n.d. – 30.6 (15.1*)	75	n.d 13.3* (n.d.)	42	n.d. – 12.0* (n.d.)	25	n.d. – 15.5* (9.2*)	75
Sertraline	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0
Sitagliptin	n.d. – 10.7 (n.d.)	33	n.d. – 16.2 (9.3*)	75	n.d. – 32.5 (15.0)	92	n.d. – 16.9 (12.0*)	83	n.d. – 15.8 (10.4*)	83	n.d. – 26.5 (18.2)	92
Sulfamethoxazole	n.d.	0	n.d.	0	n.d.	0	n.d. Í	0	n.d.	0	n.d.	0
Temazepam	n.d.	0	n.d.	0	n.d. – 7.2* (n.d.)	8	n.d.	0	n.d. – 4.4* (n.d.)	8	n.d. – 4.7* (n.d.)	8
Tramadol	n.d. – 27.0 (19.6)	83	3.9* - 39.9 (19.8)	100	n.d. – 57.2 (34.6)	92	n.d. – 44.8 (28.9)	92	n.d. – 47.4 (27.4)	92	20.7 – 52.4 (40.5)	100
Triamterene	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0
Tramadol	n.d. – 27.0 (19.6)	83	3.9* - 39.9 (19.8)	100	n.d. – 57.2 (34.6)	92	n.d. – 44.8 (28.9)	92	n.d. – 47.4 (27.4)	92	20.7 – 52.4 (40.5)	100
Triamterene	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0	n.d.	0
Trimethonrim	n.d. – 19.0	92	2.0* - 8.9	100	2.8* - 19.3	100	n.d. – 11.1	92	2.3* - 12.1	100	7.3 – 22.9	100
minetiopiim	(2.7)		(5.3)		(12.4)		(5.4)		(5.5)		(14.2)	
Venlafaxine	n.d. – 2.6*	42	n.d. – 5.2	75	n.d. – 8.5*	83	n.d. – 4.3	75	n.d. – 5.0	75	n.d. – 8.2	83
	(n.d.)	•	(2.6*)		(4.9)	•	(2.9*)		(3.1)	•	(4.5)	•
Verapamil	n.d.	0	n.d.		n.d.	0	n.d.	0	n.d.	0	n.d.	0

\* Below LOQ

<sup>1</sup> data for 11 months only available (April 2016 missing).

n.d. No detect

(Med) Median % Detection frequency (100% = 12 months)



**Figure 1**. Map of 11 sampling sites within the sampling network. River flows recorded from a gauge in each river (orange triangle) from each sampling day (m<sup>3</sup>/s) are pictured top left. WWTPs that serve the city (3) are represented by the red rectangles. Sites F1-F5 are along the smaller River Foss, while sites O1-O6 are along the larger River Ouse.



Figure 2. A) Routine matrix spikes run alongside environmental samples during the 12-month monitoring campaign in WWTP influent, effluent, surface water and reagent water. The dotted lines represent the 70 – 120% acceptable recovery range. B) %RSD of matrix spike replicates. An RSD below 20% is desirable (depicted with dotted line). The median, 25th and 75th quartiles are presented while the whiskers represent the 10th to 90th percentile, compounds outside this range are depicted with an X.



Figure 3. Estimated % removal in WWTP A (trickling filter), WWTP B (carbon activated sludge). Hydrocodone not shown, estimated removal in WWTP A -307% and in WWTP B -597%. Matrix recovery was outside the 70 -120% desired range is identified for (a) influent and (b) effluent.



**Figure 4.** Total pharmaceutical concentration (summed) of all detected analytes at each sampling site from each month during 2016 in the River Foss. Sampling locations (blue circles) in relation to Environment Agency Flow gauges (orange triangles) are depicted along the river.



**Figure 5.** Total pharmaceutical concentration (summed) of all detected analytes at each sampling site from each month during 2016 along the River Ouse. Sampling locations (blue circles) in relation to Environment Agency Flow gauges (orange triangles) are depicted along the river.



Figure 6. Annual median concentration from all sampled sites in (A) the River Foss and (B) River Ouse. Pairwise t-tests were conducted between neighbouring sites and significant differences are denoted by the corresponding number. Sites F1-F2, O1-O2 =1; F2-F3, O2-O3 =2; F3-F4, O3-O4 =3; F4-F5, O4-O5 =4; O5-O6 =5.



Figure 7. Median seasonal concentration from sites F3-F4 in the River Foss (A) and O3-O4 in the River Ouse (B) for select pharmaceuticals. Temporal variations were tested using Friedman's Test and results are reported for each compound where a significant result was found, p<0.05 (\*), p<0.01 (\*\*), p<0.0005 (\*\*\*), p<0.0001 (\*\*\*).</p>



Figure 8. The annual average PEC/MEC ratios are plotted for the River Foss (open circles) and the River Ouse (closed circles). PECs were calculated for each river based on experimental WWTP removals and the average flow from sampling days. PEC/MEC ratios were calculated for site F2-F5 and O3-O6 and averaged, error bars represent the standard deviation.

# Highlights

- 11 sites from two nested river systems sampled monthly for one year.
- Seasonal and spatial variation due to flow, usage and compound stability.
- Removal efficiency estimated for 24 pharmaceuticals in two WWTPs.
- Disagreement between measured concentrations and exposure model predictions.