

1 **Microalgae and bacteria dynamics in high rate algal ponds** 2 **based on modelling results: long-term application of** 3 **BIO_ALGAE model**

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10 **1. Introduction**

11 High rate algal pond (HRAP) technology for wastewater treatment was
12 developed in California by Prof. Oswald in the 1950s as an alternative to conventional
13 waste stabilisation ponds (WPS) (Oswald and Gotaas, 1957). The smaller footprint of
14 HRAP systems coupled with the benefit of production of valuable products (e.g.
15 biofuels, bioplastics) from microalgae feedstock makes them more attractive over WPS
16 (García et al. 2000a; Faleschini et al. 2012).

17 HRAPs are based on microalgae and bacteria interactions in wastewater exposed
18 to light. Microalgae photosynthesis provides oxygen necessary for the degradation of
19 organic compounds present in wastewater by aerobic bacteria. During bacterial
20 oxidation of organic matter, carbon dioxide (CO₂) is produced and is available for both
21 photosynthesis and nitrification (Oswald 1988). The many processes that occur in
22 microalgae-bacteria systems are quite difficult to control (García et al. 2006; Awuah
23 2006; Fuentes et al. 2016). Moreover, these processes depend on ever-changing
24 environmental variables such as solar radiation and temperature.

25 Although these systems have been studied for many years, still today the
26 physical, chemical and biochemical reactions that occur in microalgae-bacteria systems

27 are less well known than processes in conventional technologies, such as activated
28 sludge. In fact, it is still very challenging to understand which are the main factors
29 affecting microorganisms growth and production (i.e. microalgae and bacteria), and
30 how their interactions affect the relative proportion of microorganisms. Recently,
31 variations of biomass production over a year in pilot-scale HRAPs were experimentally
32 evaluated by Mehrabadi et al. (2016). These authors observed that changes in
33 microalgae concentration were clearly linked to seasonal fluctuations in temperature and
34 light intensity in the absence of nutrient limitation. Other studies have shown that
35 HRAP operating conditions play an important role on biomass composition, and of
36 course the efficiency for removing pollutants. In a study conducted by Park and Craggs
37 (2011), hydraulic retention time (HRT) clearly influenced microalgae proportion
38 dynamics. A low hydraulic retention time (HRT, 2 days) yielded much more microalgae
39 biomass than bacteria (80% of average of total biomass), while high HRT (8 days) had
40 lower microalgae proportion (56% of average of total biomass). Note that these authors
41 estimated microalgae proportion indirectly by measurements of chlorophyll-a
42 concentration. At present time is not trivial to have a direct measure of microalgae and
43 bacteria proportion in such mixed cultures.

44 Mathematical models have proven to be useful tools to understand and optimize
45 the functioning of biological wastewater treatment systems, including microalgae-
46 bacteria systems (Park and Craggs 2011; Zhou et al. 2014). Solimeno et al. (2017a)
47 developed the mechanistic BIO_ALGAE model to understand the internal functioning
48 of complex microalgae-bacteria systems. One relevant feature of the model is that it
49 allows microbial biomass concentration prediction, and thus evaluation of the relative
50 proportions of microorganisms (Solimeno et al. 2015; 2017a; 2017b).

51 The River Water Quality Model 1 (RWQM1) (Reichert et al. 2001) and the
52 ASM3 model (Iacopozzi et al. 2007) (International Water Association, IWA) were
53 selected to describe microalgae and bacteria processes, respectively. Inorganic carbon as
54 a limiting substrate for the growth of microalgae is one of the major innovative
55 processes of BIO_ALGAE. Moreover, temperature, photorespiration, pH dynamics,
56 solar radiation, light attenuation and rate of transfer of gases to the atmosphere are
57 considered main limiting factors for microalgae growth. BIO_ALGAE was
58 implemented in the COMSOL MultiphysicsTM software, which solves differential
59 model equations using the finite elements method (FEM). This model was previously
60 calibrated and validated with high quality experimental data from duplicated pilot
61 HRAPs receiving real wastewater (Solimeno et al. 2017a). Calibration was conducted
62 by adjusting the following 6 parameters selected after a Morris's sensitivity analysis: 3
63 parameters related to microalgae and heterotrophic bacteria specific growth rate and
64 decay of heterotrophic bacteria, and 3 parameters related to the transfer of gases to the
65 atmosphere. These parameters were carefully calibrated and validated in our previous
66 work comparing field data over 4 intensive days of experiments to predict daily
67 fluctuations of the components in the ponds, and the relative proportions of microalgae
68 and bacteria in a short-time scale. A long-term validation is essential to demonstrate (1)
69 the capacity of the model to predict seasonal variations of microalgae and bacteria
70 biomass, and (2) the effect of different HRT operating strategies on the HRAP
71 performance, biomass production and biomass proportions.

72 Therefore, the aim of the present study is to validate the BIO_ALGAE model
73 with experimental data from a pilot HRAP gathered during two different seasons
74 (summer and winter), and operating at different HRT (4 and 8 days). Moreover, the

75 potential of the model is demonstrated by means of practical study cases in which
76 microalgae production, the relative proportion of microalgae and bacteria, and the
77 ammonium removal efficiency were compared over an annual cycle. Hence, the purpose
78 of this study was to study HRT and season effects on HRAP performance: microalgae
79 production, the relative proportion of microalgae to bacteria, and ammonium removal
80 efficiency.

81

82 **2. Material and methods**

83 *2.1. Experimental data*

84 The data used for simulations were obtained from previous studies conducted by
85 the authors in a pilot HRAP (García et al. 2000a; 2002). A detailed description of the
86 system can be found in these studies. In brief, the pilot HRAP was installed outdoors on
87 the roof of the Group of Environmental Engineering and Microbiology (GEMMA)
88 building (Universitat Politècnica de Catalunya-BarcelonaTech, Barcelona, Spain,
89 latitude: 41° 23' 24.7380" N; longitude: 2° 9' 14.4252" E). The data used in this study
90 were collected from July 1993 to October 1993 (Period I), and from November 1993 to
91 February 1994 (Period II), corresponding to low and high HRT of the pilot (4 and 8
92 days, referred as HRAP_{4d} and HRAP_{8d}, respectively). In practice, low HRT is used in
93 warmer periods (summer-autumn), while high HRT are used in colder periods (autumn-
94 winter) to maintain stable contaminant removal efficiencies (García et al. 2006).

95 The pilot HRAP was a typical race track built in PVC with a water surface area
96 of 1.54 m² and a water depth of 0.34 m, and a nominal volume of 0.47 m³ (Fig. 1). A
97 single paddlewheel was set to provide a rotational speed of 5 rpm and a mid-channel

98 velocity of approximately 9 cm s^{-1} , avoiding biomass settling. HRAP received primary
99 treated urban wastewater from a near street sewer, which was continuously pumped to
100 the pond. Primary treatment was conducted in a 0.5 m^3 storage tank. HRT of the HRAP
101 was controlled by wastewater flow.

102 Samples of HRAP influent, HRAP mixed liquor (identical to HRAPs effluent,
103 because of almost perfect complete mixing) were taken once a week, at 2:00 PM ± 1
104 hour. Description of the methods used for analyses can be found in García et al. (2000a,
105 2002). Water temperature, pH and DO (dissolved oxygen) were measured weekly, at
106 9:00 AM ± 1 and at 2:00 PM ± 1 hour. Maximum and minimum water temperature and
107 irradiance recorded over the two periods investigated are shown in Table 1a.

108 Irradiance and air temperature were obtained from a nearby meteorological
109 station.

110

111 *2.2. Model implementation*

112

113 Simulations were conducted using the BIO_ALGAE model. A detailed
114 description of the components, the biokinetic processes, and the chemical and physical
115 equations were presented in our previous works (Solimeno et al. 2015; 2017a; 2017b).
116 To simplify presentation of the simulation results, Tables S1 and S2 in Supplementary
117 Material (SM) present the biokinetic processes and the matrix of stoichiometric
118 parameters. Values of biokinetic, physical and chemical parameters are shown in SM,
119 Tables S3-S4. Mathematical expressions of the stoichiometric coefficients of each
120 process are also shown in SM, Table S5.

121 The model was implemented in COMSOL Multiphysics™ v5.1 software. A
122 simplified 1D domain was used to represent a vertical section of the pilot HRAP.
123 Assuming that each section behaves similarly due to perfect mixing of culture medium,
124 this reasonable simplification allowed a reduction of computational cost. The culture
125 was mixed with the paddlewheel, which ensured almost complete stirred reactor
126 behaviour due to the small volume of the high rate pond in relation with the big size of
127 the blades of the paddlewheel.

128 According to the Beer-Lambert law, irradiance decays exponentially as it passes
129 through the HRAP mixed liquor. Therefore, a depth-averaged irradiance I [$\mu\text{mol m}^{-2} \text{s}^{-1}$]
130 was used to represent irradiance at any point of the pond (Solimeno et al. 2017a). The
131 penetration pathway corresponded to the depth of the HRAP (0.3 m).

132

133 *2.3. Validation procedure*

134

135 BIO_ALGAE includes 93 parameters describing microalgae, bacteria, physical
136 and chemical processes (Tables S3-S4, SM). Most of these parameters were obtained
137 from the existing RWQM1 [15], ASM1, and ASM3 models (Henze et al. 2000;
138 Iacopozzi et al. 2007). Parameters related to temperature, photorespiration, carbon
139 limitation and light attenuation were obtained from other literature cited in SM (Table
140 S3-S4). Morris uncertainty method was applied in our previously work (Solimeno et al.
141 2016; 2017a) to identify the parameters which had the greatest influence on the
142 simulation response (Morris 1991). Results from our previous works have shown that
143 the values of maximum growth rate of microalgae (μ_{ALG}), the maximum growth rate and
144 the decay of heterotrophic bacteria (μ_{H} and $k_{\text{death,H}}$) and the parameters related to the

145 transfer of gases to the atmosphere (K_{a,O_2} , K_{a,CO_2} and K_{a,NH_3}), were very sensitive and
146 need to be calibrated in each application of the model (Solimeno et al. 2015; 2016).

147 In the present work, these parameters were set based on a previous calibration
148 and validation effort using experimental data from duplicate HRAPs located at the
149 Delhi wastewater pond treatment plant (California) during 4 days of experiments
150 (Solimeno et al. 2017a). Results of this earlier effort indicated that the model was able
151 to match experimental data accurately. Influent HRAP average concentrations observed
152 in each experimental period were used as constant input values to run simulations
153 (Table 1b). Influent concentrations of nitrate and nitrite were lower than the analytical
154 method's detection limit and therefore considered to be zero in the input for the model.
155 Ammonium nitrogen comprised almost 90% of dissolved Kjeldahl nitrogen (García et
156 al. 2000a), and therefore the concentration of organic nitrogen present in the influent
157 wastewater was omitted from the model.

158 Fractions of influent chemical oxygen demand (COD) were estimated from
159 rational values for primary effluents in Activated Sludge Model No1 (ASM1) (Henze et
160 al. 2000). Accordingly, the proportion of each fraction was defined as: 22% readily
161 biodegradable soluble organic matter (S_S), 50% slowly biodegradable particulate
162 organic matter X_S , 10% inert soluble organic matter (S_I), 8% inert particulate organic
163 matter (X_I) and 10% heterotrophic bacteria (X_H). In the present work microalgae and
164 bacteria biomass are transformed from COD to total suspended solids (TSS) assuming a
165 ratio $COD/TSS = 0.80$ (Khorasandi et al. 2014) in order to compare experimental and
166 simulation results. Note that some authors apply different ratios, for example, Von
167 Sperling (2007) found values of 1–1.5 in waste stabilization pond effluents.

168 The concentrations of components in the mixed liquor of the HRAP measured at
169 the beginning of the two experimental periods are shown and described in Table 2.
170 Unfortunately, the concentration of each particulate component (X_{ALG} , X_S , X_I , X_H , X_{AOB}
171 and X_{NOB}) in the mixed liquor was not known (where X_{ALG} is microalgae concentrations
172 and X_{AOB} , X_{NOB} are ammonium and nitrite oxidizing bacteria, respectively).

173 Therefore, initial ratio of X_{ALG} , X_S , X_I , X_H , X_{AOB} and X_{NOB} concentrations were
174 quantified from initial TSS value (from M1 pond) based on previous simulation tests.

175 This assumption also provided an initial relationship between pH, dissolved
176 oxygen, and nutrients (i.e. nitrogen and carbon).

177 Validation was performed by comparing measured data with simulation patterns
178 using graphs of the two periods (with different HRT). Tested components during
179 validation were: pH, dissolved oxygen (S_{O_2}), bicarbonate (S_{HCO_3}), ammonium (S_{NH_4}),
180 nitrate (S_{NO_3}), nitrite (S_{NO_2}) and TSS. Model data were compared to experimental data
181 by the root mean square error (RMSE).

182

183 *2.4. Case studies: relative proportion of microalgae and bacteria, and biomass*
184 *production forecasting over a year cycle*

185

186 Practical case studies were conducted to evaluate the variations in biomass
187 production and the relative proportion of microalgae and bacteria over a year cycle
188 (from January to December). The experiments were conducted over an annual cycle in
189 order to investigate the influence of different HRT operating strategies and seasonal
190 variations of temperature and irradiance on the relative proportion of microalgae and
191 bacteria, and biomass production over a year cycle.

192 In these studies, we simulated the evolution of microalgae, bacteria and TSS
193 concentrations starting from the initial mixed liquor concentration used for the
194 validation of the model at the beginning of the month of February, and using average
195 influent wastewater concentration showed in Table 3. In addition, ammonium and
196 ammonia concentration ($S_{NH4}+S_{NH3}$) were evaluated as indicators of removal efficiency
197 because they are very sensitive to changes in the environmental conditions in HRAPs.
198 According to European standard (European Council 1991), HRAP performance is
199 suitable when the concentration of ammonium and ammonia in the effluent is lower
200 than 10 gN m^{-3} . Three scenarios were evaluated: 1) the HRAP operating at 4-day HRT
201 (HRAP_{4d}) over the whole year; 2) the HRAP operating at 8-day HRT (HRAP_{8d}) over
202 the whole year, and 3) the HRAP operating with different HRT, from April to
203 September at 4-day HRT and from October to March at 8-day HRT (HRAP_{8-4-8d}).

204 The standard method of changing HRT during different seasons to maintain
205 removal efficiency was investigated according to the results of previous experimental
206 research carried out in Barcelona (García et al. 2000b). Water temperature data taken
207 weekly at 9:00 AM ± 1 hour, and at 2:00 PM ± 1 hour measured over the one-year
208 monitoring period, and irradiance data from the meteorological station of physical
209 department of University of Barcelona (around 2 km far from the pilot HRAP) were
210 used to run simulations for study cases.

211

212 **3. Results and discussion**

213

214 In this section, first of all, air temperature and irradiance changes over the course
215 of all experiments were showed in Fig.2. As can be seen temperature and irradiance

216 were greater in Period I than in Period II. Also in Period I the general trend of
217 temperature and irradiance was a progressive decrease from July to October, while
218 changes in Period II were more subtle. Following, performance assessment of the model
219 in predicting experimental data under different HRTs were presented.

220

221 *3.1. HRAP_{4d} validation (Period I)*

222

223 Figure 3 shows the results of the validation in the HRAP_{4d} from July to October.
224 Simulations were able to follow measured pH and S_{O2} trends during the whole
225 experimental period (Fig. 3a-b). As can be seen, both variables have a daily oscillations
226 pattern due mostly to microalgae photosynthetic activity. This trend is in agreement
227 with previous simulation results obtained during calibration in our previous study
228 (Solimeno et al. 2017a), and also with previous experimental studies (García et al.
229 2006). Simulated daily minimum and maximum values were generally higher and lower
230 than values measured at 9:00 AM and 2:00 PM, respectively, because the peaks of
231 microalgae activity do not necessary coincide with these hours. From simulations, pH
232 values ranged from 7.4 to 10.1, with an average of 8.5, while S_{O2} concentration ranged
233 from 0 gO₂ m⁻³ to 28.1 gO₂ m⁻³, with an average of 11.2 gO₂ m⁻³. It is possible to see
234 how at the end of this period daily fluctuations of pH and S_{O2} were slightly smoother
235 than at the beginning of the study. At night, S_{O2} concentration decreased to be usually
236 less than 5 gO₂ m⁻³, and even in some few cases almost 0 due to the lack of
237 photosynthesis and the intense microbial respiration.

238 Simulations followed the trend observed for measured S_{HCO3}, S_{NH4}, S_{NO3} and
239 S_{NO2} with different degree of success (Fig. 3c-f). Simulated S_{HCO3} and S_{NH4} curves

240 matched quite well with the experimental data, and present a clear oscillation pattern
241 mostly related to photosynthesis, with lower values of both variables during daytime.
242 Microalgae grow during daytime using bicarbonate as carbon source, and subsequently
243 pH raises favoring conversion of ammonium to ammonia, which is lost in part through
244 volatilization. Moreover, microalgae uptake also contributes to ammonium decrease
245 during daytime. These trends are also in agreement with simulation results carried out
246 during calibration in our previous study (Solimeno et al. 2017a), and with previous
247 experimental studies (García et al. 2006). Daily fluctuations of S_{HCO_3} and S_{NH_4} tend to
248 soften towards the end of Period I (Fig 3c-d); the same pattern was observed for pH and
249 S_{O_2} . This is indicative of a lower photosynthetic activity due to decrease of incident
250 irradiance and temperature.

251 As can be seen in Figure [S2d3d](#), S_{NH_4} simulated concentrations were relatively
252 low and constant during July and August (the first 60 days), and increased from mid-
253 September, corresponding with the decrease in incident irradiance and temperature. This
254 also corresponds with the lower overall microalgae activity. Higher values of S_{NO_3} and
255 S_{NO_2} were observed towards the end of the period, when S_{NH_4} concentration was also
256 higher. The model was able to simulate these trends described for S_{NO_3} and S_{NO_2} quite
257 well, and it can be seen that photosynthesis influences these compounds much less due
258 to the much lower daily oscillation trends.

259 Simulated TSS concentration fits experimental data with a good degree of
260 accuracy (Fig. 3g). The model is able to predict curves of microalgal (X_{ALG}) and
261 bacterial biomass concentrations (X_{H} , X_{AOB} and X_{NOB}) (Fig. 3h-i). Simulated X_{ALG}
262 concentration exhibits an oscillation trend, reflecting microalgae growth during daytime
263 (crest) and decay at night (trough); heterotrophic bacteria concentration simulations do

264 not exhibit this pattern. During the period, X_{ALG} gradually decreased, following the
265 decreases in irradiance and temperature, while X_H remained relatively constant. In July
266 and August (the first 60 days), high irradiance and temperature produced a high
267 photosynthetic activity, which at the same time produced high daily peaks of S_{O_2} (often
268 greater than $25 \text{ gO}_2 \text{ m}^{-3}$). Concentrations of DO in the culture above 250% air saturation
269 ($22.6 \text{ gO}_2 \text{ m}^{-3}$) can dangerously inhibit microalgae activity (Costache et al. 2013).
270 Subsequently these peaks seem to limit microalgae growth due to photorespiration. As
271 can be seen in Figure 4a, the photorespiration factor ($f_{PR}(O_2)$) reduced microalgae
272 growth from 20 to 40% (values of the factor from 0.8 to 0.6, respectively). Thus, excess
273 of oxygen caused less microalgae production that could had been avoided with
274 improved oxygen transfer to the atmosphere. A detailed description of photorespiration
275 factor is provided in SM. Also the drop in temperature from mid-September (day 65)
276 had impact on X_{ALG} , causing a reduction of growth between 10 to 20% through the
277 thermic photosynthetic factor (f_{T-FS}) (Figure 4b) (see also SM for a detailed description
278 of this factor).

279 Nitrifying bacteria (X_{AOB} and X_{NOB}) concentration was very low in comparison
280 to X_H . This observation has already been reported in previous simulation studies, and
281 for other types of wastewater treatment systems (Krasnits et al. 2009; Samsó and García
282 2013; Solimeno et al. 2017a)

283 Altogether, simulation results have predicted that much of the organic matter
284 present in the mixed liquor corresponds to X_{ALG} (55% in average of TSS) and X_H (26%
285 in average of TSS). X_{AOB} and X_{NOB} are comparatively very low (0.35%), and the
286 remaining solids are attributable to X_S (5.5%) and X_I (13.2%).

287

3.2. HRAP_{8d} validation (Period II)

288

289

290 Figure 5 shows the results of the validation in the HRAP_{8d} from November to
291 February. Again, the model exhibited the oscillation trend for pH and S_{O2} during the
292 whole experimental period (Fig. 5a-b), with lower values in comparison to Period I. The
293 degree of model fit was slightly lower than in Period I. Simulation results indicated that
294 pH values ranged from 7.2 to 9.6, with an average of 8.1, while S_{O2} concentrations
295 ranged from 0.9 gO₂ m⁻³ to 20 gO₂ m⁻³ with an average of 11 gO₂ m⁻³. Daily
296 fluctuations of pH and S_{O2} were shorter than in Period 1. In comparison to Period 1,
297 nighttime S_{O2} concentration rarely decreased to below 5 gO₂ m⁻³ due to the lower
298 overall microbial respiration and lower temperature (which increased transfer from the
299 atmosphere to the mixed liquor).

300 Model validation results for HRAP_{4d} conditions were comparable for S_{HCO3}, S_{NH4}
301 and S_{NO2} data. Conversely, simulated S_{NO3} concentration did not match experimental
302 data as closely as in Period I. Simulated S_{HCO3} and S_{NH4} exhibited the oscillation pattern
303 already mentioned in Period I, but with shorter daily fluctuations, similar to the last part
304 of Period I (when irradiance and temperature decreased). This is indicative of a lower
305 overall photosynthetic activity in comparison to Period I. Much higher values of S_{NO2}
306 were observed towards the end of Period II, when S_{NH4} was higher.

307 As can be seen in Figure 5g, the model was able to simulate TSS concentration
308 with a good degree of accuracy. With respect to HRAP_{4d}, predicted concentrations of
309 TSS, X_{ALG}, X_H, X_{AOB} and X_{NOB} were lower than the data (although HRT was higher
310 than in Period I) (Fig. 5g-i). Lower X_{ALG} concentration was mostly due to the
311 temperature (and to irradiance to a lesser extent). Figure [S4j-5j](#) shows reduction of

312 growth from 10 to 30% through the thermic photosynthetic factor (f_{T_FS}), and as can be
313 seen was much lower in Period II than in Period I (compare with Fig. 4b). Although
314 COD influent in Period II ($195 \pm 50 \text{ g O}_2 \text{ m}^{-3}$) was slightly higher than the in Period I
315 ($180 \pm 84 \text{ gO}_2 \text{ m}^{-3}$), X_H concentration was lower due to the lower temperature. Bacteria
316 thermal factor (f_{T_MB}) shows reduction of X_H growth from 40 to 60% (Fig. 5j).

317 X_{ALG} was higher (58% in average of TSS) than X_H (22% in average of TSS).
318 X_{AOB} and X_{NOB} biomass was comparatively much lower (2.4%), but still higher than the
319 estimated value for Period I. The remaining solids were attributable to X_S (6%) and X_I
320 (11.6%).

321 A comparative evaluation indicates that HRT should be high enough to
322 guarantee treatment performance and to prevent wash-out effects (Larsdotter 2006).
323 Simulation results indicated that a higher HRT during the winter would probably be
324 necessary than during the summer as result of the lower growth rate of microalgae as
325 well as bacteria.

326 Table 4 presents the RMSE values obtained comparing the experimental data
327 with the model simulations obtained from the validation of the model for each period
328 investigated. Values of RMSE near zero indicate that the model fits experimental data
329 well [28]. RMSE values for pH, dissolved oxygen, bicarbonate, nitrogen species and
330 total suspended solids are in good accordance with the RMSE values calculated during
331 the calibration of the model (Solimeno et al. 2017a).

332
333 *3.3. Study case: relative proportion of microalgae and bacteria, biomass*
334 *production and ammonium removal efficiency of HRAP_{4d} over a year cycle.*

335

336 In this case study, the relative proportion of microalgae and bacteria, and the
337 production of microalgae are predicted with the HRAP operating continuously with 4-
338 day HRT. Figure 6a presents simulations of X_{ALG} , X_H and TSS concentrations. X_{ALG}
339 was different between seasons, being lower in colder months (from November to
340 March) and higher in warmer months (from April to October). The photorespiration
341 effect limited microalgae growth during the warmer months, keeping the concentrations
342 around 225 gTSS m^{-3} . X_H concentration was quite constant over the year due to the
343 constant influent wastewater features. As can be seen from Figure 6b, microalgae
344 proportion with respect to bacteria increased from April to October up to 60-75% and
345 dropped down to 27-33% from November to March. Trends suggested by these results
346 are in accordance with the experimental studied by Park and Craggs (2011). In their
347 study, the proportion of X_{ALG} (estimated indirectly) in the microalgae-bacteria biomass
348 of an HRAP operating at 4-day HRT with CO_2 addition in summer was estimated to be
349 around 80%.

350 X_{ALG} and TSS production are shown in Figure 6c. Predictions indicate that
351 during the warmer month with a 4-day HRT, it is possible to reach up to $20 \text{ gTSS m}^{-2}\text{d}^{-1}$
352 of X_{ALG} production. Although pH values in summer are very high (> 9 , Fig. 7a), the
353 model indicates that microalgae are not carbon limited (carbon Monod function= 0.99).
354 Furthermore, S_{O_2} in excess limits microalgae growth through photorespiration (average
355 $f_{\text{PR}}(\text{S}_{\text{O}_2}) = 0.62$ in summer) (Fig. 7a).

356 Ammoniacal nitrogen concentration (sum of ammonium plus ammonia
357 $\text{S}_{\text{NH}_4} + \text{S}_{\text{NH}_3}$, from now on “ammonium”) was used as an indicator of efficiency of HRAP
358 treatment wastewater. As can be seen in Figure 7b, S_{NH_4} has a clear seasonal pattern. In

359 colder months, only approximately an average of 40% of the influent ($49 \text{ g N-NH}_4 \text{ m}^{-3}$)
360 is removed, while in warmer months average removal rate goes up to 90%.

361

362 *3.4. Study case: relative proportion of microalgae and bacteria, biomass*
363 *production and ammonium removal efficiency of HRAP_{8d} over a year cycle.*

364

365 In this case study, the HRAP is continuously operated with 8-day HRT. Figure
366 8a presents simulations of X_{ALG} , X_{H} and TSS concentrations. X_{ALG} changed less over
367 the year in comparison to the HRAP_{4d}. X_{H} concentrations were quite constant over the
368 year, and had similar concentrations to HRAP_{4d}. As can be seen from Figure 8b,
369 microalgae proportion in comparison to bacteria was higher from April to October (76-
370 78%), and slightly dropped down to 65-68% from November to March. In this case
371 study, X_{ALG} were more abundant than X_{H} over the entire year. These trends are not in
372 agreement with the experimental study by Park and Craggs (2011), where the
373 proportion of microalgae (estimated indirectly) of an HRAP operating at 8-days HRT
374 with CO₂ addition in summer was around 55.6%, much lower than in a 4-day HRT. Park
375 and Craggs (2011) indicated that microalgae growth was limited due to low light
376 availability in the pond. Irradiance was attenuated by the high biomass concentration up
377 to 430 g VSS m^{-3} , while in our numerical experiment the biomass concentration in term
378 of TSS is maintained below of 400 g TSS m^{-3} .

379 X_{ALG} and TSS production are compared in Figure 8c. Predictions indicate that
380 with an 8-day HRT, it is possible to reach up to $10.6 \text{ g TSS m}^{-2}\text{d}^{-1}$ of X_{ALG} production in
381 warmer months, which resulted 50% lower than X_{ALG} production predicted in 4-day
382 HRT. In this case study pH is also very high in summer (> 9 , Fig. 9a), however the

383 model indicates that microalgae are not carbon limited. Again, S_{O_2} in excess limits
384 microalgae growth through photorespiration (average $f_{PR}(S_{O_2}) = 0.42$ in summer) (Fig.
385 9b). Excess of S_{O_2} was much higher than $HRAP_{4d}$. With an 8-day HRT the mass influent
386 organic matter concentration in the pond is reduced, therefore oxygen demand by X_H to
387 oxidize organic matter is lower than oxygen produced by microalgae during the
388 photosynthesis due to the high concentration of X_{ALG} (260 gTSS m^{-3} .in summer). As
389 can be seen in Figure 9c, the model prediction indicated that average ammonium
390 removal rate goes up to 98% of the influent ($49 \text{ g N}_{NH_4} \text{ m}^{-3}$) over the whole year.

391

392 3.5. Study case: ~~enhanced~~optimization of microalgae production and
393 ammonium removal efficiency over a year cycle

394

395 In this case study, the HRAP is operated with changing HRT. Higher HRT (8-
396 day) was used in the colder months (from October to March) and lower HRT (4-day) in
397 the warmer months (from April to September) ($HRAP_{8-4-8d}$). This strategy was selected
398 from results obtained in the previous case studies.

399 Figure 10a presents simulations of X_{ALG} , X_H and TSS concentrations. X_{ALG} and
400 microalgae/bacteria proportion (not shown, but can be deduced) changed slightly over
401 the year in comparison by the other two case studies. X_{ALG} production was enhanced
402 (Fig. 10b). With $HRAP_{8-4-8d}$, the production increased to 30% and 35% in $HRAP_{4d}$ and
403 $HRAP_{8d}$, respectively.

404 As can be seen in Figure 10c the model prediction indicated that average
405 removal rate of ammonium goes up to 92% ($49 \text{ g N}_{NH_4} \text{ m}^{-3}$) over the whole year.

406

407 **4. Conclusion**

408 In this work the BIO_ALGAE model was validated for a long-term period using
409 data from a pilot HRAP operating at different HRT (4 and 8 days) during the summer
410 and winter seasons (respectively). The model accurately matched HRAP dynamics
411 using the calibrated values of 6 parameters obtained in previous work by the authors.

412 BIO_ALGAE has demonstrated by means of practical study cases to be a useful
413 tool to understand microalgae and bacteria interactions in wastewater treatment, and in
414 particular to study the effect of different HRAP operating strategies on the relative
415 proportion of microalgae and bacteria, biomass production, and removal of ammonium.

416 Moreover, the model could be used to optimize biomass production. Moreover, the
417 ~~model was able to enhanced biomass production.~~

418

419 E-supplementary data of this work can be found in online version of the paper

420

421 **Acknowledgements**

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423 supported by the Spanish Ministry of Economy and Competitiveness through the project
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425 AP2012-6062 scholarship provided by the Spanish Ministry of Education and Science.

426

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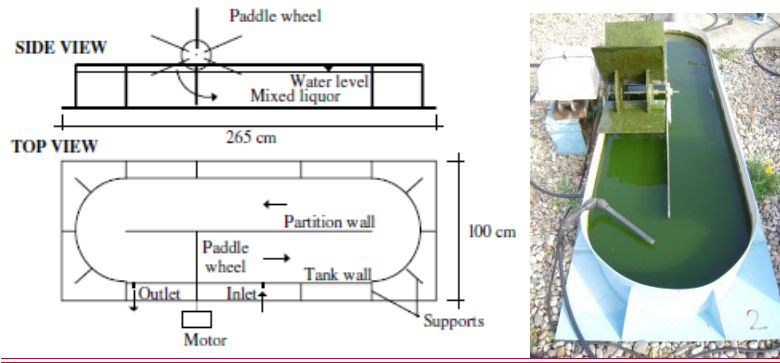
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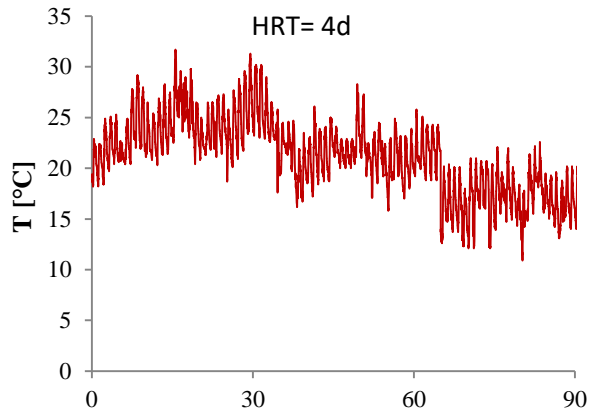


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551

552 **Fig. 1.** Diagram of top and side views of the pilot HRAP on the left and a picture on the right. The system was
 553 located roof of the Group of Environmental Engineering and Microbiology (GEMMA) building (Universitat
 554 Politécnica de Catalunya-BarcelonaTech, Barcelona, Spain).

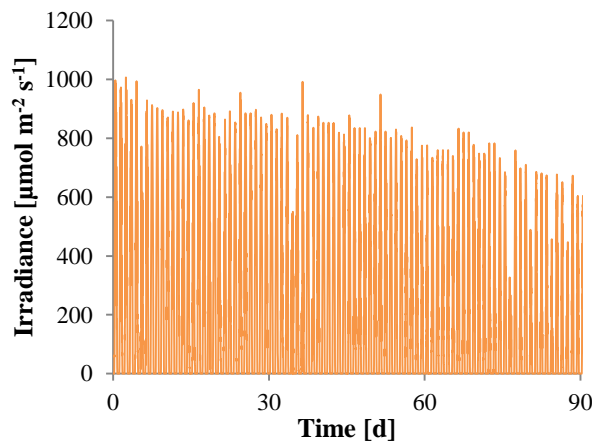
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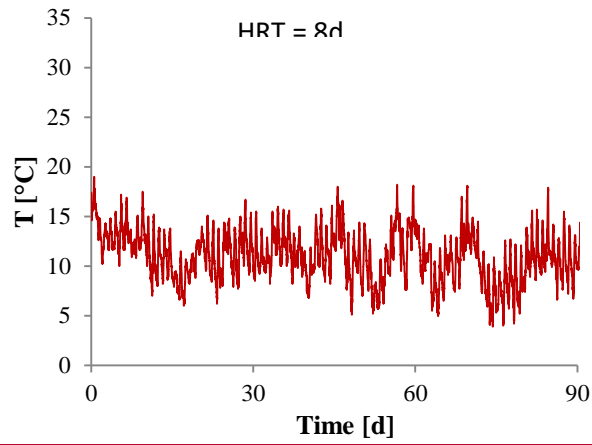
HRT= 4d

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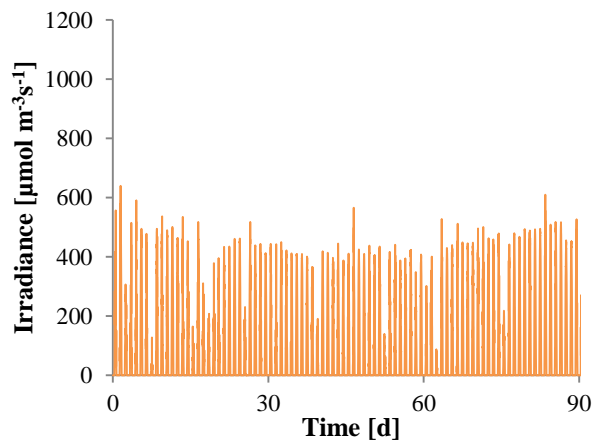
a) b)



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c) _____ c)



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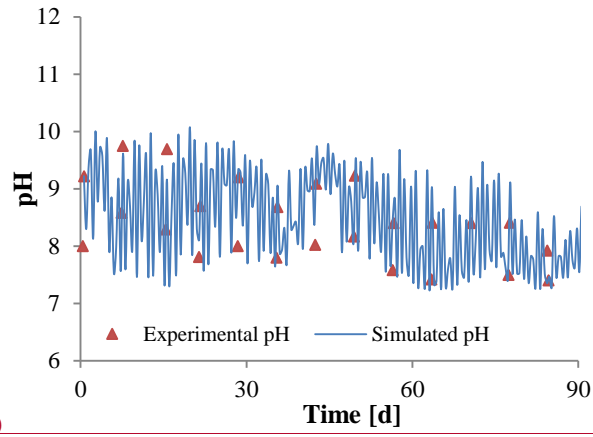
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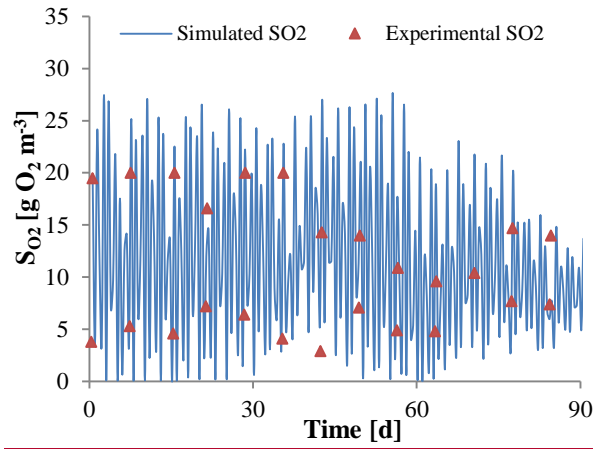
Fig. 2. Changes in air temperature and irradiance in Barcelona over Period I (July 21st – October 14th, 1993, HRAP_{4d}) (a, b), and over Period II (November 10th, 1993 – February 8th, 1994, HRAP_{8d}) (c, d). HRT in each period also shown.

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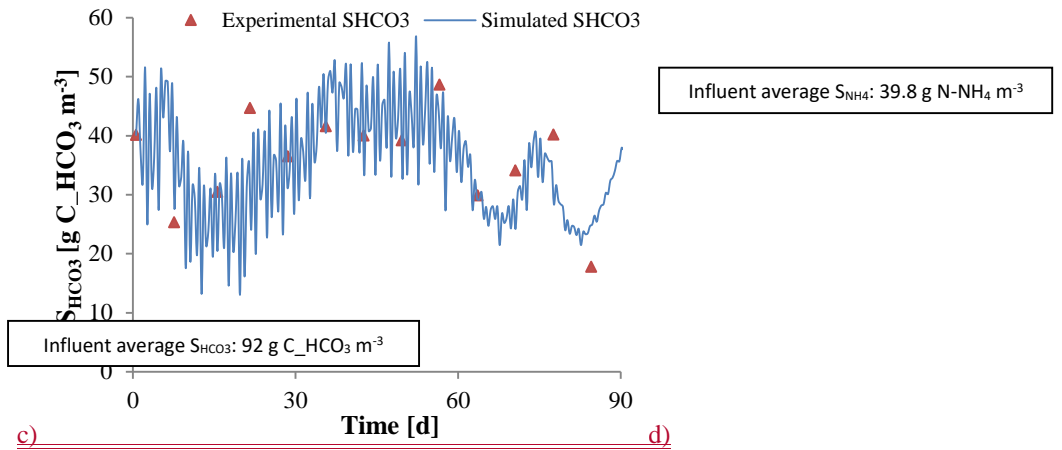


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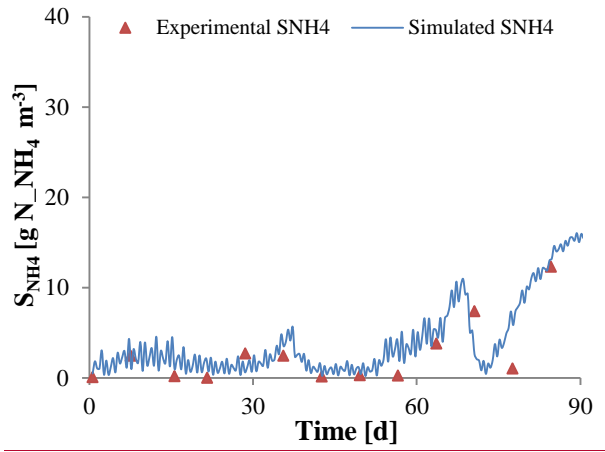
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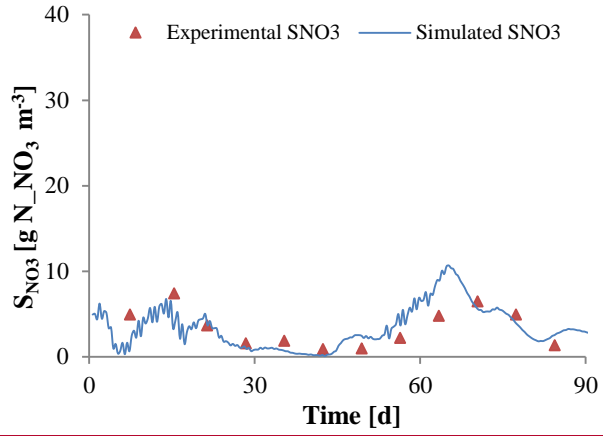
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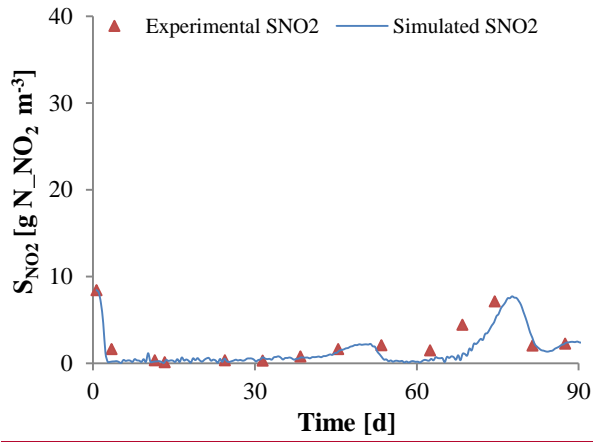
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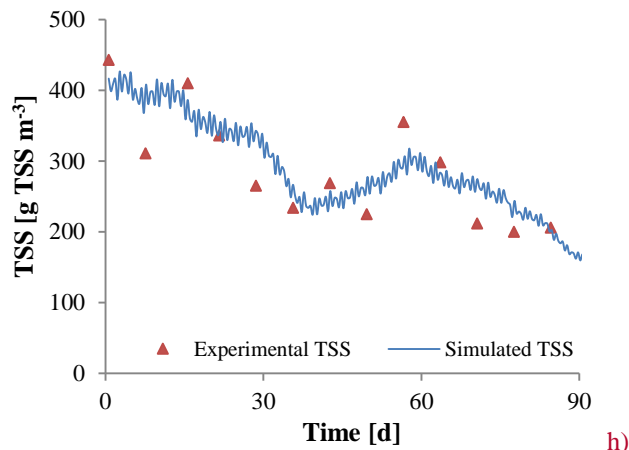
e)

f)

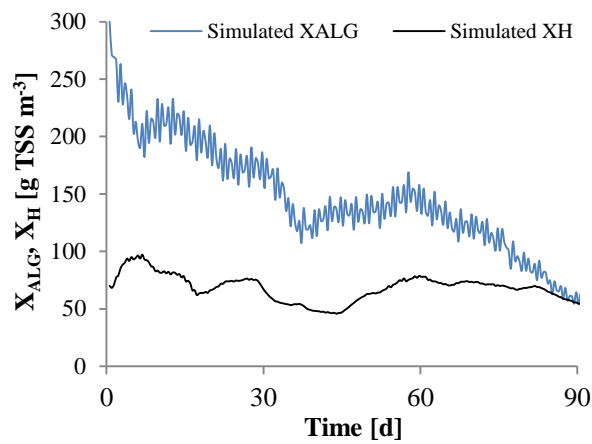
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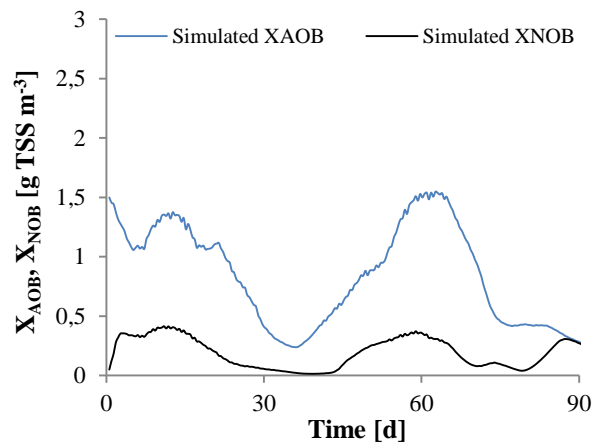
g)



h)



Time [d]

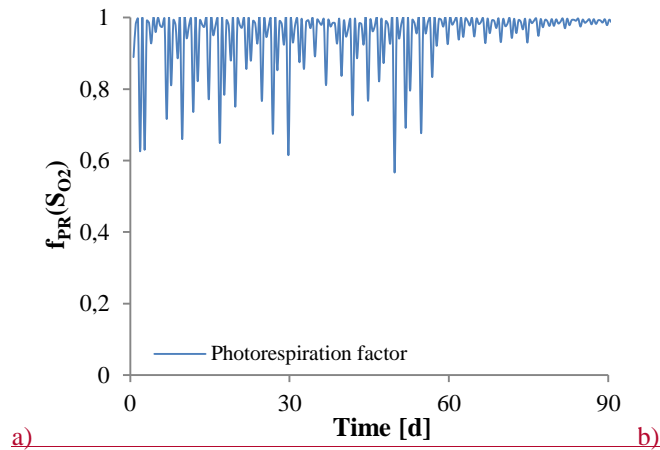


Time [d]

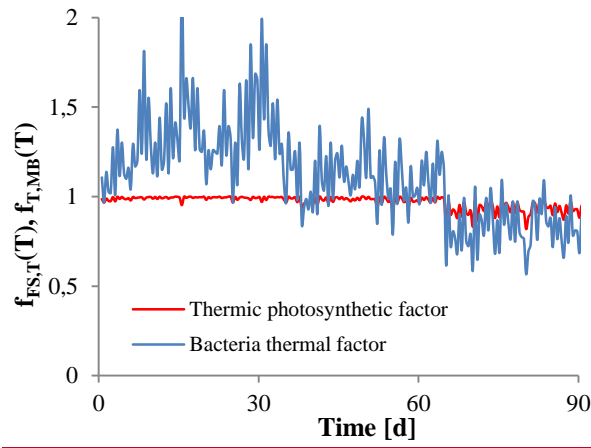
i)

Fig. 3. Changes in experimental (red triangles) and simulated (blue and black lines) a) pH, b) S_{O_2} , c) S_{HCO_3} , d) S_{NH_4} , e) S_{NO_3} f) S_{NO_2} , g) TSS, h) X_{ALG} and X_H , and i) X_{AOB} X_{NOB} concentrations over the Period I (July 21st – October 14th, 1993) in the HRAP_{4d}. Note that in a) and b) values were measured at 9:00 AM \pm 1 hour and 2:00 PM \pm 1 hour. All other values measured at 2:00 PM \pm 1 hour. Higher values of pH and S_{O_2} observed at 2:00 PM \pm 1 hour. S_{NO_3} and S_{NO_2} concentrations were not detected in influent wastewater.

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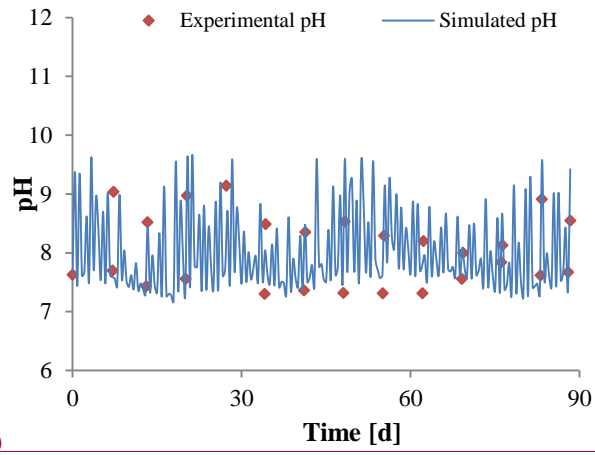


580 **Fig. 4.** a) Changes in the values of $f_{PR}(S_{O_2})$ and b) $f_{FS,T}(T)$ and $f_{T,MB}(T)$ factors over Period I (July 21st – October
581 14th, 1993) in HRAP_{4d}.

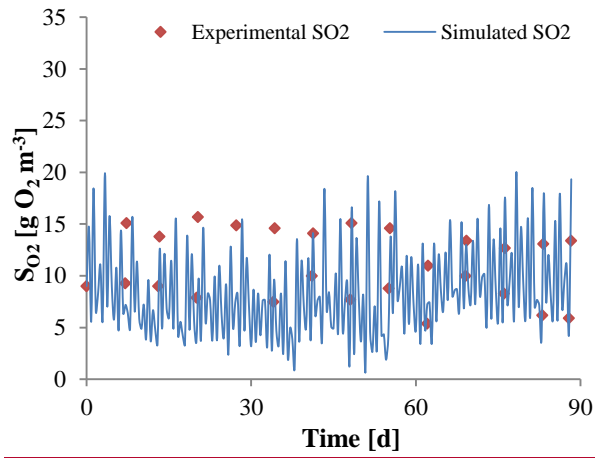
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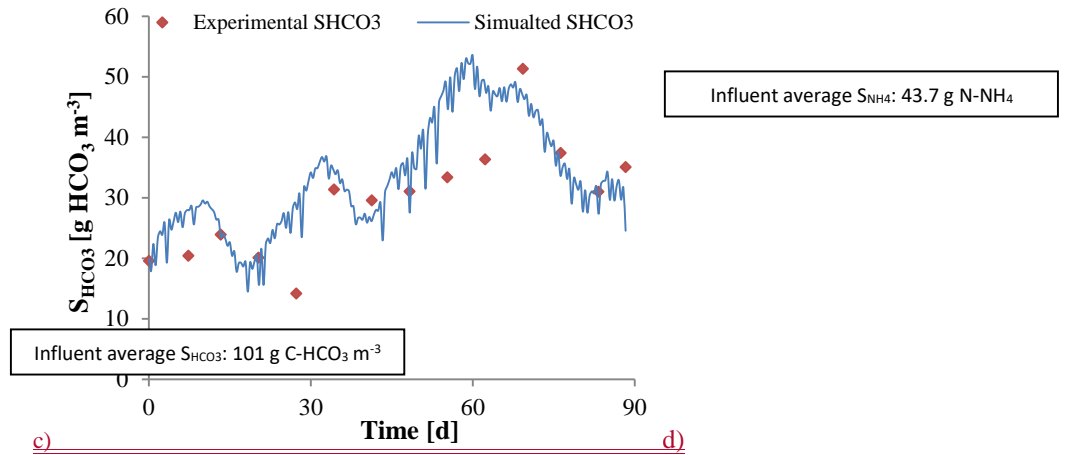
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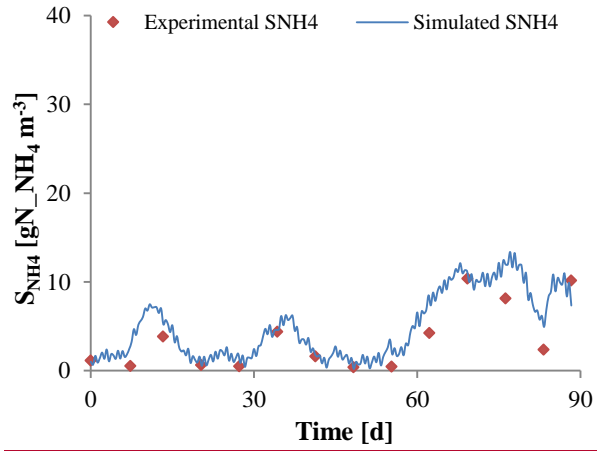
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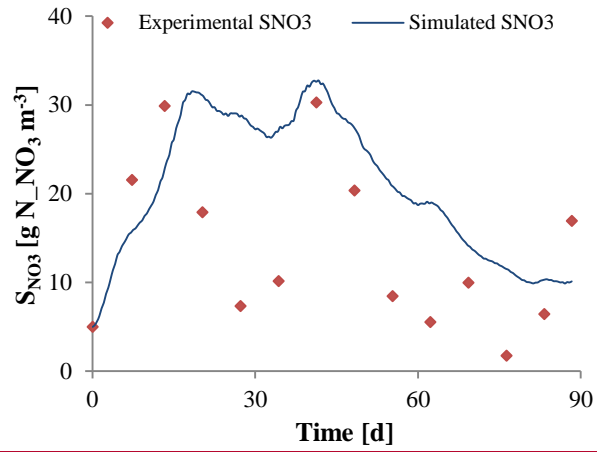
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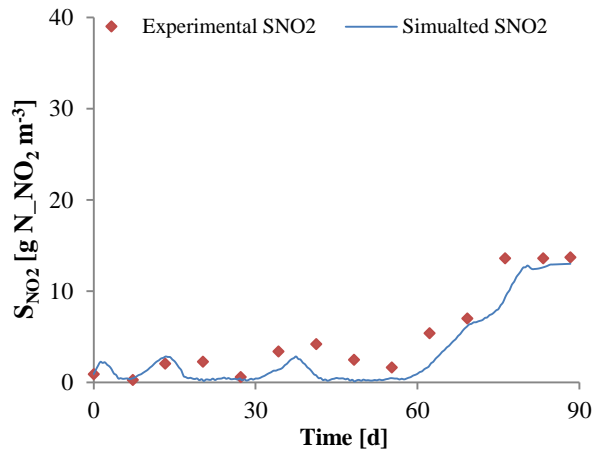
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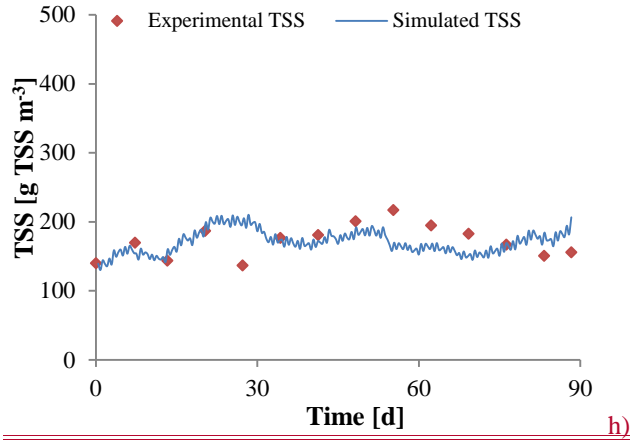
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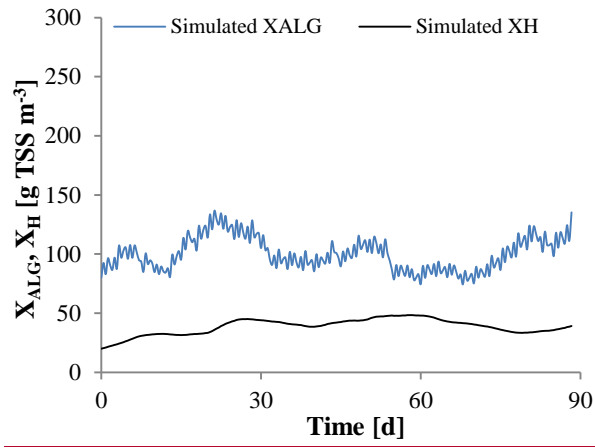


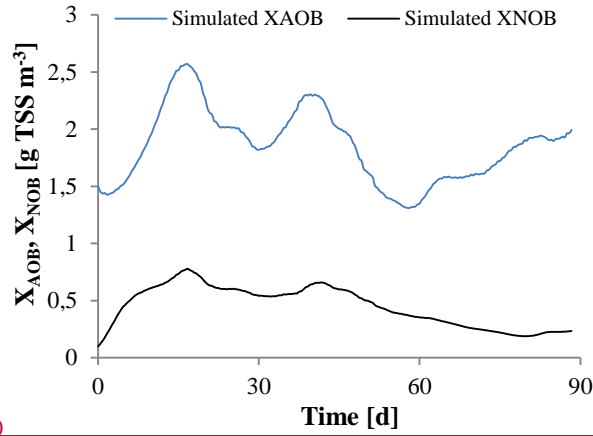
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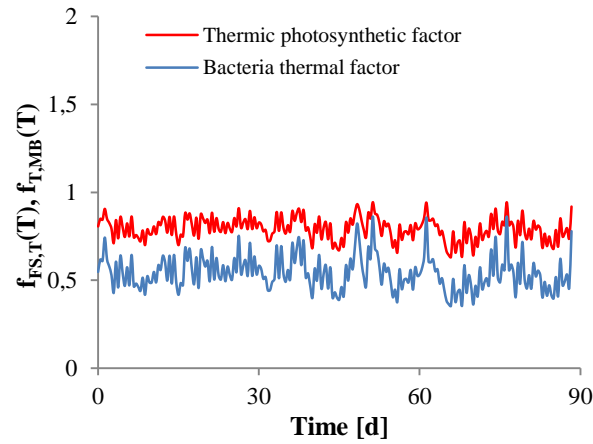


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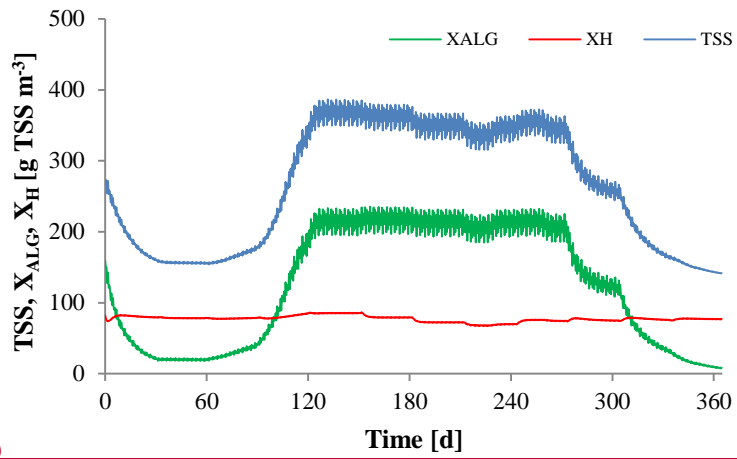


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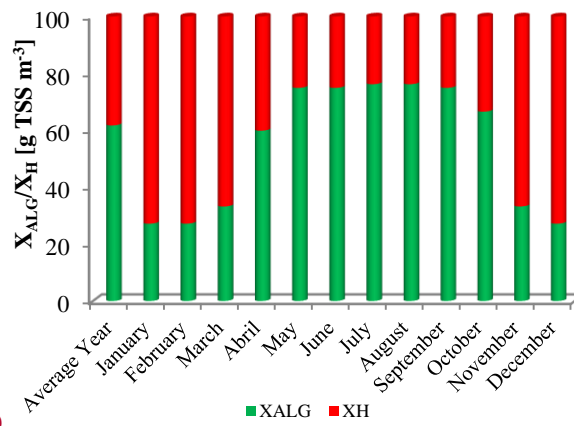
594 **Fig. 5.** Changes in experimental (red diamonds) and simulated (blue and black lines) a) pH, b) SO_2 , c) $SHCO_3$, d) SNH_4
 595 e) SNO_3 , f) SNO_2 , g) TSS, h) X_{ALG} and X_H , i) X_{AOB} and X_{NOB} concentrations and j) changes in the values of $f_{FS,T}(T)$
 596 and $f_{T,MB}(T)$ factor over the Period II (November 10th, 1993 – February 8th, 1994) in the HRAP_{8d}. Note that in a)
 597 and b) values were measured at 9:00 AM \pm 1 hour and 2:00 PM \pm 1. All other values measured at 2:00 PM \pm 1 hour.
 598 Higher values of pH and SO_2 observed at 2:00 PM \pm 1 hour. SNO_3 and SNO_2 concentrations were not detected in influent
 599 wastewater.

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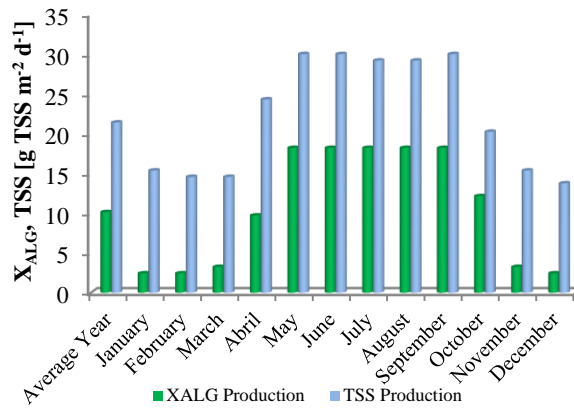
a)



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b)

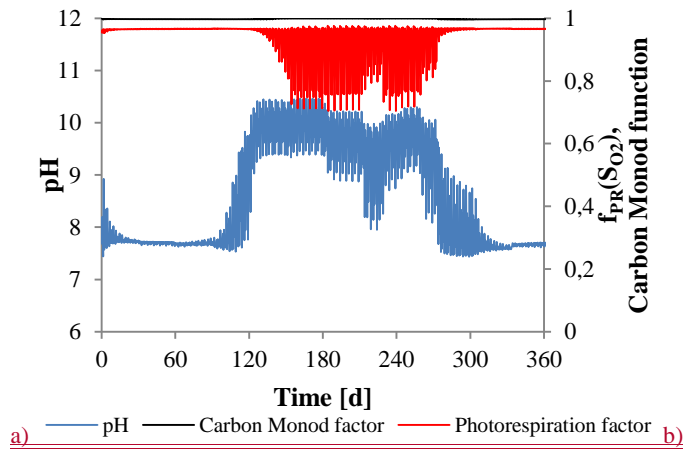
c)



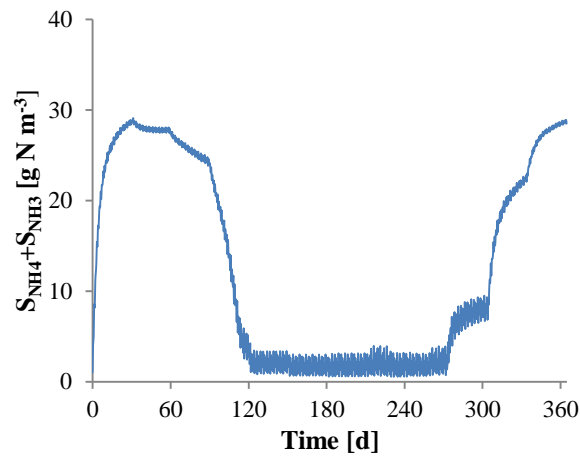
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604 **Fig. 6.** a) Simulated TSS (blue line), X_{ALG} (green line) and X_H (red line) concentration over a year (from January to
 605 December) in HRAP_{4d}. Average annual and monthly b) X_{ALG} and X_H concentration proportion, and c) X_{ALG} and TSS
 606 in HRAP_{4d}.

607



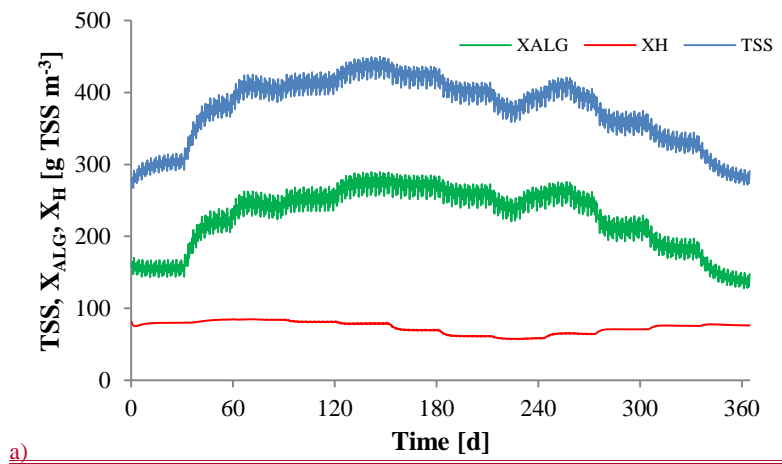
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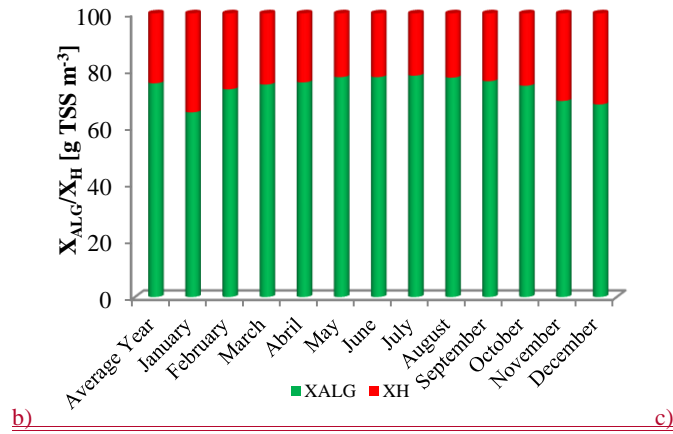
610 **Fig. 7.** Changes in a) pH, $f_{PR}(S_{O_2})$ and carbon Monod function value and b) $S_{NH_4}+S_{NH_3}$ concentration over a year in
 611 HRAP_{4d}.

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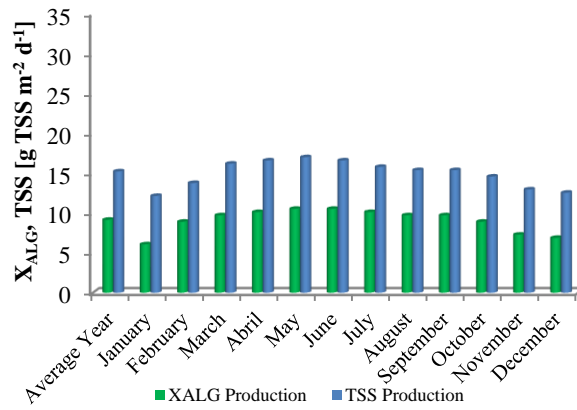


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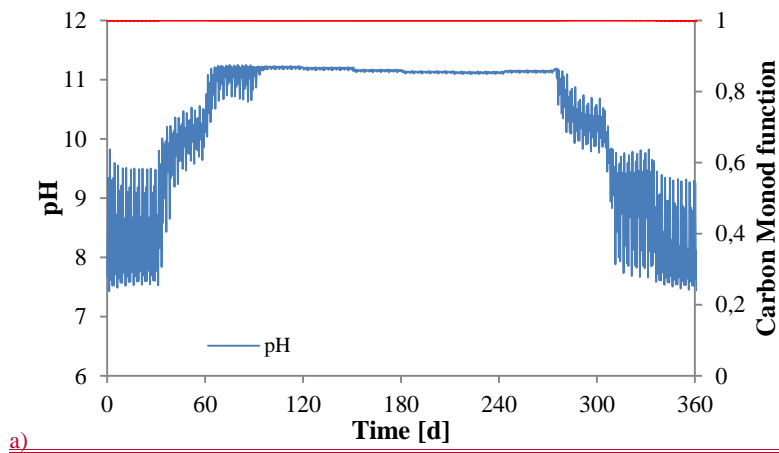


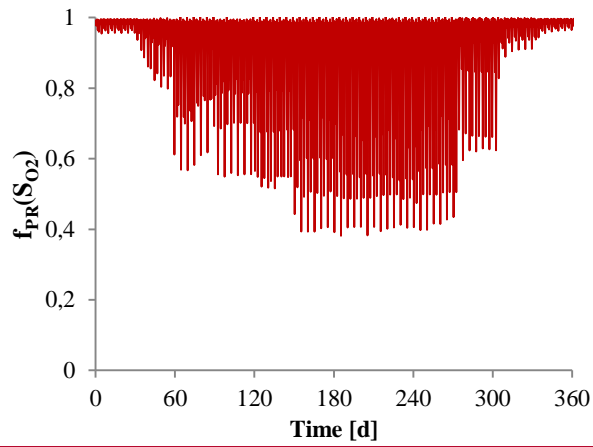
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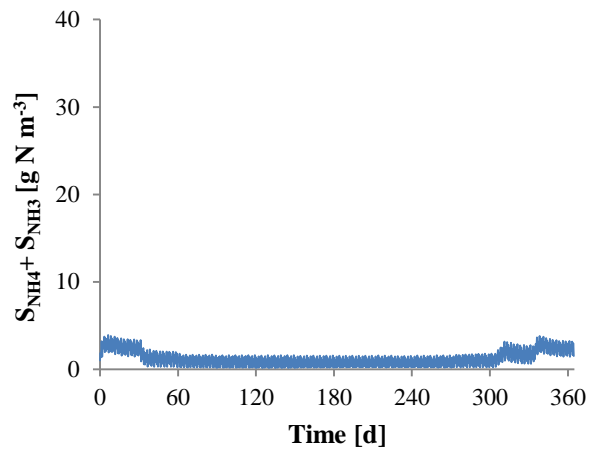
616 **Fig. 8.** a) Simulated TSS (blue line), X_{ALG} (green line) and X_H (red line) concentration over a year (from January to
617 December) in HRAP_{8d}. Average annual and monthly b) X_{ALG} and X_H concentration proportion, and c) X_{ALG} and TSS
618 in HRAP_{8d}.

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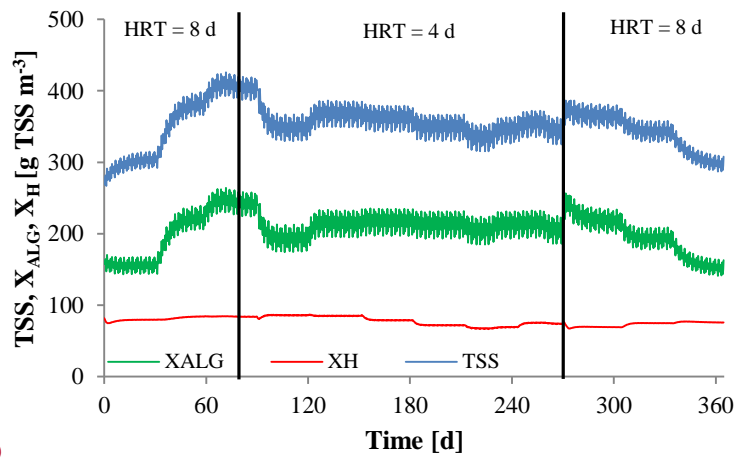
b) c)



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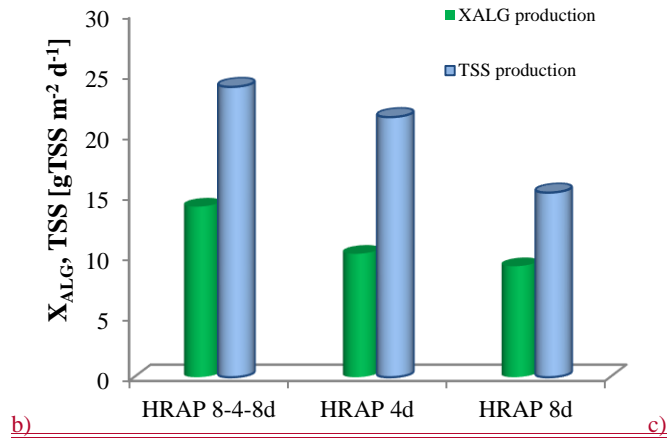
622 **Fig. 9.** Changes in a) pH value and carbon Monod function, b) $f_{PR}(S_{O_2})$ values and c) $S_{NH_4^+}+S_{NH_3}$ concentrations
 623 over a year in HRAP_{8d}.



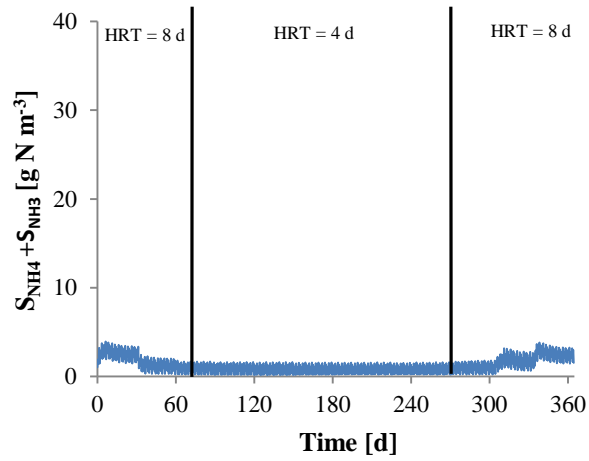
a)

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Fig. 10. a) Simulated TSS (blue line), X_{ALG} (green line) and X_H (red line) concentration over a year (from January to December) in HRAP_{8-4-8d}. Vertical black lines indicate HRT change. b) Comparison of average annual X_{ALG} and TSS production over a year as function of different HRT operating strategies and c) Changes in S_{NH4}+S_{NH3} concentration over a year in HRAP_{8-4-8d}. Vertical black lines indicate the change of HRT.