

# 1 Removal of the pesticide tebuconazole in constructed wetlands:

## 2 Design comparison, influencing factors and modelling

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### 8 Abstract

9 Constructed wetlands (CWs) are a promising technology to treat pesticide  
10 contaminated water, but its implementation is impeded by lack of data to optimize  
11 designs and operating factors. Unsaturated and saturated CW designs were used to  
12 compare the removal of triazole pesticide, tebuconazole, in unplanted mesocosms  
13 and mesocosms planted with five different plant species: *Typha latifolia*, *Phragmites*  
14 *australis*, *Iris pseudacorus*, *Juncus effusus* and *Berula erecta*. Tebuconazole removal  
15 efficiencies were significantly higher in unsaturated CWs than saturated CWs, showing  
16 for the first time the potential of unsaturated CWs to treat tebuconazole  
17 contaminated water. An artificial neural network model was demonstrated to provide  
18 more accurate predictions of tebuconazole removal than the traditional linear  
19 regression model. Also, tebuconazole removal could be fitted an area-based first order

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20 kinetics model in both CW designs. The removal rate constants were consistently  
21 higher in unsaturated CWs (range of 2.6–10.9 cm d<sup>-1</sup>) than in saturated CWs (range of  
22 1.7–7.9 cm d<sup>-1</sup>) and higher in planted CWs (range of 3.1–10.9 cm d<sup>-1</sup>) than in unplanted  
23 CWs (range of 1.7–2.6 cm d<sup>-1</sup>) for both designs. The low levels of sorption of  
24 tebuconazole to the substrate (0.7–2.1%) and plant phytoaccumulation (2.5–12.1%)  
25 indicate that the major removal pathways were biodegradation and metabolization  
26 inside the plants after plant uptake. The main factors influencing tebuconazole  
27 removal in the studied systems were system design, hydraulic loading rate and plant  
28 presence. Moreover, tebuconazole removal was positively correlated to dissolved  
29 oxygen and all nutrients removal.

## 30 Capsule

31 System design, plant presence and species and HLR influence tebuconazole treatment  
32 in CWs, and the removal can be described by an artificial neural network model.

33 **Keywords:** Artificial neural network; Biofilm reactors; Contaminants of emerging  
34 concern; Fungicides; Phytoremediation

35

## 36 1. Introduction

37 Tebuconazole is a triazole pesticide that is widely used in agriculture for crop  
38 protection due to its broad spectrum of antifungal activities (Shikuku et al., 2014) and  
39 included as an active ingredient in wood preservatives (Miyauchi et al., 2005).  
40 Concentration levels of tebuconazole (Table S1) ranging from ng L<sup>-1</sup> to µg L<sup>-1</sup> have been

41 found in both rural and urban water bodies (Bollmann et al., 2014; Casas & Bester,  
42 2015; Shikuku et al., 2014). However, tebuconazole is reported to be toxic to aquatic  
43 life and human health at  $\mu\text{g L}^{-1}$  level (EFSA, 2014). In the last few decades, the  
44 occurrence of pesticides, including tebuconazole, in the aquatic environment has  
45 become a worldwide issue of increasing environmental concern. Thus, due to its wide-  
46 spread use and detection as well as its potentially harmful effects, tebuconazole was  
47 selected as the model pesticide in this study.

48       Constructed wetlands (CWs) have become widely used to treat pesticide  
49 contaminated wastewater as an economical, robust and sustainable technology  
50 (Vymazal & Březinová, 2015). Previous research on removal of pesticides in CWs has  
51 been conducted mostly in saturated systems, such as free water surface CWs or  
52 horizontal subsurface flow CWs, while unsaturated systems, such as vertical flow CWs,  
53 have been less studied. Vegetated and non-vegetated saturated surface flow CWs  
54 have been reported to result in removal of 45%–90% of the tebuconazole at an inflow  
55 concentration of 0.1–10  $\mu\text{g L}^{-1}$  in agricultural landscapes in Europe (Passeport et al.,  
56 2013; Tournebize et al., 2013). Unsaturated CWs have usually better removal  
57 efficiency of the typical wastewater constituents BOD and ammonium, due to better  
58 oxygen transfer capability in their wetland beds (Wu et al., 2014). However, there are  
59 no results of direct comparisons in pesticide removal efficiencies among different  
60 types of CWs. Thus, the most effective CW type to treat pesticides has not yet been  
61 determined. Unsaturated CWs have different hydrological characteristics including  
62 water flow pathway and hydraulic retention time compared with saturated CWs.  
63 These features suggest possibly different contaminant removal efficiencies and

64 mechanisms (Gregoire et al., 2009; Kadlec & Wallace, 2008; Vymazal, 2007). Thus,  
65 comparisons of pesticide removal performance, kinetics and mechanisms in different  
66 CW designs are necessary to provide better information for future applications.

67 To date, the factors influencing removal of the pesticide tebuconazole in  
68 different CWs have been rarely investigated. For instance, one popularly used  
69 pesticide, chlorpyrifos, has been reported that the removal efficiency and removal  
70 rate constant were negatively affected by increased influent concentrations from 100  
71  $\mu\text{g L}^{-1}$  to 500  $\mu\text{g L}^{-1}$  to 1  $\text{mg L}^{-1}$  levels through phytoremediation (Prasertsup &  
72 Ariyakanon, 2011). These concentrations are high, especially if considering that typical  
73 concentrations in urban storm water are usually below 100  $\mu\text{g L}^{-1}$  (Bollmann et al.,  
74 2014; Casas & Bester, 2015). Thus, the effect of the influent concentration of pesticide  
75 on removal under real environmental levels is unknown. Different hydraulic loading  
76 rates (HLRs) affect pollutant/microbial contact time and reaction rates, which has an  
77 effect on pollutants, such as BOD, nitrogen and some pharmaceuticals,  
78 biodegradation (Lin et al., 2008; Zhang et al., 2017). Despite this, the effect of HLR on  
79 pesticide removal in CWs has not received much attention, even though removal  
80 kinetics models, such as the zero or first order kinetics models, are calculated based  
81 on pollutant removal under different HLRs. Thus, we lack the information on pesticide  
82 removal efficiencies under different HLRs that is needed to be able to determine  
83 pesticide removal kinetics. It is expected that different plant species may influence  
84 pesticide removal in CWs differently due to their different root structure, root exudate  
85 release, compound uptake ability and associated different microbial communities. Lv  
86 et al. (2016c) observed that tebuconazole removal in saturated CW mesocosms was

87 influenced by the identity of the plant species, while plant uptake and substrate  
88 sorption made limited contributions towards tebuconazole removal. However,  
89 whether these factors also influence tebuconazole removal in unsaturated CW is  
90 unknown. Understanding the factors influencing removal of the pesticide  
91 tebuconazole in different CW designs would undoubtedly improve the design and  
92 operation of CWs for the treatment of not only tebuconazole but also other triazole  
93 pesticides.

94         Reliable numerical models can be used to increase the understanding of  
95 pollutant removal processes occurring in CWs and to improve existing design criteria  
96 of CWs (Langergraber, 2007). Linear regression has been the most widely used model  
97 in CWs for predictions of pollutant removal (Rousseau et al., 2004). However, linear  
98 regression provides rather crude approximations of the complex assortment of  
99 nonlinear relationships present in environmental systems (May & Sivakumar, 2009).  
100 Artificial neural network (ANN) modelling is a technique inspired by biological neuron  
101 processing, which addresses an interconnected structure of processing elements. ANN  
102 is widely used in solving complex and nonlinear problems (Schmidhuber, 2015). In  
103 recent years, ANN has been successfully applied to predict the removal abilities of  
104 organic matter (COD and BOD<sub>5</sub>) (Akratos et al., 2008), TSS (Naz et al., 2009), different  
105 phosphorous species (ortho-P and TP) (Akratos et al., 2009) and nitrogen (NH<sub>4</sub><sup>+</sup>-N and  
106 TN) (Guo et al., 2014; Kotti et al., 2016) in various types of CWs. However, no study  
107 has been conducted on ANN model-based simulation for pesticides or other emerging  
108 organic contaminants.

109           Consequently, the main objectives of the present study were the following: (1)  
110 to compare the removal efficiency, kinetics and mechanism of tebuconazole removal  
111 in both unsaturated and saturated CWs with different plant species; (2) to investigate  
112 the main influencing factors (system design, HLR, initial concentration and plant  
113 species) of tebuconazole removal in both types of CW designs; and (3) to compare  
114 ANN with traditional linear regression models in order to explore a simple and robust  
115 methodology suitable for predicting tebuconazole removal in CWs.

## 116 2. Materials and methods

### 117 2.1 Mesocosm-scale CWs and experimental conditions

118           Each mesocosm-scale CW was made of a black plastic container with both a  
119 height and diameter of 20 cm. Each container was filled with a 4 cm layer of gravel ( $\emptyset$   
120 0.8 to 1.2 cm) on the bottom, a geotextile, a 10 cm layer of sand ( $\emptyset$  0.05 to 0.1 cm with  
121 average porosity of 37%) and finally a 4 cm layer of gravel. All mesocosm-scale CWs  
122 were intermittently pulse fed by water artificially spiked with tebuconazole from the  
123 surface. The outlet height was set at 3 cm for unsaturated CWs (Fig. 1a) and 15 cm for  
124 saturated CWs (Fig. 1b). The system was setup and used for a previous experiment  
125 along summer 2014 and winter 2015 by Lv et al. (2016c). Both unsaturated and  
126 saturated CWs consisted of an influent tank and triplicates of six planting types:  
127 unplanted and planted with *Juncus effusus* (*Juncus*), *Typha latifolia* (*Typha*), *Berula*  
128 *erecta* (*Berula*), *Phragmites australis* (*Phragmites*) and *Iris pseudacorus* (*Iris*). In total,  
129 36 mesocosm-scale CWs were constructed, 18 for the unsaturated and 18 for the  
130 saturated design. Artificially spiked tebuconazole water was prepared in 300 L doses

131 and constantly mixed by a submerged centrifugal pump placed at the bottom of the  
132 influent tank. New influent was prepared every 2–5 days, and the influent load was  
133 controlled by a timer and pump. Two concentrations of tebuconazole (10 and 100  $\mu\text{g}$   
134  $\text{L}^{-1}$ ) and four hydraulic loading rates (1.7, 3.4, 6.9 and 13.8  $\text{cm d}^{-1}$ ) were used. The  
135 corresponding hydraulic retention time (HRT) for the saturated CWs were 2, 1, 0.5 and  
136 0.25 days, respectively. The wastewater was prepared with “Pioner Grøn” (Brøste  
137 Group, Denmark) N:P:K full strength nutrient solution added to tap water  
138 (supplementary material). An additional carbon source for basic microbial community  
139 survival using acetic acid was used to simulate a 20  $\text{mg L}^{-1}$  TOC load. The experiment  
140 lasted from July to August 2015 (57 days) after a two-month stabilization period. The  
141 air temperature ranged from 15 to 25 °C and the relative air humidity from 51 to 78%  
142 (Fig. 1c).

## 143 **2.2 Sampling and analysis**

144 Before each sampling, the mesocosms were allowed to stabilize for three  
145 complete hydraulic cycles (calculated by the saturated mesocosms), after which the  
146 effluent quality was assumed to be representative. The triplicates samples of the  
147 influent were collected directly from the influent tank using a 1 L amber flask. Similar  
148 1 L amber flasks were connected to the CW effluent flow valve and left *in-situ* for 2-  
149 10 hours, in order to collect a minimum of 800 mL of composite water samples. In  
150 total, eight sampling campaigns were conducted. For each campaign, a total of 42  
151 samples were collected: the influent (3) plus effluent samples (3 x 6) for each design  
152 (x2). The volume of each effluent was noted to calculate water loss by  
153 evapotranspiration (Equation S1, supplementary materials). Dissolved oxygen (DO),

154 pH, temperature and electrical conductivity (EC) were measured *in-situ*. The nutrients  
155 of total nitrogen (TN), total organic carbon (TOC) NH<sub>4</sub>-N, NO<sub>3</sub>-N and PO<sub>4</sub>-P were  
156 analysed within 12 hours. A detailed description of the measurements can be found  
157 in supplementary materials. The pesticide tebuconazole was pre-concentrated by  
158 solid-phase extraction (SPE) prior to further analysis using an HPLC system (Thermo  
159 Scientific Ultimate 3000) equipped with a diode array detector (Lv et al., 2016c).  
160 Tebuconazole removal efficiencies were corrected for water loss due to  
161 evapotranspiration according to Equation S2. The removal of tebuconazole from the  
162 water samples was simulated to fit both area-based (Equation S3) and volume-based  
163 (Equation S4) first order kinetics models.

164 The weight of each CW mesocosm was measured at the end of the experiment  
165 to roughly estimate the fresh biomass of the different plant species according to  
166 Equation S5. Approximately 100 g of plant aerial tissue and 100 g of substrate (1 cm Ø  
167 cores at 4–14 cm depth) were collected at the end of experiment. Substrate and plant  
168 tissue samples were extracted by ultrasonication, and the substrate extracts were  
169 analyzed directly while the plant extracts were further cleaned by two-stage  
170 saponification and SPE prior to analysis by HPLC (Lv et al., 2017b). Total substrate  
171 organic carbon content (SOC) was estimated by the loss on ignition method (LOI) using  
172 a muffle furnace and calculated by Equation S6. All the equations for the calculations  
173 are described in detail in the supplementary materials.

### 174 **2.3 Modelling**



175 All measured variables, besides the data of tebuconazole removal, influent  
176 water quality (tebuconazole influent concentration, HLR, water temperature, pH, EC,  
177 DO and evapotranspiration) and nutrient removal (TOC, TN, TP, NH<sub>4</sub>-N and NO<sub>3</sub>-N) in  
178 the two studied CW designs were used in the principal component analysis (PCA).  
179 Moreover, one more parameter labelled as “plant” was included in the PCA. The value  
180 was set as 0 for the unplanted CWs and 1 for planted CWs. The loading factors,  
181 corresponding to the principal components that explained most of the variation in the  
182 original data were extracted to be used in the ANN and linear regression models as  
183 input parameters.

184 The multi-layer perceptron (MLP) type of ANN, the most popular method used  
185 in hydrological modelling (Govindaraju, 2000a; Govindaraju, 2000b), was used. In MLP,  
186 the artificial neurons are arranged in a layered configuration (Fig. 2) containing a single  
187 input layer, a single processing (hidden) layer and a single output layer. For the linear  
188 regression analysis, tebuconazole removal efficiency was simulated as the summation  
189 of each selected variable multiplied by a factor. For the saturated CWs, all the input  
190 data from the present study was used for model training (144 data points). Moreover,  
191 data from the previous experiment by Lv, et al. (2017) for tebuconazole removal was  
192 used to validate the saturated CW model. For the unsaturated CWs, since this is the  
193 first study reporting tebuconazole removal under unsaturated conditions, the input  
194 data was randomly split into two subsets (2:1 ratio), with the larger subset used for  
195 training and the other subset used for validation. Thus, 96 and 48 data points were  
196 used to train and validate the models, respectively. The Mean Absolute Error  
197 (Equation S8) was used to evaluate the precision of both ANN and linear regression

198 models during model validation in the present study. The detailed data training and  
199 simulation of the ANN and linear regression models are described in the  
200 supplementary materials.

## 201 **2.4 Statistical analysis and software**

202 Statistical analyses were carried out using the XLStat Pro® statistical software  
203 (XLStat, Paris, France). Analysis of variance (ANOVA) followed by Tukey's HSD test was  
204 used to identify significant differences in water quality (pH, EC, and DO), influencing  
205 factors (plant species, HLR, influent concentration and system designs) on  
206 tebuconazole removal, reaction rate constants, tebuconazole concentration by  
207 substrate sorption and plant uptake, and nutrient (TOC, TN, NH<sub>4</sub>-N and TP) removal at  
208 the 0.05 significance level ( $p < 0.05$ ). Plant height and leaf chlorophyll differences at the  
209 beginning and end of the experiment were compared with the student-T test. The data  
210 were checked for normality and homogeneity of variance prior to statistical analysis.  
211 If variables were not normally distributed, they were log-transformed. Composite  
212 values of water quality (pH, EC, and DO) and nutrient removal (TOC, TN, NH<sub>4</sub>-N and TP)  
213 from each mesocosm throughout the whole study were visualized using beanplots by  
214 the program BoxPlotR. PCA, ANN and linear regression models were performed in  
215 StatSoft Statistica version 7 (StatSoft). Visualized co-occurrence network figures were  
216 illustrated using the Gephi platform (Bastian et al., 2009).

## 217 **3. Results**

### 218 **3.1 Plant vitality and water quality**

219 The height and leaf chlorophyll of each wetland plant species were not  
220 significantly different at the end of the experiment when compared with that at the  
221 beginning of the experiment (Fig. S1). Thus, the tebuconazole concentration levels  
222 tested in the present setup did not reveal toxic effects on the plants, which is in  
223 agreement with previous studies (Lv et al., 2016a; Lv et al., 2016c).

224 The values of DO, pH and EC in the effluent under different HLRs and influent  
225 concentration levels during the whole experiment were not significantly different  
226 between the planted mesocosms for either unsaturated or saturated CWs (Table S2).  
227 Thus, for easier visualization of the results, these parameters for all influent and  
228 effluent of unplanted mesocosms and planted mesocosms were integrated and  
229 displayed in beanplots for the comparison (Fig. 3). The DO values were significantly  
230 higher in the effluent (2.1–9.2 mg/L) than in the influent (0.2–3.8 mg/L), and  
231 significantly higher in planted (4.2–9.2 mg/L) than unplanted mesocosms (2.1–8.5  
232 mg/L) for both unsaturated and saturated CWs. Moreover, DO in unsaturated CWs  
233 were significantly higher than in the unplanted mesocosms from both the saturated  
234 and unsaturated designs. The pH in the unplanted mesocosms (8.6–9.0) was  
235 significantly higher than in the planted mesocosms (7.7–8.8) for both saturated and  
236 unsaturated CWs. No difference was observed in pH between the CW designs. The EC  
237 values of the influent and effluent for all mesocosms were similar and ranged between  
238 610 to 680  $\mu\text{S cm}^{-1}$ , and they were also not significantly different between CW designs.

239 The nutrient removal performances in all CW systems were combined and  
240 visualized in beanplots (Fig. S2). The removal of TP, TN and  $\text{NH}_4\text{-N}$  in the unplanted

241 mesocosms (40%-60%) was significantly lower than those in the planted mesocosms  
242 (80%-100%) for both unsaturated and saturated CWs. TOC (26% to 94%) and NO<sub>3</sub>-N  
243 (29% to 99%) removal was not significantly different between mesocosms or CW  
244 design. Nutrient removal was not different between different types of planted  
245 mesocosms for both CWs designs. Generally, the nutrient removal in unsaturated CWs  
246 tended to be higher, but not significantly, than the removal in the corresponding  
247 saturated mesocosms under all the operation conditions throughout the study.

### 248 **3.2 Tebuconazole removal**

249 Tebuconazole was removed in both unsaturated and saturated CWs (Fig. 4) at  
250 removal efficiencies reaching up to 99.8%. A significant effect of system design on  
251 tebuconazole removal efficiency was observed through a four-way ANOVA test (Table  
252 1). Efficiencies were generally significantly higher in the unsaturated CWs than in the  
253 corresponding saturated CWs for the corresponding mesocosm type and HLR. The  
254 results of the four-way ANOVA test also showed that tebuconazole removal was  
255 significantly affected by HLR and mesocosm types (plant species) for both unsaturated  
256 and saturated CWs. Tebuconazole removal efficiencies showed similar patterns for  
257 both influent concentration levels, showing an increase from 21.0% to 99.8% at an  
258 HLR decrease from 13.8 cm d<sup>-1</sup> to 1.8 cm d<sup>-1</sup>. Table S3 shows that all planted  
259 mesocosms achieved significantly higher tebuconazole removal (33% to 99.8%)  
260 compared with unplanted controls (21% to 66.1%). Moreover, mesocosms planted  
261 with *Berula* (71% to 99.8%) showed significantly higher removal efficiency than the  
262 other plant species in both unsaturated and saturated CWs. Regarding the influent

263 concentration ( $10 \mu\text{g L}^{-1}$  to  $100 \mu\text{g L}^{-1}$ ) in both CW designs, removal was not  
264 significantly affected by this factor (Table 1).

### 265 **3.3 Kinetics of tebuconazole removal**

266 The area-based first order kinetics model was applied to determine the  
267 tebuconazole removal rate constants in both unsaturated and saturated CWs (Table  
268 2). The area-based removal rate constants ( $k$ ) were not influenced by influent  
269 concentration levels ( $10$  and  $100 \mu\text{g L}^{-1}$ ). For unsaturated CWs, the  $k$  value was  
270 significantly lower in the unplanted mesocosms ( $2.6 \pm 0.8 \text{ cm d}^{-1}$ ) than in the planted  
271 mesocosms (ranging from  $5.3$  to  $10.9 \text{ cm d}^{-1}$ ). Moreover, the  $k$  value of *Berula* ( $10.9 \pm$   
272  $2.6 \text{ cm d}^{-1}$ ) mesocosms was significantly higher than the  $k$  values for the other planted  
273 mesocosms. For saturated CWs, the  $k$  value was also significantly lower in the  
274 unplanted mesocosms ( $1.7 \pm 0.5 \text{ cm d}^{-1}$ ) than in the planted mesocosms (ranging from  
275  $3.1$  to  $7.9 \text{ cm d}^{-1}$ ). The *Berula* mesocosms also had significantly higher  $k$  values ( $7.9 \pm$   
276  $1.2 \text{ cm d}^{-1}$ ) than the other planted mesocosms for saturated CW design. Additionally,  
277 the area-based removal rate constants for the unsaturated CWs were significantly  
278 higher than the corresponding mesocosms for saturated CWs, except for *Typha* and  
279 *Phragmites* mesocosms.

280 The volume-based first order kinetics model was additionally applied to describe  
281 tebuconazole removal in saturated CWs. The volume-based removal rate constants  
282 ( $k_v$ ), half-life and  $R^2$  are presented in Table 2. The  $k_v$  (volume-based removal rate)  
283 values were not affected by the influent concentration levels ( $10$  and  $100 \mu\text{g L}^{-1}$ ).

284 *Berula* planted mesocosms had significantly higher  $k_v$  values ( $3.7 \pm 0.8 \text{ d}^{-1}$ ) than the  
285 other mesocosms (ranging from 0.8 to  $1.6 \text{ d}^{-1}$ ).

### 286 **3.4 Substrate sorption and plant uptake**

287 The average values of the substrate tebuconazole concentrations and the  
288 substrate total organic carbon (SOC) content at the end of the experiment were higher  
289 in the unsaturated CWs than in the corresponding saturated mesocosms (Fig. S3a and  
290 b). However, tebuconazole concentrations normalized for the SOC content in  
291 unsaturated CWs were significantly lower than the corresponding saturated  
292 mesocosms (Fig. S3c), except for *Berula* mesocosms. Moreover, tebuconazole  
293 normalized concentrations for unplanted mesocosms were generally significantly  
294 higher than for the planted mesocosms in both designs. Based on the mass balance of  
295 the total tebuconazole spiked into each mesocosm, it can be estimated that sorption  
296 to the substrate represents only 1.6%–2.1% and 0.7%–1.5% of the tebuconazole  
297 removed in the unsaturated and saturated CWs, respectively.

298 Regarding phytoaccumulation, tebuconazole concentration in the aboveground  
299 tissue of the different plants ranged from 0.7 to  $3.8 \text{ mg kg}^{-1} \text{ DW}$  (Fig. S3d) at the end  
300 of the experiment. Assuming tebuconazole translocation factors (range of 0.27 to 3.9)  
301 and biomass aboveground/roots ratios (range of 0.3 to 0.7) based on previous  
302 research by Lv et al. (2016b), it can be estimated that phytoaccumulation represented  
303 3.6%–12.1% and 2.5%–11.7% of the tebuconazole removed in the unsaturated and  
304 saturated CWs, respectively.

### 305 **3.5 Co-occurrence networks**

306 Co-occurrence networks were computed to facilitate the visualization of the  
307 correlations between all measured parameters for unsaturated CWs (Fig. 5a) and  
308 saturated CWs (Fig. 5b). Only the significant ( $p$ -value  $< 0.01$ ) and strong ( $|\text{Pearson's } r|$   
309  $\geq 0.4$ ) correlations (Table S4) are shown in Fig. 5. Tebuconazole removal showed  
310 significant and strong positive correlations with DO, evapotranspiration, plant and  
311 removal of TOC, TP, TN,  $\text{NH}_4\text{-N}$ , and  $\text{NO}_3\text{-N}$ , and negatively correlated with HLR.  
312 Moreover, the detailed correlations between tebuconazole removal and the  
313 parameters were analysed (Fig. S4). Notably, besides supporting the previously  
314 identified significant and strong correlations, the positive correlation between TN and  
315  $\text{NH}_4\text{-N}$  removal and tebuconazole removal showed a significantly higher slope for  
316 planted CWs than for all unplanted CWs. Further analysis showed no significant  
317 correlation between removal of TP, TN and  $\text{NH}_4\text{-N}$  with evapotranspiration in the  
318 planted CWs (Fig. S5). However, once again, the difference in TN and  $\text{NH}_4\text{-N}$  removal  
319 between planted and unplanted CWs is clear.

### 320 **3.6 Modeling and validation**

321 All the measured parameters were analysed by PCA, and the extracted “best  
322 loading factors” of the candidate variables on the principal components axes are  
323 reported in Table 3. Four principal components were found to explain 80% and 83%  
324 of the variance in the original dataset for unsaturated and saturated CWs, respectively.  
325 For both designs, the first two principal components explained the majority of the  
326 variability (around 65%) in the dataset. The first principal components were highly  
327 correlated to DO, evapotranspiration, plant and removal of TOC, TN, TP,  $\text{NH}_4\text{-N}$ , and  
328  $\text{NO}_3\text{-N}$ , which are parameters presenting large loading factors. Concerning the second

329 principal component, HLR had a large loading factor, implying that it should also be  
330 used as input variable for the models. Thus, the factors DO, evapotranspiration, HLR,  
331 plant and removal of TOC, TN, TP, NH<sub>4</sub>-N, NO<sub>3</sub>-N were selected as the input  
332 parameters for the ANN and linear regression models.

333 Both the ANN and linear regression models provided reliable (MAE ≤ 0.32 and  
334 R<sup>2</sup> ≥ 0.73) simulations of tebuconazole removal for both unsaturated CWs (Fig. 6a) and  
335 saturated CWs (Fig. 6c). However, the ANN model showed lower MAE and higher R<sup>2</sup>  
336 values than the linear regression model for both designs (Fig. 6a and c). Regarding  
337 model validation, the ANN model showed a slightly lower value of MAE (0.22) and  
338 higher R<sup>2</sup> (0.85) than those of the linear regression model (MAE= 0.24, R<sup>2</sup>=0.81) for  
339 unsaturated CWs. For the saturated CWs (Fig. 6d), the ANN model showed a  
340 significantly lower value of MAE (0.44) and higher R<sup>2</sup> (0.61) than those for the linear  
341 regression model (MAE= 0.50, R<sup>2</sup>=0.29).

#### 342 4. Discussion

343 The effluent DO was significantly higher for unsaturated CWs than saturated  
344 CWs, as expected, as intermittently pulse-loaded subsurface CWs have higher oxygen  
345 transfer rates (Headley et al., 2013). Due to the oxygen release from wetland plant  
346 roots (Brix, 1997), significantly higher effluent DO values were observed in the planted  
347 CWs compared to the unplanted CWs for the same CW design. However, the  
348 operational differences and plant presence did not affect pH and EC, as all CWs  
349 presented similar values. Theoretically, the CWs with higher oxygen transfer rates  
350 should result in higher nutrient (TOC, NH<sub>4</sub>-N and TP) removal efficiencies (Nivala et al.,



351 2007). In the present study, however, nutrient removal was not significantly different  
352 between unsaturated and saturated CWs. This was probably due to the fact that the  
353 synthetic influent was not real wastewater, the nutrients concentrations were low,  
354 and the systems were not limited by oxygen availability. On the other hand, the  
355 significantly higher nutrient removal efficiencies in the planted CWs compared with  
356 unplanted CWs can be attributed to plant uptake (Brix, 1994).

357 Previous studies have reported that saturated surface and subsurface flow CWs  
358 could achieve 45%–98% tebuconazole removal efficiencies at 0.1–100  $\mu\text{g L}^{-1}$  inflow  
359 concentrations in mesocosms and experimental field-scale CWs (Elsaesser et al., 2011;  
360 Elsaesser et al., 2013; Lv et al., 2016c). This present study is, to the best of our  
361 knowledge, the first to describe tebuconazole removal (33%–99.8%) in unsaturated  
362 CWs. Passeport et al. (2013) and Tournebize et al. (2013) observed that field-scale  
363 saturated surface flow CWs (surface area of 1280  $\text{m}^2$ ) could provide up to 36% removal  
364 of tebuconazole from agricultural drainage in different years (2008 and 2009) at a HLR  
365 of 5.6–5.9  $\text{cm d}^{-1}$ . For this HLR regime, it can be estimated from Fig. 4 (and calculated  
366 from Eq. S2) that the tebuconazole removal in the unsaturated and saturated CW  
367 mesocosms, in our study would be 49%–66% and 42%–61%, respectively. Moreover,  
368 if considering a typical HRT in saturated subsurface flow systems of higher than 1 d  
369 (Kadlec & Wallace, 2008), the present mesocosms would provide more than 49% and  
370 62% tebuconazole removal efficiencies in unplanted and planted mesocosms,  
371 respectively. Moreover, the mesocosms planted with *Berula* could each achieve  
372 significantly higher removal efficiencies above 90% under these typical HLR and HRT  
373 regimes.

374 Besides the significantly higher tebuconazole removal efficiencies, the area-  
375 based reaction rate constants were also significantly higher for unsaturated CWs (2.6–  
376 10.9 cm d<sup>-1</sup>) than that for saturated CWs (1.7–7.9 cm d<sup>-1</sup>). These results indicate that  
377 changing CW design from saturated to unsaturated could improve tebuconazole  
378 removal. Due to lack of area-based reaction rate constants available from previous  
379 studies, the values from the present study are the first reference data for future  
380 investigations. Volume-based first order kinetics model can also adequately describe  
381 the tebuconazole degradation in the saturated CWs. The volumetric removal rate  
382 constants (0.8–3.3 d<sup>-1</sup>) are similar to those reported in previous experiments (0.6–2.9  
383 d<sup>-1</sup>) by Lv et al. (2016c) under the same system setup and operating conditions,  
384 revealing a stable capacity for tebuconazole removal during a two-year period. Thus,  
385 both the areal and volumetric degradation rates found in this study can be used as  
386 reference values for future implementation of CWs for tebuconazole contaminated  
387 water treatment.

388 The better tebuconazole removal in unsaturated CWs compared with saturated  
389 CWs may be due to the higher DO levels and different hydraulics, which has also been  
390 observed to generate different microbial communities (Lv et al., 2017a). DO was  
391 positively correlated with tebuconazole removal in the CWs (Fig. 5), which indicates  
392 that tebuconazole degradation is favoured by aerobic conditions. It has been  
393 previously shown that many micropollutants, such as alkylphenols (t-nonylphenol and  
394 4-p-nonylphenol), hormones (estrone, 17 $\beta$ -estradiol and 17 $\alpha$ -ethinylestradiol) and  
395 pharmaceuticals (ibuprofen), show higher depletion through biodegradation under  
396 aerobic conditions than under anaerobic conditions (Abargues et al., 2012; Hijosa-

397 Valsero et al., 2010). In the present study, the DO levels in the saturated CWs were  
398 relatively high ( $>2 \text{ mg L}^{-1}$ ) and higher than expected from the full-scale saturated CW  
399 systems. Thus, tebuconazole removal efficiency in full-scale saturated CWs with lower  
400 DO levels is expected to be lower than reported here.

401 Besides the significant effect of CW design on tebuconazole removal, the  
402 positive effect of the presence of wetland plants in the systems is supported by the  
403 significantly higher tebuconazole removal efficiencies and area-based removal rate  
404 constants for planted CWs than unplanted CWs in both designs. In fact, the presence  
405 of wetland plants also promotes nutrients removal. The different plant species have  
406 different root structure, root oxygen and exudate release, and tebuconazole uptake  
407 ability (Lv et al., 2016b). Tebuconazole removal showed positive correlation with  
408 evapotranspiration indicating that passive uptake of tebuconazole into the plant with  
409 the transpiration stream play a role in removal. Nutrient (TN and  $\text{NH}_4\text{-N}$ ) removal,  
410 however was not correlated with evapotranspiration (Fig. S5), indicating that the  
411 increased activity of the microbial community in the planted systems (Lv et al., 2017a)  
412 promoted nutrients metabolisation. Since tebuconazole removal also showed a  
413 positive correlation in both CW designs with nutrients removal (Fig. S4), it is also  
414 possible that tebuconazole removal can be coupled with the co-metabolisation of  
415 some nutrients. However, metabolisation pathways need to be further investigated  
416 through molecular methods.

417 Tebuconazole removal decreased with increasing HLR, which may be due to the  
418 threshold of the microbial community biodegradation and plant uptake capacity for

419 tebuconazole in both CWs under higher compound loading. The influent  
420 concentration (10 or 100  $\mu\text{g L}^{-1}$ ) did not affect tebuconazole removal in both designs.  
421 This finding is contradictory with a previous study in which the removal efficiency and  
422 removal rate constant of one popularly used pesticide, chlorpyrifos, decreased with  
423 increasing influent concentrations (above 100  $\mu\text{g L}^{-1}$ ) (Prasertsup & Ariyakanon, 2011).  
424 This difference may be attributed to the chemical properties of the compounds or to  
425 the toxic effects of higher concentrations (above 100  $\mu\text{g L}^{-1}$ ). In fact, the present results  
426 are consistent with previous work on tebuconazole (Lv et al., 2016c) where under  
427 saturated conditions and different seasons (summer and winter), influent  
428 concentration did not affect system performance.

429 A mechanistic approach to tebuconazole removal indicates that hydrolysis is  
430 expected to be negligible due its chemical properties (Table S1). Photodegradation can  
431 also be excluded as the mesocosms were operated under subsurface flow conditions.  
432 Tebuconazole sorption to sediment was observed, and the tebuconazole  
433 concentration normalized by SOC was generally lower in unsaturated CWs than in  
434 saturated CWs, which may indicate higher biodegradation rate of tebuconazole in the  
435 unsaturated CW. The generally higher SOC in the planted mesocosms than in the  
436 unplanted mesocosms may be caused by higher microbial biomass growth (Zhang et  
437 al., 2010), favoured by the oxygen translocation and roots exudation capacity of the  
438 plants. Moreover, the estimated substrate sorption (0.7–2.1%) and plant  
439 phytoaccumulation (2.5–12.1%) indicate a limited contribution of both mechanisms  
440 to tebuconazole removal in both designs. In a previous study with saturated  
441 mesocosms, substrate sorption and plant phytoaccumulation were also found to

442 contribute less than 13% to tebuconazole removal (Lv et al., 2016c). The direct role of  
443 plants is still not clear. On one hand, there is a strong positive correlation between  
444 tebuconazole removal and evapotranspiration. On the other hand, no significant  
445 phytoaccumulation is observed, which indicates that tebuconazole could be removed  
446 through metabolisation inside the plant tissue after uptake. Thus, tebuconazole  
447 degradation inside the plant tissue as well as microbial degradation are identified as  
448 the main pathways for tebuconazole depletion in both unsaturated and saturated  
449 CWs. The functionality analysis of the microbial communities from both interstitial  
450 water and biofilm sampled from the present experimental setup (Lv et al., 2017a)  
451 pointed out a strong correlation between tebuconazole removal and the biofilm in  
452 both saturated and unsaturated CWs. Thus, microbial degradation by the substrate-  
453 fixed biofilm seems to be another relevant pathway. Further studies should clarify the  
454 microbial metabolic pathways through which pesticide degradation occurs.

455       Tebuconazole removal in CWs was modelled for the first time in this study. The  
456 different modelling results from the non-linear ANN model and the traditional linear  
457 regression model were compared. The ANN model showed better predictive ability to  
458 forecast tebuconazole removal in both CW designs, even though the simulation results  
459 were relatively good for both the ANN and the traditional linear regression models. In  
460 order to test the broad applicability of the model, it is common to use independent  
461 research data for model validation/prediction (Akratos et al., 2009; Kotti et al., 2016;  
462 Naz et al., 2009). Due to the lack of independent data available, the training and  
463 validation data for unsaturated CWs were from the same study. Thus, the ANN model  
464 only showed slightly better (higher  $R^2$  and lower MAE) forecasting performance than

465 the linear regression model. The better fitting of the ANN was more obvious in the  
466 saturated CWs, for which the data for validation was adapted from a different dataset.  
467 The result is supported by relevant studies reporting that ANN modelling, in  
468 comparison with linear models, can improve the accuracy for predicting COD and  
469 BOD<sub>5</sub> concentrations in wastewater treatment plant (Abyaneh, 2014) and total  
470 nitrogen dynamics in streams (Amiri & Nakane, 2009). Thus, the application of a  
471 nonlinear algorithm, such as the ANN-MLP, and the multi fold training and validation  
472 schemes we adopted improved the accuracy of the forecasting model, by covering a  
473 substantial part of the non-linear mechanisms and factors influencing tebuconazole  
474 removal. The present result demonstrates that tebuconazole degradation is a complex  
475 process and cannot be easily predicted by simple linear regressions. Therefore, the  
476 modelling results show that ANN was more stable and can be trusted to simulate and  
477 predict tebuconazole removal, which also deserves to be utilized for other studies on  
478 removal of emerging organic pollutants.

## 479 5. Conclusions

480 The present study showed significantly higher tebuconazole removal in unsaturated  
481 CWs than saturated CWs, supporting the high potential of unsaturated CWs for the  
482 treatment of tebuconazole contaminated water. Tebuconazole removal was fitted  
483 with an area-based first order kinetics model in both unsaturated and saturated CWs.  
484 The obtained degradation rates can be used as reference data for future applications  
485 of CWs in the treatment of tebuconazole contaminated water.

486 Tebuconazole sorption by the substrate (0.7–2.1%) and plant  
487 phytoaccumulation (2.5–12.1%) made limited contributions to tebuconazole removal.  
488 Thus, biodegradation and plant metabolisation were the main removal pathways in  
489 both CW designs. The main factors influencing tebuconazole removal in the  
490 mesocosms were system design, plant presence and species, and HLR. Plants had a  
491 positive effect, while increasing HLR had a negative effect. Influent concentration did  
492 not show a significant effect on tebuconazole removal. Tebuconazole removal was  
493 also correlated with dissolved oxygen and removal of other pollutants, indicating that  
494 tebuconazole degradation could be coupled with co-metabolisation processes.  
495 Artificial neural network (ANN) modelling was demonstrated to be a more accurate  
496 model than linear regression modelling to simulate tebuconazole removal in CWs.

497

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