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## Experimental Investigation and Modeling of the Transformation of Illicit Drugs in a Pilot-Scale Sewer System

Jiaying Li

*University of Queensland*

Jianfa Gao

*University of Queensland*

Phong K. Thai

*University of Queensland*

Adam Shypanski

*University of Queensland*

Ludwika Nieradzik

*University of Queensland*

*See next page for additional authors*

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# Experimental Investigation and Modeling of the Transformation of Illicit Drugs in a Pilot-Scale Sewer System

## Abstract

In-sewer stability of illicit drug biomarkers has been evaluated by several reactor-based studies, but less has been done in sewer pipes. Experiments conducted in sewer pipes have advantages over lab-scale reactors in providing more realistic biomarker stability due to the flow and biological dynamics. This study assessed the transportation and transformation of seven illicit drug biomarker compounds in a pilot-scale rising main and a gravity sewer pipe. Biomarkers presented diverse stability patterns in the pilot sewers, based on which a drug transformation model was calibrated. This model was subsequently validated using transformation data sets from the literature, aiming to demonstrate the predictability of the pilot-based transformation coefficients under varying sewer conditions. Furthermore, transformation coefficients for five investigated biomarkers were generated from four studies, and their prediction capabilities under the pilot-sewer conditions were jointly assessed using performance statistics. The transformation model was successful in simulating the in-sewer stability for most illicit drugs. However, further study is required to delineate the sources and pathways for those compounds with potential formations to be simulated in the transformation model. Overall, the transformation model calibrated using the pilot-sewer data is a credible tool for the application of wastewater-based epidemiology.

## Keywords

transformation, experimental, modeling, sewer, system, investigation, pilot-scale, drugs, illicit

## Disciplines

Engineering | Science and Technology Studies

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## Authors

Jiaying Li, Jianfa Gao, Phong K. Thai, Adam Shypanski, Ludwika Nieradzic, Jochen F. Mueller, Zhiguo Yuan, and Guangming Jiang

1 Experimental investigation and modelling of the  
2 transformation of illicit drugs in a pilot-scale sewer  
3 system

4 *Jiaying Li<sup>a</sup>, Jianfa Gao<sup>b</sup>, Phong K. Thai<sup>b</sup>, Adam Shypanski<sup>a</sup>, Ludwika Nieradzik<sup>a</sup>, Jochen F.*  
5 *Mueller<sup>b</sup>, Zhiguo Yuan<sup>a</sup>, Guangming Jiang<sup>a,c,\*</sup>*

6 <sup>a</sup>Advanced Water Management Centre, The University of Queensland, St Lucia, QLD 4072,  
7 Australia

8 <sup>b</sup>Queensland Alliance for Environmental Health Sciences, The University of Queensland,  
9 Brisbane, QLD 4102, Australia

10 <sup>c</sup>School of Civil, Mining and Environmental Engineering, University of Wollongong,  
11 Wollongong, NSW 2522, Australia

12

13 **KEYWORDS**

14 Wastewater-based epidemiology, illicit drugs, biotransformation, pilot sewer system, model  
15 validation

16 **ABSTRACT**

17 In-sewer stability of illicit drug biomarkers has been evaluated by several reactor-based studies  
18 but less has been done in sewer pipes. Experiments conducted in sewer pipes have advantages

19 over lab-scale reactors in providing more realistic biomarker stability due to the flow and  
20 biological dynamics. This study assessed the transportation and transformation of seven illicit  
21 drug biomarker compounds in a pilot-scale rising main and a gravity sewer pipe. Biomarkers  
22 presented diverse stability patterns in the pilot sewers, based on which a drug transformation  
23 model was calibrated. This model was subsequently validated using transformation datasets  
24 from literature, aiming to demonstrate the predictability of the pilot-based transformation  
25 coefficients under varying sewer conditions. Furthermore, transformation coefficients for five  
26 investigated biomarkers were generated from four studies and their prediction capabilities  
27 under the pilot sewer conditions were jointly assessed using performance statistics. The  
28 transformation model was successful in simulating the in-sewer stability for most illicit drugs.  
29 However, further study is required to delineate the sources and pathways for those compounds  
30 with potential formations to be simulated in the transformation model. Overall, the  
31 transformation model calibrated using the pilot-sewer data is a credible tool for the application  
32 of wastewater-based epidemiology.

### 33 INTRODUCTION

34 Wastewater-based epidemiology (WBE) has been developed rapidly over the last decade in a  
35 bid to achieve objective and timely assessment of community health and consumption  
36 behaviours via analysing trace levels of substances (termed as biomarkers) in wastewater,  
37 including illicit drugs, pharmaceuticals and new psychoactive substances.<sup>1</sup> Back-estimating  
38 the catchment-wide usage of illicit drugs is an emerging area of WBE, which is demonstrated  
39 to be a useful complementary tool to conventional drug monitoring approaches.<sup>2-5</sup> In order to  
40 improve the accuracy of back-estimation, researches have been widely conducted to address  
41 the uncertainties associated with sampling method and chemical analysis,<sup>5-9</sup> while a  
42 comprehensive understanding of biomarkers stability in real sewers is still ongoing.<sup>10-17</sup>

43 Biomarkers are subjected to physiochemical and biological processes during their transport in  
44 rising main and gravity sewer pipes, where the hydraulic retention time (HRT) may last for  
45 hours.<sup>18</sup> Neglecting the biomarkers transformation (e.g., the degradation or formation in  
46 sewers) will lead to an under- or over-estimation of drug consumption in a catchment.<sup>4, 10, 12,</sup>  
47 <sup>13, 15, 19</sup> This uncertainty varies depending on the stability of biomarkers and the  
48 characteristics of sewer systems such as HRT distributions, which was suggested to be  
49 negligible for the stable biomarkers<sup>8</sup> but significantly increased for the unstable biomarkers  
50 with >40% median mass losses in the catchments.<sup>15</sup>

51 Depending on the different experimental scales and conditions utilized, research on  
52 biomarker stability can be divided into four categories: 1) *in-water study* using clean/sterile  
53 bottles or other containers where abiotic processes such as chemical hydrolysis occur in  
54 water;<sup>14, 20, 21</sup> 2) *in-wastewater study* that is conducted in raw wastewater where suspended  
55 biomass and certain microbial activities contribute to the transformation process;<sup>12, 14, 21-25</sup> 3)  
56 *sewer reactor study* employing lab reactors with intact/suspended biofilms or activated  
57 sludge to mimic the biologically active sewer environments;<sup>10-12, 14, 16, 19</sup> 4) *real sewer pipe*  
58 *study* using the sewer pipes with the same (or similar) size and operational conditions as the  
59 real sewer networks.<sup>10, 15, 17, 26, 27</sup>

60 Among the *real sewer pipe studies*, two of them assessed the change of biomarkers using 24-  
61 h composite samples, however, with limited understanding of the concurrent in-pipe  
62 hydraulics or biological activities.<sup>15, 26</sup> Two other studies spiked biomarkers in a real rising  
63 main pipe and evaluated their variations from the pipe upstream to a downstream sampling  
64 point, coupled with the measurements of flow dynamics and biological activities in the pipe.  
65 However, due to the poor accessibility of real sewer pipes and the narrow HRT windows, the  
66 obtained data points were insufficient for kinetics evaluation.<sup>10, 27</sup> In comparison to the

67 inherent limitations of static lab-reactors and complex real sewers<sup>28</sup>, a pilot-scale sewer can  
68 be more beneficial for the study of biomarker stability in sewer pipes by providing multiple  
69 sampling points, online monitoring, controllable flow, and other environmental factors. This  
70 has been demonstrated in a recent *real sewer pipe study* investigating the fate of  
71 pharmaceutical biomarkers.<sup>17</sup>

72 The temporal transformation data obtained in pilot-scale sewers can be very valuable for the  
73 modelling of biomarker stability. Transformation modelling is a useful approach with which  
74 to utilize information on biomarker stability in the application of WBE in reality. So far, three  
75 *sewer reactor studies* investigated the transformation modelling of illicit drugs,<sup>10, 14, 16</sup> but  
76 only one study validated the estimated transformation coefficients in a real rising main pipe.<sup>10</sup>  
77 It is thus imperative to calibrate the illicit drug transformation model using dynamic data  
78 from sewer pipes (e.g. pilot-scale sewers) instead of lab reactors. More importantly, the  
79 transformation coefficients derived from different studies need to be systematically evaluated  
80 for their transferability across diverse sewer conditions.

81 The first objective of this study is to measure and model the stability of illicit drug  
82 biomarkers in the pilot-scale sewer pipes. Experiments were conducted in a pilot-scale sewer  
83 system, including a rising main pipe and a gravity sewer pipe with online monitoring and  
84 control system. A drug transformation model was calibrated with the pilot-sewer data and  
85 was subsequently validated using literature data under varying sewer conditions. The second  
86 objective is to systematically evaluate the transformation coefficients generated by previous  
87 studies, through the comparison of the prediction capabilities using performance statistics.  
88 Collectively speaking, this work intends to advance WBE through not only providing  
89 valuable data on illicit drug transformation in sewer pipes, but also enhancing the  
90 generalizability and applicability of the transformation model to the application of WBE.

## 91 MATERIALS AND METHODS

### 92 **Compounds for Investigation**

93 This study investigated the parent compounds of major illicit drug biomarkers by spiking  
94 them into the pilot-scale sewer pipes, including cocaine (COC), ketamine (KET), 3,4-  
95 methylenedioxyamphetamine (MDMA), morphine (MOR), and methadone (MTD).  
96 According to the analysis results of wastewater samples at the experiment site, the spiking  
97 concentrations (2.5-8 ppb) for most biomarkers were higher than their native residues in raw  
98 wastewater (<0.5 ppb). The native methamphetamine (METH) was investigated without  
99 being spiked because of its relatively high background concentrations (around 1 ppb). The  
100 native benzoylecgonine (BE) was evaluated as a specific metabolite of COC without being  
101 spiked. Other metabolites of the parent compounds were not spiked in separate tests. The  
102 flow tracer rhodamine was added into the spiking mixture solution. Rhodamine signal was  
103 measured by a portable Cyclops<sup>®</sup>-7 Submersible Rhodamine Sensor coupled with a Cyclops<sup>®</sup>  
104 Explorer.

### 105 **Tests in the pilot-scale sewer system**

106 Experiments were conducted in a pilot sewer system located at the Luggage Point  
107 Wastewater Treatment Plant, Queensland, Australia. This study employed two 300-m long  
108 sewer pipes, one rising main and one gravity pipe, on a controllable platform equipped with  
109 programmable logic controller, pumps, meters, sensors, etc.<sup>17, 29</sup> The experimental procedures  
110 in this study were the same as reported previously.<sup>17</sup> The layout and parameters of the sewer  
111 system are illustrated in Supplementary Information S1.1.

112 The *Rising main pipe* is completely filled with wastewater and anaerobic biofilm were  
113 cultivates on the pipe's inner surface with a thickness of 1~2 mm.<sup>29</sup> The internal diameter is

114 100 mm, leading to a biofilm-area-to-wastewater-volume (A/V) ratio of 40 m<sup>-1</sup>. This rising  
115 main pipe is constructed to spiral up from the ground layer (inlet) to the top layer (outlet) and  
116 wastewater is driven by pumping events in a plug-flow regime. As shown in Figure S1.1,  
117 eight sampling ports are distributed along the sewer pipe, i.e. at 0, 15, 45, 75, 105, 135, 195,  
118 and 240 m from the inlet for port #1 to #8, respectively.

119 Triplicate batch tests were carried out over 3 consecutive days (Day 1, 2 and 3) in the rising  
120 main. To achieve a typical hydraulic condition, the main pump was turned on for 1 min every  
121 hour with a flow rate of 236 L min<sup>-1</sup>, producing a wastewater slug of 30 m and an intermittent  
122 shear stress of 0.6 N m<sup>-2</sup> in the pipe. With the first pumping event of each test, a mixture  
123 solution of biomarkers and rhodamine was spiked into influent using an external peristaltic  
124 pump, resulting in the first and the only spiked wastewater slug at time 0 (t<sub>0</sub>). This spiked  
125 wastewater slug was pushed 30 m downstream by the subsequent non-spiked wastewater  
126 slugs at every following pumping event and arrived at the final sampling port after 7 h of  
127 HRT in the pipe. Through matching the length of every wastewater slug with the distance  
128 between two sampling ports (i.e. 30 m), the central area of this spiked slug could be captured  
129 at #1 to #8 sampling ports in sequence after each hourly pumping event. Meanwhile, during  
130 every 1-h pump-off period, samples of the spiked wastewater were collected at 15 min  
131 intervals through the sampling port where the spiked slug was located. Concurrently with the  
132 transportation of the spiked wastewater slug, a rhodamine sensor was connected to #1 to #8  
133 sampling ports for the monitoring of the rhodamine concentration.

134 The *Gravity sewer pipe* comprises both water and air phases, between which gas transfer  
135 leads to 1~4 mg L<sup>-1</sup> dissolved oxygen in the bulk liquid phase. A removable section of pipe  
136 showed the prevailing existence of sediments at the bottom. This gravity sewer pipe has an  
137 internal diameter of 225 mm and is constructed to spiral down from the top layer (inlet) to the



138 ground layer (outlet) with a slope of 0.56%, where wastewater flow is driven by gravitation.  
139 The traveling time of wastewater from the pipe inlet to the outlet ranged around 8~10 min.  
140 Moreover, this pilot gravity sewer pipe has the unique capability of allowing recirculation.  
141 Under the recirculation mode, wastewater effluent at the pipe outlet is collected in a  
142 recirculation tank and re-directed to the feeding tank by a recirculation pump. The  
143 recirculation mode enables wastewater to flow in the gravity sewer pipe as long as required.  
144 The average flow velocity was  $0.38 \text{ m s}^{-1}$ ; the in-pipe water depth was around 5 cm; and the  
145 average A/V ratio was approximately  $27.5 \text{ m}^{-1}$  (S1.2 and Figure S1.2). The average shear  
146 stress under this flow condition was  $0.5 \text{ N m}^{-2}$ . A sampling port is installed near the outlet,  
147 allowing access to wastewater in the pipe (Figure S1.1).

148 Triplicate tests were conducted over 3 consecutive days (Day 4, 5 and 6) in the gravity sewer  
149 pipe under recirculation mode. At the beginning of each test, a mixture solution of  
150 biomarkers and rhodamine was directly spiked into the pipe and continuously mixed with the  
151 flowing wastewater. Meanwhile, a rhodamine sensor was connected to the sampling port for  
152 online readings during experimental periods. As shown in preliminary tests, the spiked  
153 rhodamine presented 2~3 signal peaks over the first few cycles of recirculation and then a  
154 sufficiently mixed stage appeared at 1.5 h after the spiking event as indicated by the  
155 consistent signal intensity. Therefore, for experiments in the recirculating gravity sewer,  $t_0$  of  
156 the non-spiked biomarker commenced at the beginning of each test. However,  $t_0$  of the  
157 spiked biomarkers was defined when a homogeneously mixed status was reached, in order to  
158 minimize the uncertainty of mixing on the evaluation of the spiked biomarkers stability. The  
159 HRTs of the recirculating wastewater in the gravity sewer pipe were 3~4 h, during which  
160 wastewater samples were collected at 15 min intervals.

161 The collected samples were pretreated on site for the analyses of biomarkers and wastewater  
162 parameters, including sulfur species, dissolved methane, volatile fatty acids (VFAs), soluble  
163 chemical oxygen demand (SCOD), total and volatile suspended solids (TSS and VSS) (for  
164 sample pre-treatment and analytical methods see S2). Temperature and pH of samples were  
165 measured on site using a portable pH/temperature meter (TPS Aqua-pH pH/Temp meter).

## 166 **Calibration of Drug Transformation Model**

167 As widely applied in previous studies,<sup>10, 14, 30-32</sup> a first-order kinetics is adopted for illicit drug  
168 transformation model in this work (eq 1):

$$C_t = C_0 \cdot e^{-\left(k_{ww} + k'_{bio} \cdot \frac{A}{V}\right) \cdot t} \quad (1)$$

169  $C_t$  is biomarker concentration ( $\mu\text{g L}^{-1}$ ) at time  $t$  (h) and  $C_0$  is the initial concentration ( $\mu\text{g L}^{-1}$ ).  
170  $k_{ww}$  ( $\text{h}^{-1}$ ) represents the processes in the bulk liquid wastewater, mainly chemical hydrolysis  
171 assuming that the sorption to suspended solids or biofilm is limited for investigated  
172 biomarkers<sup>5, 14, 16, 33</sup> and the biological activity of suspended solids is negligible compared to  
173 the sewer biofilms or sediments.<sup>34-36</sup> It is further postulated that  $k_{ww}$  remains the same under  
174 aerobic and anaerobic conditions, according to the findings of McCall *et al.*<sup>14</sup> and Ramin *et*  
175 *al.*<sup>16</sup> (about the minor impact of redox condition on abiotic transformation rates).  $k'_{bio}$  ( $\text{m h}^{-1}$ )  
176 includes  $k'_{bioa}$  and  $k'_{bioan}$ , representing the biofilm effect under aerobic and anaerobic  
177 condition with the normalization of A/V ratio, respectively. The effects of mass transfer  
178 limitation on  $k'_{bio}$  is considered to be negligible for illicit drug compounds under the  
179 experimental conditions as discussed in S1.2. In conclusion, the overall in-sewer  
180 transformation rate ( $\text{h}^{-1}$ ) depends on  $k_{ww}$  and the  $k'_{bio}$  coupled with a specific A/V ratio.  
181 Consequently, the percentage contributions from wastewater and biofilm processes to the  
182 overall transformation of a biomarker are quantified by the ratio of  $k_{ww}$  against  $k'_{bio} \cdot \frac{A}{V}$ . This

183 study assumed that the mature sewer biofilms in the pilot sewers were under steady state with  
184 negligible biomass growth over the course of experiments.

185 To calibrate the transformation model, experimental datasets obtained from the pilot rising  
186 main pipe (Day 1 and 2) and the pilot gravity sewer (Day 4 and 5) pipe were used to estimate  
187  $k'_{bioan}$  and  $k'_{bioa}$ , respectively, through a Bayesian procedure described in Li *et al.*<sup>10</sup> The  
188 datasets of Day 3 (rising main) and Day 6 (gravity sewer) were reserved for the subsequent  
189 model validation. As shown in a previous field-scale study,<sup>10</sup> rhodamine signal was used to  
190 normalize the biomarker concentrations which effectively minimized the potential hydraulic  
191 uncertainty during wastewater transport. Therefore, for the spiked biomarkers in this pilot  
192 study,  $C_t$  is normalized by the ratio of the initial rhodamine signal against the signal at  $t$ .  $k_{ww}$   
193 is estimated using the data on biomarkers transformation in the bulk wastewater (Figure  
194 S3.1), which is collected from the experiments using the control sewer reactor without  
195 biofilms and from previous *in-wastewater studies*.<sup>10, 12</sup> The estimation of  $k'_{bio}$  is carried out in  
196 R (Version 3.2) which executes the Bayesian method in OpenBUGS  
197 (<http://www.openbugs.net>).

### 198 **Validation of the pilot-based model**

199 In order to evaluate the validity of the transformation coefficients ( $k$  values) estimated by the  
200 abovementioned pilot-based model across diverse sewer environments, experiment datasets  
201 obtained from the pilot rising main at Day 3 and the pilot gravity sewer at Day 6 together  
202 with the literature data collected from three independent sewer reactor studies<sup>10, 14, 16</sup> are used  
203 as the observation inputs representing different testing sewer conditions (datasets in Table  
204 S3.2). For each investigated biomarker, the pilot-based  $k$  values are used to generate  
205 predictions (mean with 95% confidence bounds) with the specific A/V ratios of the four  
206 studies (for descriptions of prediction scenarios see Table S3.2). Under each scenario,

207 rejection probability is determined by counting the percentage of experimental observations  
208 (i.e. the data on biomarker transformation) located outside the corresponding predictive  
209 region. The lower rejection probability reflects the higher validity of the pilot-based  $k$  values.

## 210 **Performance comparisons of different $k$ -value sets**

211 To date, several lab-scale studies have assessed the in-sewer stability of illicit drug  
212 biomarkers and estimated the transformation coefficients under their specific testing  
213 conditions.<sup>10, 14, 16</sup> Although these  $k$ -value sets demonstrated adequate prediction performance  
214 in the corresponding studies, their predictive abilities have neither been validated beyond  
215 their original testing conditions nor jointly compared under the realistic sewer conditions.  
216 Importantly, measurements of this pilot study are valuable to model validation, which were  
217 obtained from an enlarged pilot-scale system with typical operations and dynamic hydraulics  
218 of real sewer systems.

219 For this reason, multiple  $k$ -value sets ( $k_{ww}^{M_i}$  and  $k_{bio}^{M_i}$ ) are obtained from four different  
220 stability studies, i.e., this pilot study and three *sewer reactor studies* (defined as M1~M4 in  
221 Table S3.2), and their prediction capabilities are jointly compared under the conditions of the  
222 pilot-scale sewer system. Multiple prediction scenarios (mean with 95% confidence bounds)  
223 under the conditions of the pilot rising main and the gravity sewer are generated for each  
224 investigated biomarker (Table S3.2). Furthermore, in order to identify the  $k$ -value set with the  
225 highest agreement of prediction to the observations, performance statistics are computed for  
226 the predictive scenarios of M1~M4 under each condition, including a stochastic validation  
227 metric (the Bayes factor) and two accuracy measures (Pearson correlation coefficient  $r$  and  
228 variance explained by predictive models based on cross-validation ( $VEcv$ )) calculated using  
229 RStudio (Version 1.0.143).

230 *1) The Bayes Factor*

231 The Bayes Factor (BF) is a typical model validation metric, which quantitatively measures  
232 the agreement between predictions and measurements together with their internal  
233 uncertainties based on stochastic characteristics.<sup>37</sup> As explained by eq 2, the Bayes Factor  
234 quantifies the ratio of the probabilities of observations under null hypothesis  $H_0$  and  
235 alternative hypothesis  $H_1$ , respectively, at each validation site (for visual illustrations see  
236 Figure S3.2):

$$BF = \frac{P(\text{data}|H_0)}{P(\text{data}|H_1)} \quad (2)$$

237 where  $H_0$  is the null hypothesis representing the better match between the observations in  
238 pilot sewers and the predictions of a calibrated model, e.g. M1;  $H_1$  is the alternative  
239 hypothesis representing the better match between observations and the predictions of other  
240 competing models, e.g., M2~M4. For each biomarker under the pilot gravity sewer or the  
241 rising main condition, up to three log(BFs) can be computed depending on the available  $k$ -  
242 value sets in literature, i.e., log(BF12), log(BF13), and log(BF14). At a validation site,  
243 log(BF)>0 means  $H_0$  is true, indicating that predictions by the  $k$ -value set of M1 is favoured  
244 by the observed data. A larger absolute value of log(BF) suggests the higher preference to the  
245 pilot-based M1 when log(BF)>0, or to the reactor-based M2, M3, or M4 when log(BF)< 0.

246 *2) Accuracy of model predictions*

247 The Pearson correlation coefficient  $r$  is widely used to assess model predictions against  
248 observed data via quantifying the strength and direction of a linear relationship between the  
249 two variables (eq 3). When observations  $x$  are perfectly linearly related to model predictions  
250  $y$  (i.e.,  $y = \beta_0 + \beta_1 x$ ,  $\beta_0$  and  $\beta_1$  are coefficients),  $r$  is a suitable indicator of predictive  
251 accuracy:<sup>38</sup>

$$r = \frac{\sum_1^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_1^n (x_i - \bar{x})^2 (y_i - \bar{y})^2}} \quad (3)$$

252 where  $n$  is the number of observations;  $x_i$  is the observed value  $i$ ;  $\bar{x}$  is the mean of the  
 253 observed values;  $y_i$  is the predicted value  $i$ ;  $\bar{y}$  is the mean of the predicted values.

254 However, when  $x$  and  $y$  are not well correlated, i.e., when noticeable noise  $\varepsilon$  appears in  $y =$   
 255  $\beta_0 + \beta_1 x + \varepsilon$ ,  $r$  becomes potentially biased and  $VEcv$  (eq 4) is recommended as the correct  
 256 measure of predictive accuracy instead:<sup>38</sup>

$$VEcv = \left(1 - \frac{\sum_1^n (x_i - y_i)^2}{\sum_1^n (x_i - \bar{x})^2}\right) \times 100(\%) \quad (4)$$

## 257 RESULTS AND DISCUSSION

### 258 Wastewater Compositions and Biological Activities

259 The observations of wastewater parameters in this study were the same as reported in Gao *et*  
 260 *al.*,<sup>17</sup> which conducted experiments in the same setup. The sewer characteristics in the pilot  
 261 sewer system, in terms of wastewater parameters and biological activities, are also compared  
 262 to the literature data. It is found that the variations of wastewater parameters in the pilot  
 263 sewers (Figure S1.3) are similar to those in other sewer studies (Table S1.2).<sup>10, 39, 40</sup> For  
 264 instance, wastewater pH kept relatively stable over 8 h in the pilot rising main ( $6.99 \pm 0.11$ ),  
 265 while pH in the pilot gravity sewer increased by  $\sim 0.25$  units during the first 2 h, likely due to  
 266 the  $\text{CO}_2$  and  $\text{H}_2\text{S}$  stripping. Wastewater temperature remained consistently stable in the pilot  
 267 sewers ( $22.9 \pm 0.6$  °C). TSS was higher in the pilot gravity sewer (500-800 mg L<sup>-1</sup>) than the  
 268 pilot rising main (200-410 mg L<sup>-1</sup>) due to resuspension of sediments. The interday deviations  
 269 of TSS were attributed to the daily variation of real wastewater, while the intraday changes of  
 270 TSS were relatively insignificant over the experimental period.

271 Based on previous studies, sulfate reducing bacteria and methanogenic archaea are the  
272 primary microorganisms responsible for not only the carbon/sulfur transformation, but also  
273 the biomarker stability in biofilms.<sup>14, 18, 41, 42</sup> In this work, sulfide and methane production  
274 rates are used as the major indicators of biological activities in the pilot sewers, which are  
275 comparable to the literature data (Table S1.2). The simultaneously increasing pattern of  
276 sulfide and methane in the pilot rising main (Figure S1.3) was also similar to that in real  
277 sewers.<sup>43-46</sup> In contrast, sulfide or methane production was detected to be negligible or even  
278 negative in the pilot gravity sewer, which could be explained by the faster sulfide oxidation  
279 than the concurrent sulfate reduction and/or the emission of sulfide and methane gas into the  
280 air phase.

281 VFAs and SCOD are the major substrates for heterotrophic bacteria and methanogens, which  
282 present varying consumption rates in different sewer environments (Table S1.2). Primarily  
283 being produced by fermentation process under anaerobic condition,<sup>39, 47</sup> the consumption rate  
284 of VFAs in the pilot rising main was lower than that in the gravity sewer (Figure S1.3). The  
285 consumption rate of SCOD was higher than VFAs in the pilot rising main, while in the pilot  
286 gravity sewer, the concurrent consumption of SCOD and VFAs was generally similar.  
287 Overall, this unique pilot sewer system has an environmental condition representative of real  
288 sewers.

### 289 **Biomarkers Stability in the Pilot Sewer System**

290 The online monitoring of rhodamine showed the dynamic flow patterns in the pilot sewer  
291 pipes during experiment periods (Figure 1). The rhodamine profiles shared the same results  
292 of Gao *et al.*<sup>17</sup> In the pilot rising main, the staged profiles of rhodamine revealed the plug-  
293 flow regime as a response to the intermittent pumping events. In the pilot gravity sewer under

294 recirculation mode, rhodamine profiles reflected the continuous mixing and indicated the  
295 commencing of a sufficiently mixed stage at 1.5 h after the spiking event.

296 Various patterns were found for the investigated biomarkers in the pilot sewer system (Figure  
297 1). Over the consecutive tests in the pilot gravity sewer or the rising main, the interday  
298 divergence of the transformations for each biomarker was relatively limited as indicated by  
299 the low standard deviations. COC, MDMA, and MTD exhibited decreasing trends during the  
300 testing periods in both the pilot rising main and the gravity sewer, where MTD was observed  
301 to have the most rapid degradation with ~25% loss in 3 h, followed by ~20% loss of COC  
302 and MDMA. During the longer HRT periods in the pilot rising main, these biomarkers still  
303 presented similar degradation trends and overall 35~40% losses were observed after 8 h. By  
304 contrast, KET and METH showed relatively good stability with <20% losses in both pipes. In  
305 addition, formations of BE and MOR were observed over the experimental periods, which  
306 were the combined results of multiple concurrent in-sewer processes, including the  
307 transformation of biomarker itself, the back-transformation of parent compounds and/or the  
308 potential deconjugation of glucuronides in the raw wastewater.<sup>11, 12</sup> Data on the  
309 transformations of MOR and MTD in the pilot sewers was also reported previously.<sup>17</sup>  
310 Similarly, BE and MOR formations were also found in other testing sewer conditions, e.g., in  
311 lab reactors<sup>11, 12, 14, 16, 23, 31, 48</sup> and real sewer pipe.<sup>10</sup>

## 312 **Drug Transformation Models**

### 313 *Calibration and validation of the transformation model*

314 Calibrations of the transformation model based on the measured data of this pilot study are  
315 presented in Figure S3.3 and the values of transformation coefficients are reported as mean  
316 with 95% credible intervals (CI) in Table S3.1. Among the estimated  $k'_{bio}$  values, the  
317 relatively high  $k'_{bioa}$  and/or  $k'_{bioan}$  for COC, MDMA, and MTD suggest the important effect



318 of biofilm on their stability in sewers. Moreover, for most biomarkers, their  $k'_{bioa}$  values are  
319 found to be higher than the corresponding  $k'_{bioan}$  values, indicating the higher biofilm-  
320 specific impact in aerobic condition compared to anaerobic condition. Effect of mass transfer  
321 resistance on biodegradation process ( $k'_{bio}$ ) is considered to be limited for most investigated  
322 biomarkers (see discussions in S1.2). Meanwhile, the relatively high  $k_{ww}$  for COC, KET, and  
323 MTD indicate their comparatively evident decreases in the bulk liquid wastewater due to the  
324 processes such as chemical hydrolysis.<sup>10, 12</sup> However,  $k$  value was not estimated for BE or  
325 MOR, which showed significant formations in the pilot sewers (Figure 1). This is because the  
326 knowledge on BE or MOR formation was limited since 1) it is difficult to accurately quantify  
327 the various contributing sources, such as the parent compounds and/or conjugated forms in  
328 the raw wastewater and 2) this study did not spike labelled BE or MOR to exclude the  
329 contribution from other precursors and hence to identify their specific transformation in  
330 sewers. In consequence, the transformation model was not calibrated for the prediction of the  
331 in-sewer formations of BE or MOR.

332 Validity of the estimated pilot-based  $k$  values in diverse sewer environments are subsequently  
333 examined. For each investigated biomarker, the measurement datasets obtained from  
334 different stability studies and the predictive regions generated by the pilot-based  $k$  values  
335 under the corresponding scenario (e.g., A/V ratio and redox condition) are jointly plotted in  
336 Figure 2. Through showing whether the literature data points are encompassed or rejected by  
337 the prediction regions, Figure 2 visually reflects the transferability and applicability of the  
338 pilot-based  $k$  values to various sewer conditions. The rejection probability for each scenario  
339 is reported in Table S3.3.

340 The prediction capability of the pilot-based  $k$  values is successfully validated for KET and  
341 METH under all sewer conditions. As shown in Figure 2, the predictive regions almost

342 completely encompass the observations from the four stability studies under aerobic and  
343 anaerobic conditions, except for a few data points over long HRTs. Validity of the pilot-  
344 based  $k$  values is successfully demonstrated for COC, MDMA, and MTD in the pilot sewers  
345 but is less successful against the testing conditions of the three *sewer reactor studies*. For  
346 these three biomarkers, relatively high rejection probabilities of the pilot-based  $k$  values are  
347 determined for different reasons: 1) the overestimation of COC decreases at the very high  
348 A/V ratio condition ( $175 \text{ m}^{-1}$ ) of Ramin *et al.*<sup>21</sup>; 2) the overestimation of MDMA  
349 transformation compared to its observed insignificant changes in lab reactors,<sup>10, 14</sup> suggesting  
350 that MDMA could be partially degraded by certain microbes that existed in the realistic pilot  
351 sewers but not in lab-scale reactors. Similarly, McCall *et al.*<sup>14</sup> also found the biofilm specific  
352 transformation for MDMA and attributed it to the divergent biofilm growing conditions and  
353 the different transformation potentials; 3) the underestimation of MTD losses compared to  
354 the observed drastic decreases in the anaerobic reactor of Li *et al.*<sup>10</sup> The diffusion limitation  
355 could affect  $k'_{bio}$  in the pilot rising main (with intermittent pumping events) due to the high  
356 biodegradability of MTD (full discussion in S1.2).

### 357 *Model performance evaluation I: the different k-value sets*

358 Multiple  $k$ -value sets ( $k_{ww}^{M_i}$  and  $k'_{bio}{}^{M_i}$ ) are generated from four stability studies as explained in  
359 Table S3.2 and illustrated in Figure S3.3. Figure 3 shows the joint distributions of  $k_{ww}^{M_i}$  and  
360  $k'_{bio}{}^{M_i}$ , which visually reflect the consistency or discrepancy among the  $k$ -value sets estimated  
361 by different studies.

362 For most investigated biomarkers, the close distributions of  $k_{ww}^{M_i}$  indicate the general  
363 congruence of  $k_{ww}^{M_i}$  derived from different stability studies. Such consistency suggests that  
364 the processes in the bulk wastewater (mainly the abiotic physicochemical process) of

365 different sewer environments could lead to similar transformation of biomarkers. However,  
366 certain deviations are found for COC, KET, and MTD under aerobic and/or anaerobic  
367 conditions, where the centroids of  $k_{ww}^{M_1}$  and/or  $k_{ww}^{M_2}$  show relative distances with those of  $k_{ww}^{M_3}$   
368 and/or  $k_{ww}^{M_4}$ . The deviations may be attributed to the assumptions about the independent  
369 relationship of  $k_{ww}$  with wastewater compositions and properties such as the redox potential  
370 or suspended solids concentration. Specific experiments are required to evaluate the effects of  
371 wastewater composition, redox potential, and pH on  $k_{ww}$  and the associated  $k_{bio}$ .

372 By contrast, distributions of  $k'_{bio}{}^{M_i}$  are more deviated for most investigated biomarkers. Since  
373  $k'_{bio}{}^{M_i}$  represents the contribution of biofilm to biomarker stability in sewers, such deviations  
374 reveal that the biofilm-specific effects in different studies are strongly related to the microbial  
375 communities, functions, and activities in biofilms. Generally speaking,  $k'_{bio}{}^{M_1}$  and/or  $k'_{bio}{}^{M_2}$  are  
376 usually found to be higher than  $k'_{bio}{}^{M_3}$  and/or  $k'_{bio}{}^{M_4}$ , except for KET with closer distributions of  
377  $k'_{bio}{}^{M_i}$  at minor levels. This could be explained by the stronger biological activities of biofilms  
378 in the pilot sewer pipes and the sewer reactors of Li *et al.*<sup>10</sup> compared to biofilms in the  
379 reactors of McCall *et al.*<sup>14</sup> and/or Ramin *et al.*<sup>16</sup> Moreover,  $k'_{bio}{}^{M_1}$  and/or  $k'_{bio}{}^{M_2}$  appear to be  
380 generally equivalent for COC, KET, and METH, while present certain dissimilarity for  
381 MDMA and MTD under anaerobic and/or aerobic conditions. Such discrepancies could be  
382 due to the different microbial communities in biofilms grown in pilot-scale sewer pipes and  
383 lab-scale sewer reactors. Also, the continuous mixing condition in the lab reactors<sup>10, 42</sup> could  
384 accelerate the transformation of certain biomarkers in biofilms.

385 The embedded histograms in Figure 3 show the contributions from wastewater and biofilm  
386 processes to the overall transformation of biomarkers in the pilot sewer system. For most  
387 biomarkers, the  $k$ -value sets of M1 and M2 usually suggest the dominant role of biofilms

388 because of the relatively high  $k'_{bio}{}^{M_1}$  and  $k'_{bio}{}^{M_2}$  values, except for MDMA and METH under  
389 anaerobic condition and for KET under both aerobic and anaerobic conditions. The  $k$ -value  
390 sets of M3 and M4 also recognize the governing effect of biofilms on MTD under anaerobic  
391 condition and on KET, MDMA and METH under both aerobic and anaerobic conditions. The  
392 overall KET variation predicted by M3 is actually negligible and akin to the observed high  
393 stability of KET under multiple experimental conditions of McCall *et al.*<sup>14</sup> On the contrary,  
394 the  $k$ -value sets of M3 and M4 indicate higher impact of wastewater processes ( $k_{ww}$ ) on  
395 MTD under aerobic condition and on COC under both aerobic and anaerobic conditions.  
396 These are consistent with the findings of McCall *et al.*<sup>14</sup> and Ramin *et al.*,<sup>16</sup> suggesting that  
397 processes in the bulk wastewater dominated the transformation of MTD (under aerobic  
398 condition) and COC, while sewer biofilms did not significantly enhance their transformation.

#### 399 *Model performance evaluation II: the Bayes Factor, $r$ , and $VE_{cv}$ values*

400 Figure 4 shows the Bayes Factor,  $r$ , and  $VE_{cv}$  across the predictive scenarios generated by the  
401  $k$ -value sets of M1~M4 under the pilot sewer conditions (for detail numbers see Table S3.4 to  
402 S3.6). Predictions of M1~M4 versus observations in the pilot sewer pipes (Day 3 and Day 6)  
403 are jointly plotted for each biomarker in Figure S3.4.

404 As indicated by the results of performance statistics, the  $k$ -value set of M1 provides high  
405 prediction accuracy for COC under the pilot rising main and for MDMA and MTD under  
406 both pilot gravity sewer and rising main conditions. In these cases, the likelihood of  $\log(\text{BFs})$   
407 locating in the positive area is dominant, suggesting the closer agreements of pilot-sewer  
408 measurements to the outputs of M1 than those of M2~M4 (Table S3.4). Moreover,  
409 experimental observations and the outputs of M1 are more closely matched (i.e., close to  $y =$   
410  $x$ ) compared to the correlations for M2~M4. Consistently, higher  $r$  and  $VE_{cv}$  values are  
411 quantified for M1 (Table S3.5 to S3.6), corroborating the higher prediction accuracy of M1

412 for the discussed cases. It should be noted that the datasets of Day 3 and Day 6 from the pilot  
413 sewers may naturally favour M1, which is calibrated under similar conditions.

414 Meanwhile, all the  $k$ -value sets of M1~M4 provide similarly high prediction accuracy for  
415 COC under the pilot gravity sewer condition. The performances of M1~M2 are also  
416 comparable for KET under the pilot rising main condition, where a strong linear relationship  
417 between observations and the predictions of M1 or M2 is found. In these cases,  $\log(\text{BFs})$   
418 mainly spread out within the area close to zero, implying the comparable prediction outputs  
419 of M1~M4 (Table S3.4). Moreover, the  $r$  and  $VEcv$  values of M1~M4 are similar and high  
420 for the discussed cases, which further verify the comparable and reasonable prediction  
421 capabilities of the  $k$ -value sets derived from different stability studies (Table S3.5 to S3.6).

422 For KET under the pilot gravity sewer and METH under both pilot gravity sewer and rising  
423 main conditions, the predicted overall transformations by all  $k$ -value sets are limited, i.e.  
424  $\leq 20\%$  as shown in Figure S3.4. The prediction accuracy of M1~M4 is found to be  
425 deteriorated. In these cases,  $\log(\text{BFs})$  distribute overwhelmingly in the positive area or switch  
426 between positive and negative at different validation sites (Table S3.4). Performance statistics  
427 reveal that the  $r$  values of M1~M4 (Table S3.5) range from low (0.22 for METH in the pilot  
428 gravity sewer), moderate (0.53 for KET in the pilot gravity sewer) to high (0.75 for METH in  
429 the pilot rising main). By contrast, the  $VEcv$  values are low under all scenarios (Table S3.6),  
430 which could be interpreted as a consequence of the poor correlations between predictions and  
431 observations of KET and METH in the pilot sewers. This is because these two biomarkers  
432 had limited but random variations with time, which can hardly be well fitted by the  
433 transformation model. Hence, prediction capabilities of M1~M4 tend to be partially  
434 acceptable for the relatively stable biomarkers without clear temporal variation pattern in  
435 sewers.

436 On the contrary, validity of the  $k$ -value sets derived from different stability studies cannot be  
437 effectively assessed for BE or MOR because their actual stability in the pilot sewers might be  
438 masked by formations from other sources and/or deconjugation processes. Such in-sewer  
439 formations bring uncertainty to the back-estimation of the drugs that could be  
440 metabolized/transformed to BE (i.e., the consumption of COC) or MOR (e.g., the  
441 consumptions of codeine, heroin, and morphine itself in glucuronide form). Information on  
442 the sources contributing to BE and MOR formation, such as the concentrations and  
443 transformation pathways of their parent compounds and/or conjugated forms, should be  
444 determined before the formation processes are incorporated into the transformation model.

445 The comprehensive evaluations of different calibrated transformation models help to choose  
446 the  $k$  values to give the best predictive performance in the application of WBE. Collectively  
447 speaking, for the biomarkers presenting evident losses in sewers (e.g., COC and MTD), both  
448 lab-scale and pilot-scale studies estimate relatively high  $k$  values to predict such decreases  
449 over time. On the other hand, most stability studies suggest the overall insignificant  
450 transformation of KET and METH in sewers. Their predicted transformations are also limited  
451 over time based on all  $k$ -value sets, although with some random variations. Importantly, due  
452 to the complex microbial processes in real sewers, the pilot-based  $k$  values exhibit advantage  
453 in predicting the fate of the biomarkers (e.g., MDMA) which are stable in lab-scale reactors  
454 but show partial degradation in the pilot sewer pipes, especially over long HRT.

455 Unfortunately, none of the available  $k$ -value sets are capable of simulating the increasing  
456 patterns of biomarkers (e.g., BE and MOR) in pilot sewers. In short, the transformation  
457 model calibrated by the measurements in the pilot-scale sewer pipes can be applied as a  
458 credible tool for the future WBE study.

459 ASSOCIATED CONTENT

460 **Supporting Information**

461 Additional information about the layout and parameters of the pilot sewer system, experimental  
462 observations of this pilot study, analytical methods for wastewater samples, and results of  
463 model validation and comparison are provided. Supporting information is available free of  
464 charge via the Internet.

465 AUTHOR INFORMATION

466 **Corresponding Author**

467 \*Guangming Jiang: E-mail: [gjiang@uow.edu.au](mailto:gjiang@uow.edu.au)

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478 REFERENCE

479 1. Daughton, C. G. Pharmaceuticals and Personal Care Products in the Environment:  
480 Overarching Issues and Overview. **2001**, *791*, 2-38.

- 481 2. Daughton, C. G., Illicit drugs in municipal sewage: Proposed new nonintrusive tool to  
482 heighten public awareness of societal use of illicit-abused drugs and their potential for  
483 ecological consequences. In *Pharmaceuticals and Personal Care Products in the*  
484 *Environment: Scientific and Regulatory Issues*, American Chemical Society: Washington  
485 DC, 2001; Vol. 791, pp 348-364.
- 486 3. Zuccato E; Chiabrando C; Castiglioni S; Calamari D; Bagnati R; Schiarea S; R., F.  
487 Cocaine in surface waters- a new evidence-based tool to monitor community drug abuse.  
488 *Environ. Health* **2005**, 4 (14), 7.
- 489 4. van Nuijs, A. L. N.; Castiglioni, S.; Tarcomnicu, I.; Postigo, C.; de Alda, M. L.;  
490 Neels, H.; Zuccato, E.; Barcelo, D.; Covaci, A. Illicit drug consumption estimations derived  
491 from wastewater analysis: A critical review. *Sci. Total Environ.* **2011**, 409 (19), 3564-3577.
- 492 5. European Monitoring Centre for Drugs and Drug *Assessing illicit drugs in wastewater*  
493 *Advances in wastewater-based drug epidemiology*; Luxembourg, 2016.
- 494 6. Ort, C.; Lawrence, M. G.; Rieckermann, J.; Joss, A. Sampling for Pharmaceuticals  
495 and Personal Care Products (PPCPs) and Illicit Drugs in Wastewater Systems: Are Your  
496 Conclusions Valid? A Critical Review. *Environ. Sci. Technol.* **2010**, 44 (16), 6024-6035.
- 497 7. Lai, F. Y.; Ort, C.; Gartner, C.; Carter, S.; Prichard, J.; Kirkbride, P.; Bruno, R.; Hall,  
498 W.; Eaglesham, G.; Mueller, J. F. Refining the estimation of illicit drug consumptions from  
499 wastewater analysis: co-analysis of prescription pharmaceuticals and uncertainty assessment.  
500 *Water Res* **2011**, 45 (15), 4437-48.
- 501 8. Castiglioni, S.; Bijlsma, L.; Covaci, A.; Emke, E.; Hernandez, F.; Reid, M.; Ort, C.;  
502 Thomas, K. V.; van Nuijs, A. L. N.; de Voogt, P.; Zuccato, E. Evaluation of Uncertainties  
503 Associated with the Determination of Community Drug Use through the Measurement of  
504 Sewage Drug Biomarkers. *Environ. Sci. Technol.* **2013**, 47 (3), 1452-1460.



- 505 9. van Nuijs, A. L. N.; Lai, F. Y.; Been, F.; Andres-Costa, M. J.; Barron, L.; Baz-  
506 Lomba, J. A.; Berset, J. D.; Benaglia, L.; Bijlsma, L.; Burgard, D.; Castiglioni, S.;  
507 Christophoridis, C.; Covaci, A.; de Voogt, P.; Emke, E.; Fatta-Kassinos, D.; Fick, J.;  
508 Hernandez, F.; Gerber, C.; Gonzalez-Marino, I.; Grabic, R.; Gunnar, T.; Kannan, K.;  
509 Karolak, S.; Kasprzyk-Hordern, B.; Kokot, Z.; Krizman-Matasic, I.; Li, A.; Li, X. Q.; Love,  
510 A. S. C.; de Alda, M. L.; McCall, A. K.; Meyer, M. R.; Oberacher, H.; O'Brien, J.; Quintana,  
511 J. B.; Reid, M.; Schneider, S.; Simoes, S. S.; Thomaidis, N. S.; Thomas, K.; Yargeau, V.; Ort,  
512 C. Multi-year inter-laboratory exercises for the analysis of illicit drugs and metabolites in  
513 wastewater: Development of a quality control system. *Trac-Trend Anal Chem* **2018**, *103*, 34-  
514 43.
- 515 10. Li, J.; Gao, J.; Thai, P. K.; Sun, X.; Mueller, J. F.; Yuan, Z.; Jiang, G. Stability of  
516 Illicit Drugs as Biomarkers in Sewers: From Lab to Reality. *Environ. Sci. Technol.* **2018**, *52*  
517 (3), 1561-1570.
- 518 11. Gao, J.; Banks, A.; Li, J.; Jiang, G.; Lai, F. Y.; Mueller, J. F.; Thai, P. K. Evaluation  
519 of in-sewer transformation of selected illicit drugs and pharmaceutical biomarkers. *Sci. Total*  
520 *Environ.* **2017**, *609*, 1172-1181.
- 521 12. Thai, P. K.; Jiang, G.; Gernjak, W.; Yuan, Z.; Lai, F. Y.; Mueller, J. F. Effects of  
522 sewer conditions on the degradation of selected illicit drug residues in wastewater. *Water*  
523 *Res.* **2014**, *48*, 538-47.
- 524 13. McCall, A. K.; Bade, R.; Kinyua, J.; Lai, F. Y.; Thai, P. K.; Covaci, A.; Bijlsma, L.;  
525 van Nuijs, A. L.; Ort, C. Critical review on the stability of illicit drugs in sewers and  
526 wastewater samples. *Water Res.* **2016**, *88*, 933-47.
- 527 14. McCall, A. K.; Scheidegger, A.; Madry, M. M.; Steuer, A. E.; Weissbrodt, D. G.;  
528 Vanrolleghem, P. A.; Kraemer, T.; Morgenroth, E.; Ort, C. Influence of Different Sewer

529 Biofilms on Transformation Rates of Drugs. *Environ. Sci. Technol.* **2016**, *50* (24), 13351-  
530 13360.

531 15. McCall, A. K.; Palmitessa, R.; Blumensaat, F.; Morgenroth, E.; Ort, C. Modeling in-  
532 sewer transformations at catchment scale – Implications on drug consumption estimates in  
533 wastewater-based epidemiology. *Water Res.* **2017**, *122*, 655-668.

534 16. Ramin, P.; Brock, A. L.; Causanilles, A.; Valverde Perez, B.; Emke, E.; de Voogt, P.;  
535 Polesel, F.; Plosz, B. G. Transformation and sorption of illicit drug biomarkers in sewer  
536 biofilms. *Environ. Sci. Technol.* **2017**, *51* (18), 10572–10584.

537 17. Gao, J.; Li, J.; Jiang, G.; Shypanski, A. H.; Nieradzick, L. M.; Yuan, Z.; Mueller, J. F.;  
538 Ort, C.; Thai, P. K. Systematic evaluation of biomarker stability in pilot scale sewer pipes.  
539 *Water Res* **2018**, *151*, 447-455.

540 18. Hvitved-Jacobsen, T.; Vollertsen, J.; Nielsen, A. H. *Sewer Processes: Microbial and*  
541 *Chemical Process Engineering of Sewer Networks*. Second ed.; CRC Press: Boca Raton,  
542 2013.

543 19. Plosz, B. G.; Reid, M. J.; Borup, M.; Langford, K. H.; Thomas, K. V.  
544 Biotransformation kinetics and sorption of cocaine and its metabolites and the factors  
545 influencing their estimation in wastewater. *Water Res.* **2013**, *47* (7), 2129-2140.

546 20. Ostman, M.; Fick, J.; Nasstrom, E.; Lindberg, R. H. A snapshot of illicit drug use in  
547 Sweden acquired through sewage water analysis. *Sci. Total Environ.* **2014**, *472*, 862-871.

548 21. Ramin, P.; Libonati Brock, A.; Polesel, F.; Causanilles, A.; Emke, E.; de Voogt, P.;  
549 Plosz, B. G. Transformation and sorption of illicit drug biomarkers in sewer systems:  
550 understanding the role of suspended solids in raw wastewater. *Environ. Sci. Technol.* **2016**,  
551 *50* (24), 13397-13408.

- 552 22. Castiglioni, S.; Zuccato, E.; Crisci, E.; Chiabrando, C.; Fanelli, R.; Bagnati, R.  
553 Identification and measurement of illicit drugs and their metabolites in urban wastewater by  
554 liquid chromatography-tandem mass spectrometry. *Anal Chem* **2006**, *78* (24), 8421-8429.
- 555 23. Baker, D. R.; Kasprzyk-Hordern, B. Critical evaluation of methodology commonly  
556 used in sample collection, storage and preparation for the analysis of pharmaceuticals and  
557 illicit drugs in surface water and wastewater by solid phase extraction and liquid  
558 chromatography-mass spectrometry. *J Chromatogr A*. **2011**, *1218* (44), 8036-59.
- 559 24. Chen, C.; Kostakis, C.; Irvine, R. J.; Felgate, P. D.; White, J. M. Evaluation of pre-  
560 analysis loss of dependent drugs in wastewater: stability and binding assessments. *Drug Test*  
561 *Anal* **2013**, *5* (8), 716-21.
- 562 25. Senta, I.; Krizman, I.; Ahel, M.; Terzic, S. Assessment of stability of drug biomarkers  
563 in municipal wastewater as a factor influencing the estimation of drug consumption using  
564 sewage epidemiology. *Sci Total Environ* **2014**, *487*, 659-65.
- 565 26. Jelic, A.; Rodriguez-Mozaz, S.; Barcelo, D.; Gutierrez, O. Impact of in-sewer  
566 transformation on 43 pharmaceuticals in a pressurized sewer under anaerobic conditions.  
567 *Water Res.* **2015**, *68*, 98-108.
- 568 27. Gao, J.; Li, J.; Jiang, G.; Yuan, Z.; Eaglesham, G.; Covaci, A.; Mueller, J. F.; Thai, P.  
569 K. Stability of alcohol and tobacco consumption biomarkers in a real rising main sewer.  
570 *Water Res.* **2018**, *138*, 19-26.
- 571 28. Ian A. Watson; Sascha E.Oswald; Steven A. Banwart; Roger S. Crouch; Thornton, S.  
572 F. Modeling the dynamics of fermentation and respiratory processes in a groundwater plume  
573 of phenolic contaminants interpreted from laboratory- to field-scale. *Environ. Sci. Technol.*  
574 **2005**, *39* (22), 8829-8839.

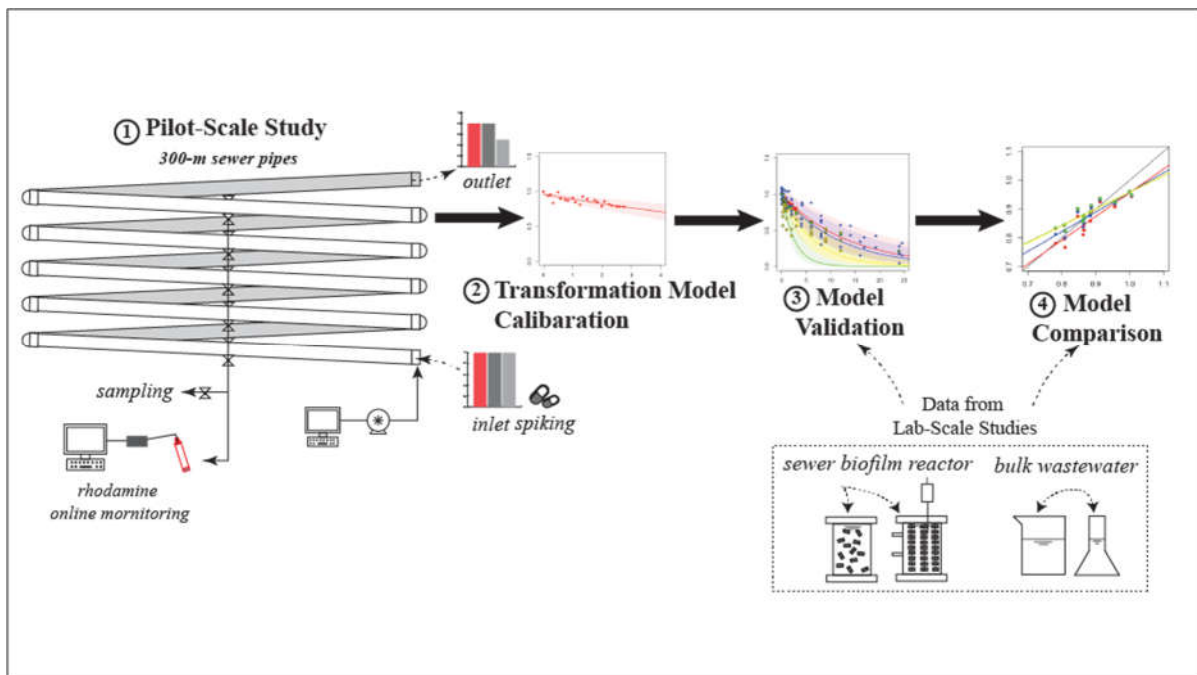
- 575 29. Shypanski, A. H.; Yuan, Z.; Sharma, K. Influence of pressure main pumping  
576 frequency on sulfide formation rates in sanitary sewers. *Environmental Science: Water  
577 Research & Technology* **2018**, *4* (3), 403-410.
- 578 30. Ramin, P.; Polesel, F.; Brock, A. L.; Plósz, B. G. The impact of temperature on the  
579 transformation of illicit drug biomarkers in wastewater. *Sci. Total Environ.* **2018**, *644*, 1612-  
580 1616.
- 581 31. Senta, I.; Krizman, I.; Ahel, M.; Terzic, S. Assessment of stability of drug biomarkers  
582 in municipal wastewater, as a factor influencing the estimation of drug consumption using  
583 sewage epidemiology. *Sci. Total Environ.* **2014**, *487*, 659-665.
- 584 32. Devault, D. A.; Levi, Y.; Karolak, S. Applying sewage epidemiology approach to  
585 estimate illicit drug consumption in a tropical context: Bias related to sewage temperature  
586 and pH. *Sci. Total Environ.* **2017**, *584*, 252-258.
- 587 33. Baker, D. R.; Ocenaskova, V.; Kviclova, M.; Kasprzyk-Hordern, B. Drugs of abuse  
588 in wastewater and suspended particulate matter--further developments in sewage  
589 epidemiology. *Environ Int* **2012**, *48*, 28-38.
- 590 34. Gutierrez, O.; Park, D.; Sharma, K. R.; Yuan, Z. Effects of long-term pH elevation on  
591 the sulfate-reducing and methanogenic activities of anaerobic sewer biofilms. *Water Res.*  
592 **2009**, *43* (9), 2549-2557.
- 593 35. Guisasola, A.; Sharma, K. R.; Keller, J.; Yuan, Z. Development of a model for  
594 assessing methane formation in rising main sewers. *Water Res.* **2009**, *43* (11), 2874-2884.
- 595 36. Liu, Y. W.; Tugtas, A. E.; Sharma, K. R.; Ni, B. J.; Yuan, Z. G. Sulfide and methane  
596 production in sewer sediments: Field survey and model evaluation. *Water Res.* **2016**, *89*, 142-  
597 150.
- 598 37. Liu, Y.; Chen, W.; Arendt, P.; Huang, H.-Z. Toward a Better Understanding of Model  
599 Validation Metrics. *J Mech Design* **2011**, *133* (7).

- 600 38. Li, J. Assessing the accuracy of predictive models for numerical data: Not  $r$  nor  $r^2$ ,  
601 why not? Then what? *Plos One* **2017**, *12* (8), e0183250.
- 602 39. Æsøy, A.; Storfjell, M.; Mellgren, L.; Helness, H.; Thorvaldsen, G.; Ødegaard, H.;  
603 Bentzen, G. A comparison of biofilm growth and water quality changes in sewers with  
604 anoxic and anaerobic (septic) conditions. *Water Sci. Technol.* **1997**, *36* (1), 303-310.
- 605 40. Sharma, K.; Ganigue, R.; Yuan, Z. pH dynamics in sewers and its modeling. *Water*  
606 *Res.* **2013**, *47* (16), 6086-6096.
- 607 41. Jiang, G.; Sharma, K. R.; Guisasola, A.; Keller, J.; Yuan, Z. Sulfur transformation in  
608 rising main sewers receiving nitrate dosage. *Water Res.* **2009**, *43* (17), 4430-4440.
- 609 42. Thai, P. K.; Jiang, G. M.; Gernjak, W.; Yuan, Z. G.; Lai, F. Y.; Mueller, J. F. Effects  
610 of sewer conditions on the degradation of selected illicit drug residues in wastewater. *Water*  
611 *Res.* **2014**, *48*, 538-547.
- 612 43. Sharma, K. R.; Yuan, Z.; de Haas, D.; Hamilton, G.; Corrie, S.; Keller, J. Dynamics  
613 and dynamic modelling of H<sub>2</sub>S production in sewer systems. *Water Res.* **2008**, *42* (10),  
614 2527-2538.
- 615 44. Foley, J.; Yuan, Z.; Keller, J.; Senante, E.; Chandran, K.; Willis, J.; Shah, A.; van  
616 Loosdrecht, M.; van Voorthuizen, E. *N<sub>2</sub>O and CH<sub>4</sub> emission from wastewater collection and*  
617 *treatment systems: technical report*; Global Water Research Coalition: London, United  
618 Kingdom, 2011.
- 619 45. Guisasola, A.; de Haas, D.; Keller, J.; Yuan, Z. Methane formation in sewer systems.  
620 *Water Res.* **2008**, *42* (6), 1421-1430.
- 621 46. Liu, Y. W.; Sharma, K. R.; Fluggen, M.; O'Halloran, K.; Murthy, S.; Yuan, Z. G.  
622 Online dissolved methane and total dissolved sulfide measurement in sewers. *Water Res.*  
623 **2015**, *68*, 109-118.

624 47. Wang, K. J.; Zeeman, G.; Lettinga, G. Alteration in Sewage Characteristics Upon  
625 Aging. *Water Sci. Technol.* **1995**, 31 (7), 191-200.

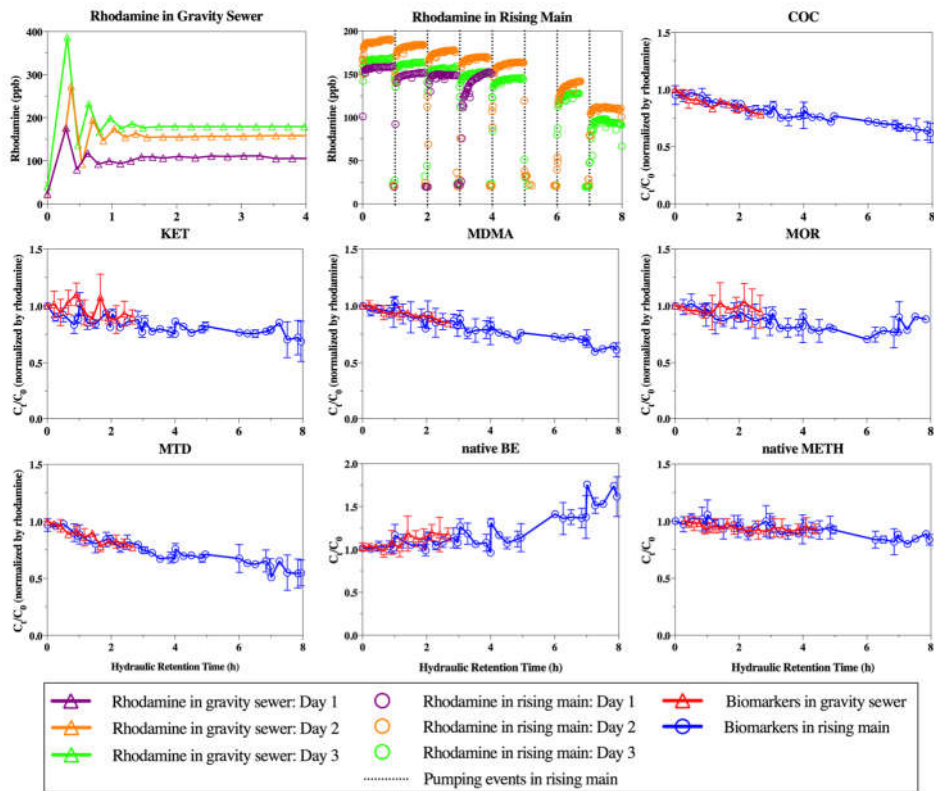
626 48. Ramin, P.; Valverde-Perez, B.; Polesel, F.; Locatelli, L.; Plosz, B. G. A systematic  
627 model identification method for chemical transformation pathways - the case of heroin  
628 biomarkers in wastewater. *Sci Rep* **2017**, 7 (1), 9390.

629

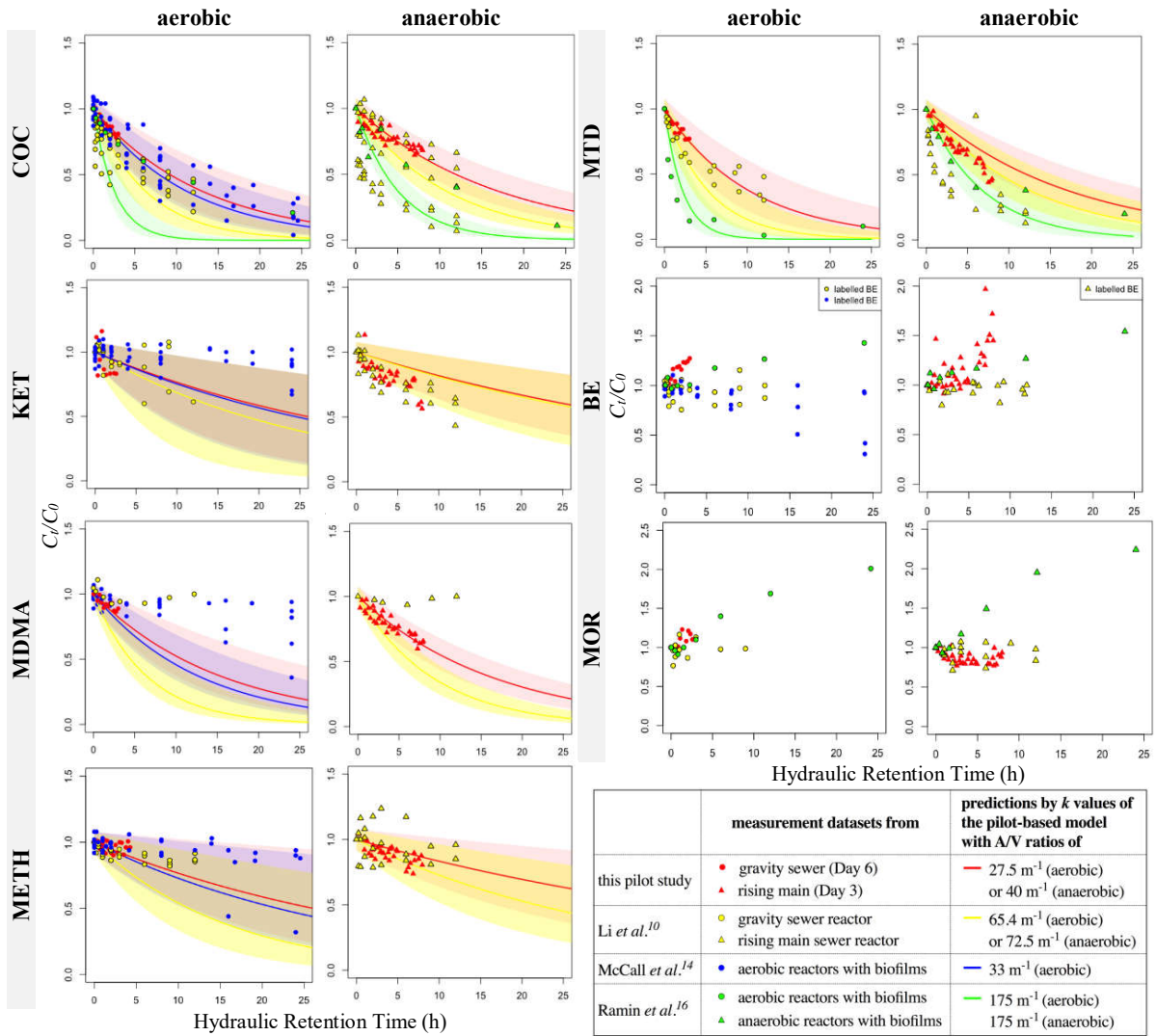


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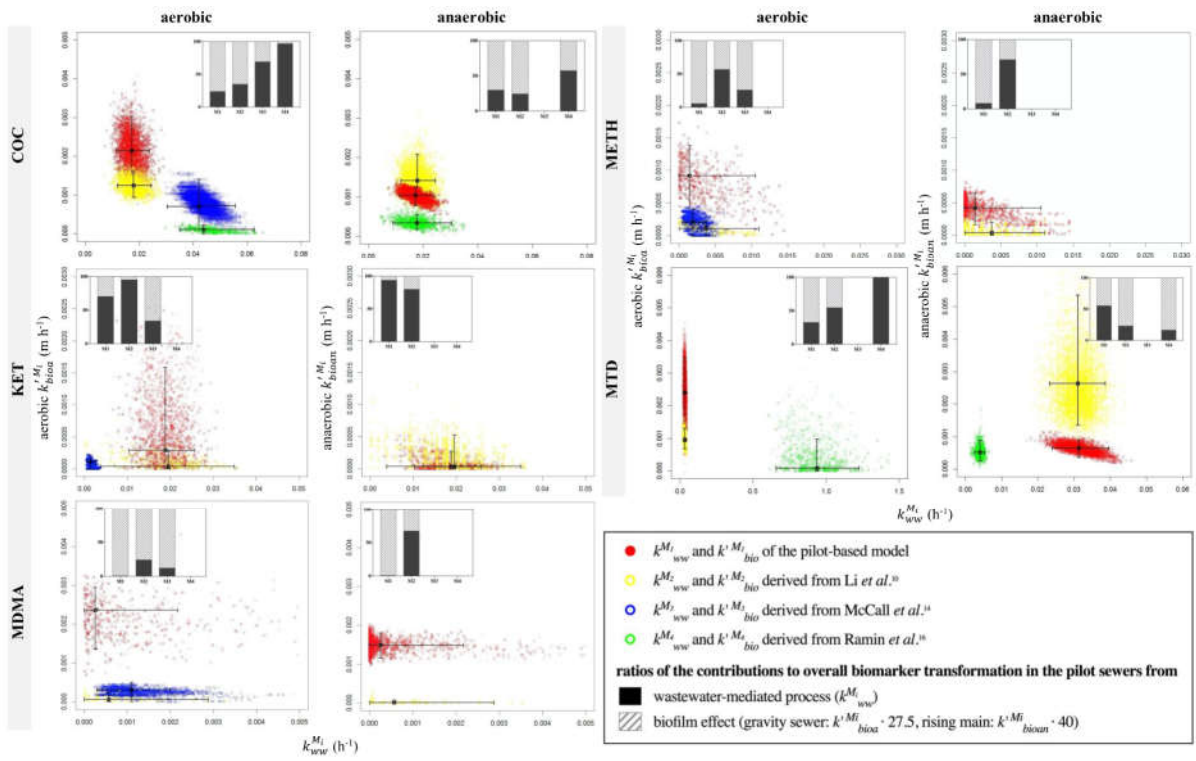


**Figure 1.** Transportation of rhodamine as a flow tracer and stability of biomarkers in the pilot rising main pipe and the pilot gravity sewer pipe. (Transformations of MOR and MTD in the pilot sewer pipes were also reported recently<sup>17</sup>)

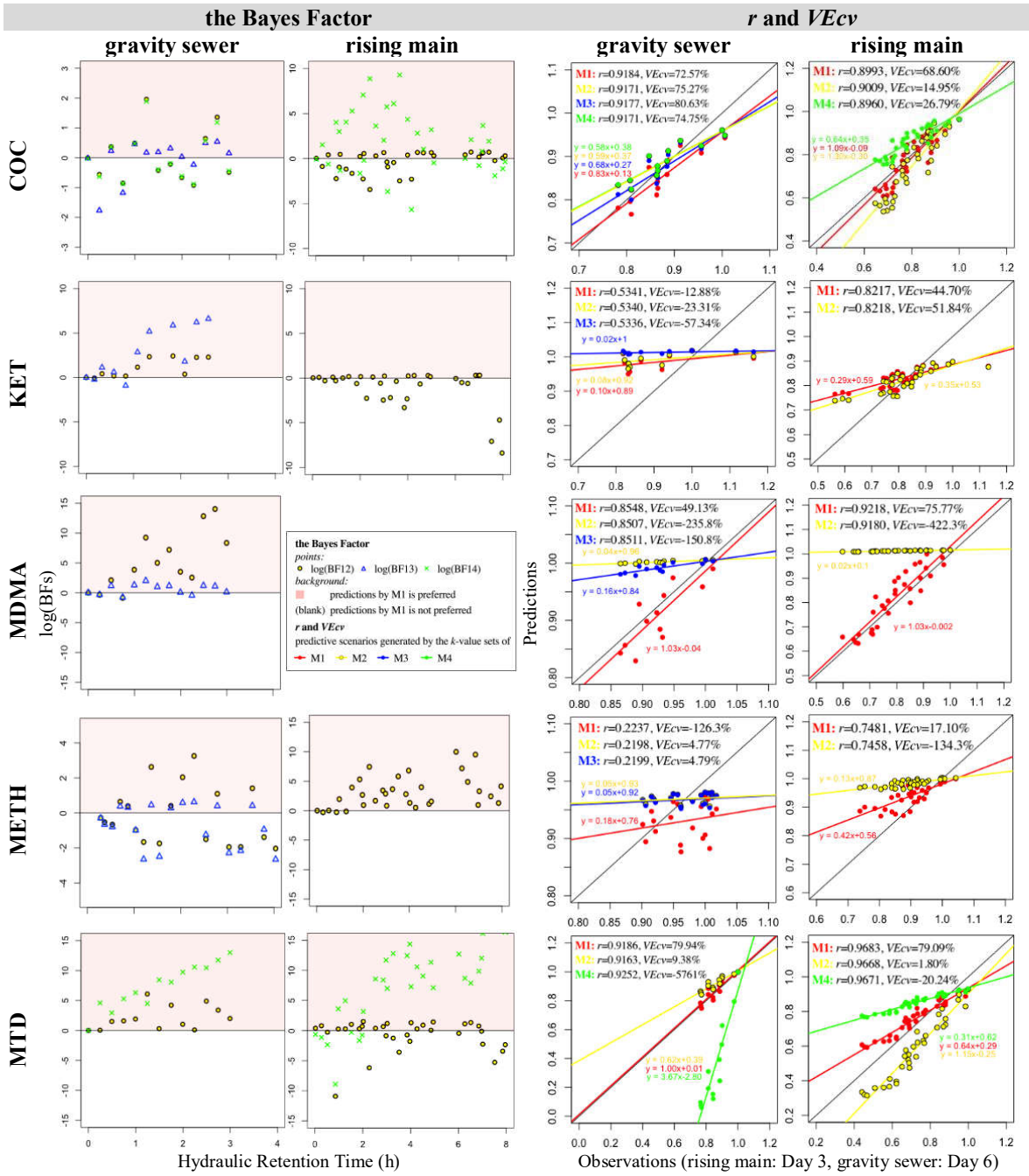


**Figure 2.** Validation of the pilot-based transformation coefficients through the predictive scenarios of the corresponding experimental observations in different stability studies.





**Figure 3.** Joint distributions of the transformation coefficients ( $k_{ww}^{M_i}$  and  $k'_{bio}^{M_i}$ ) derived from the four stability studies (M1~M4) under aerobic and anaerobic conditions. The embedded histograms indicate the contributions from wastewater processes against biofilm effects to the overall biomarker transformation in the pilot sewer pipes.



**Figure 4.** Model performance comparison: results of the Bayes factor,  $r$ , and  $VEcv$  for comparing the predictive scenarios generated by the  $k$ -value sets of four stability studies (M1 to M4) under the conditions of the pilot sewer pipes.