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Additional Information

1 **Unveiling microbial structures during raw microalgae digestion and co-digestion with**  
2 **primary sludge to produce biogas using semi-continuous AnMBR systems**

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11

12 **Abstract**

13 Methane production from microalgae can be enhanced through anaerobic co-digestion with  
14 carbon-rich substrates and thus mitigate the inhibition risk associated to its low C:N ratio.

15 Acclimated microbial communities for microalgae disruption can be used as a source of  
16 natural enzymes in bioenergy production. However, co-substrates with a certain microbial  
17 diversity such as primary sludge might shift the microbial structure. Substrates were  
18 generated in a Water Resource Recovery Facility (WRRF) and combined as follows:

19 *Scenedesmus* or *Chlorella* digestion and microalgae co-digestion with primary sludge. The  
20 study was performed using two lab-scale Anaerobic Membrane Bioreactors (AnMBR).

21 During three years, different feedstocks scenarios for methane production were evaluated  
22 with special focus on the microbial diversity of the AnMBR. A 57% of the population was  
23 shared between the different feedstock scenarios, revealing the importance of

24 Anaerolineaceae members besides *Smithella* and *Methanosaeta* genera. The addition of  
25 primary sludge enhanced the microbial diversity of the system during both *Chlorella* and

26 *Scenedesmus* co-digestion and promoted different microbial structures. Aceticlastic  
27 methanogen *Methanosaeta* was dominant in all the feedstock scenarios. A more remarkable

28 role of syntrophic fatty acid degraders (*Smithella*, Syntrophobacteraceae) was observed  
29 during co-digestion when only microalgae was digested. However, no significant changes

30 were observed in the microbial composition during anaerobic microalgae digestion when  
31 feeding only *Chlorella* or *Scenedesmus*. This is the first work revealing the composition of  
32 complex communities for semi-continuous bioenergy production from WRRF streams. The  
33 stability and maintenance of a microbial core over-time in semi-continuous AnMBRs is here  
34 shown supporting their future application in full-scale systems for raw microalgae digestion  
35 or co-digestion.

### 36 **Keywords**

37 16S rRNA gene; anaerobic digestion; AnMBR; biogas; codigestion; microalgae

### 38 **1. Introduction**

39 The search of new sources of energy to reduce the CO<sub>2</sub> emissions of fossil fuels and  
40 mitigate this worldwide energy-dependence are among the principal motivations for moving  
41 forward more sustainable technologies and lifestyles. During the last decades, biofuel  
42 implementation has attracted the interest of the scientific community (Correa et al., 2019). As  
43 a forward step, the concept of water resource recovery facilities (WRRF) has emerged for  
44 energy, nutrients, biosolids and reclaimed water recovery from sewage (Colzi Lopes et al.,  
45 2018). Related to this concept, a promising water-energy nexus is the anaerobic treatment of  
46 sewage and the valorization of the resulting effluent for microalgae biomass generation  
47 (González-González et al., 2018). This is a convenient loop, as microalgae can be harvested  
48 and later turned into biogas (González-Fernández et al., 2015) in the previous anaerobic  
49 treatment stage (Xie et al., 2018) or as a side-stream in future WRRF (Seco et al., 2018).

50 The biochemical composition of microalgae makes them suitable for bioenergy  
51 production through anaerobic digestion processes (Klassen et al., 2016). However,  
52 pretreatments used to improve their biodegradability are expensive making the methane  
53 production from microalgae unfeasible (Carrillo-Reyes et al., 2016). Therefore, feasible  
54 bioenergy generation from microalgae in future WRRFs needs biological strategies for  
55 microalgae cell disruption and degradation of the hydrolysed components. Raw conversion of

56 microalgae into biogas is possible when applying high solids retention times (SRT) in  
57 continuous bioreactors under mesophilic and thermophilic conditions (Greses et al., 2018;  
58 Klassen et al., 2016). As early reported by Zamalloa et al. (2012), the Anaerobic Membrane  
59 Bioreactor (AnMBR) allows to increase the biomass retention whilst maintains low hydraulic  
60 retention times (HRT), making possible the continuous anaerobic digestion of microalgae.

61 As early remarked by Rivière et al. (2009), the definition of microbial cores in  
62 engineering systems can provide valuable information during operational parameter  
63 optimization processes. Zamalloa et al. (2012) was the first work relying on microbial groups  
64 of microalgae anaerobic digestion through 16S rRNA gene fingerprinting. More recently,  
65 saccharolytic hydrolyzers and fermenters, as well as proteolytic bacteria from Bacteroidetes  
66 and Firmicutes phyla have been identified during *Chlamydomonas reinhardtii* anaerobic  
67 digestion (Klassen et al., 2017). However, differences in the microalgae species can lead to  
68 different microalgae-degrading communities as their composition varies among their  
69 phylogeny (Baudeflet et al., 2017). Moreover, common microalgae that grow over sewage or  
70 anaerobic effluents have more resistant cell walls and can therefore require higher  
71 microbiological hydrolytic potentials. In this context, acclimation of anaerobic sludge is a  
72 necessary step prior to continuous conversion of microalgae harvested from sewage-related  
73 streams into biogas in WRRFs (Gonzalez-Fernandez et al., 2018). The effect of the type of  
74 microalgae over the acclimated microbial community structures has not been thoroughly  
75 explored yet in the literature as most of the studies are focused on a single microalga.

76 The longer the SRT, the more favorable environment for slow-growing microorganisms  
77 that might be able to disrupt the microalgae cell walls (Greses et al., 2017). However, more  
78 efficient biomethanization of microalgae could be obtained with more balanced C:N ratios  
79 through the addition of a co-substrate with a high carbon content. The protein content of  
80 microalgae has an important drawback as the degradation of these compounds results in the

81 release of nitrogen forms that can accumulate in anaerobic systems as free ammonia.  
82 Methanogens are sensitive to free ammonia and therefore, strategies to mitigate this inhibition  
83 risk are needed to enhance continuous energy production. According to Sialve et al. (2009),  
84 mass ratios between 20 and 35 have a positive effect over methane yield as well as over  
85 microalgae anaerobic digestion and mitigate the inhibition risk.

86 The favorable effect of co-digestion for microalgae anaerobic digestion was recently  
87 reported by Solé-Bundó et al. (2019). The authors achieved a 65% improved biomethanization  
88 when combining primary sludge and *Chlorella* biomass streams from a wastewater treatment  
89 plant. Also, the degradation of *Scenedesmus* with pig manure resulted in a 50% increase of the  
90 methane yield (Astals et al., 2015). Unfortunately, these studies did not evaluated the effect of  
91 SRT over microalgae co-digestion, despite the importance of this parameter to achieve high  
92 microalgae disruption rates (Greses et al., 2018). Also, although several studies have explored  
93 different microalgae co-digestion scenarios (Herrmann et al., 2016; Mahdy et al., 2014; Solé-  
94 Bundó et al., 2018), none of them have been performed in a semi-continuous system operated  
95 under high SRT. Solé-Bundó et al. (2019) reported a 330 mL CH<sub>4</sub>·gVS production from  
96 *Chlorella* and primary sludge in continuous stirred tank reactors (CSTR) but they applied a low  
97 SRT of 20 days and a protease treatment to the microalgae biomass. Furthermore, the  
98 microbiological aspects were not explored in the abovementioned systems and hence, there is  
99 a lack of knowledge on the different groups involved in microalgae co-digestion compared to  
100 a single digestion. Only Li et al. (2017) reported the dominance of Bacteroidetes,  
101 Proteobacteria, Firmicutes and Spirochaetae during co-digestion of *Chlorella* and chicken  
102 manure. However, this study applied a pre-treatment of the microalgae. As reported by  
103 Córdova et al. (2018), microalgae pre-treatment leads on important changes in microbial  
104 patterns, functionality, strategies and interactions during microalgae anaerobic digestion.  
105 According to these authors, delta and gamma Proteobacteria were dominant for untreated

106 *Chlorella* biomass digestion, but Clostridia was the most important group after applying an  
107 alkali-treatment to the same algal biomass. On the other hand, some of the co-substrates that  
108 can be added during microalgae digestion (*e.g.* primary and secondary sludge or manure)  
109 commonly have an inner microbial diversity that can disturb the microbial core developed  
110 during microalgae degradation. These aspects need to be evaluated in continuous systems to  
111 advance towards the design of management tools based on microbial community composition,  
112 like specific biomarker monitoring, in bioenergy production systems.

113         Several combinations of reactor configuration, temperature, SRT, HRT and feedstock  
114 composition that have not been yet evaluated in the literature. In our study, we use microalgae  
115 and primary sludge taken from a WRRF plant (Seco et al., 2018) combining both anaerobic  
116 and microalgae technologies for sewage treatment. Although microalgae digestion has been  
117 thoroughly reported with reliance on the microbial populations (Córdova et al., 2018; Klassen  
118 et al., 2017; Sanz et al., 2017), the microbial core for raw microalgae and primary sludge co-  
119 digestion has not been revealed in the literature. Furthermore, most of the studies including  
120 microbial characterization of systems for biogas production for biogas have been performed  
121 using traditional anaerobic digester configurations. On the contrary, the present work explores  
122 and characterizes the microbial communities of two semi-continuous AnMBRs converting raw  
123 microalgae into biogas. Hence, this study reveals important information about the stability over  
124 time of microbial populations acclimated to microalgae digestion and evaluates the effect over  
125 the microbial core behind this process when adding an extra carbon-source (such as primary  
126 sludge from the same WRRF) to balance the C:N ratio and mitigate the free ammonia inhibition  
127 risk. It should be highlighted that this is the first study reporting information obtained using the  
128 same acclimated biomass to degrade in a semi-continuous process two common microalgae  
129 grown on sewage streams such as *Chlorella* and *Scenedesmus* without any pretreatment.

## 130         **2. Materials and Methods**

### 131 **2.1. Bioreactor operational conditions**

132 Two different lab-scale mesophilic AnMBRs were operated to produce biogas from  
133 microalgae under the operational conditions summarized in Table 1. Both reactors were  
134 operated under mesophilic conditions (35°C). The first AnMBR (digester, Figure S1a) had a  
135 12.4 L volume, 9.9 L working volume, considering the tank and the external hollow-fibre  
136 ultrafiltration membrane tank (0.42 m<sup>2</sup> surface, 0.05 µm pore size, PUR-ON® Koch Membrane  
137 Systems). The second AnMBR (co-digester, Figure S1b) had a 14 L volume (9 L working  
138 volume) and was equipped with an identical external membrane tank to the first AnMBR. A  
139 reservoir tank was coupled to the co-digester AnMBR and used for microbial analysis purposes  
140 as detailed later. The digester was inoculated with mesophilic sludge from a full-scale digester  
141 located in the municipal WTP Carraixet (València, Spain). The co-digester was inoculated  
142 with the stored biomass from the digester, available in the reservoir.

143 The digester was first operated for 20 months at different SRT conditions: 50, 70 and 100  
144 days. During these months, the HRT was set at 50 days (for 50 and 70 days SRT) and later at  
145 15 days (for 70 and 100 days) to increase the OLR of the system from 0.2 to 0.4 gCOD·L<sup>-1</sup>·d<sup>-1</sup>.  
146 <sup>1</sup>. The AnMBR co-digester started running in parallel to the AnMBR digester after 20 months,  
147 fed with the same microalgae feedstock than the AnMBR digester plus the primary sludge. The  
148 SRT of the co-digester was fixed at 100 days SRT, as it was optimized in the previous AnMBR  
149 digester performance. Both AnMBRs were running in parallel for additional 12 months.

### 150 **2.2. Feedstock sources**

151 Microalgae and primary sludge were obtained from a membrane photobioreactor pilot plant  
152 (MPBR) and a primary settler respectively, both located in the municipal WWTP “Cuenca del  
153 Carraixet” (Valencia, Spain). The MPBR pilot plant is used to remove nutrients from the  
154 anaerobic effluent of an AnMBR pilot plant treating sewage (González-Camejo et al., 2019).  
155 The experimental work of this research has lasted almost three years (32 months), in which  
156 *Scenedesmus* and *Chlorella* have separately dominated the MPBR culture. According to

157 microscopic observation and quantification (Pachés et al., 2012), during the first 24 months  
158 more than the 90% of the phytoplankton observed in the MPBR were identified as *Scenedesmus*  
159 spp. Later, a shift in the microalgae population of the MPBR occurred and instead more than  
160 90% of the cells were *Chlorella* spp. This microalga was dominant in the MPBR for the 8  
161 remaining months.

162 A cross-flow ultrafiltration hollow-fiber membrane unit (HF 5.0-43-PM500, PURON®  
163 Koch Membrane Systems) was used for microalgae harvesting and concentration to the  
164 required values prior to feed the AnMBRs to an organic loading rate (OLR) of 0.2-0.4 gVS·L<sup>-1</sup>·d<sup>-1</sup>  
165 (see Table 1). Microalgae feedstock was prepared in a single batch for both systems and  
166 then adjusted to the different concentrations for single- or co-digestion. The primary sludge  
167 was collected from the gravity thickener, sieved through an aperture of 0.5 mm sieve and  
168 diluted to 22.8 gCOD·L<sup>-1</sup> to feed the AnMBR co-digester according to Table 1 OLR conditions  
169 (62%-38% proportion of primary sludge and microalgae based on gVS determination). The  
170 physicochemical characterization of feedstock samples was performed according to APHA  
171 (2012) standard procedures. Feedstock sources were separately stored at 4°C (for no longer  
172 that 3 weeks) to preserve its characteristics and avoid degradation.

## 173 **2.2. Performance analysis**

174 Physicochemical analysis and biogas production were carried out per triplicate and three  
175 times a week as in a previous study (Zamorano-López et al., 2019a). At least the data retrieved  
176 during three pseudo-steady state weeks were considered to calculate the methane yield, the  
177 biodegradability, the solids content of the system (in terms of total suspended solids, TSS) and  
178 the total COD (TCOD). The methane yield was calculated on a COD basis, considering the  
179 COD of the methane produced and measured in the biogas over the total influent COD  
180 associated to each feedstock scenario. The biodegradability of the system was thus calculated



181 on this basis using the theoretical potential of  $350 \text{ mLCH}_4 \cdot \text{gCOD}_{\text{inf}}^{-1}$  (TMP  $0^\circ\text{C}$ , 1 atm) and  
182 expressed as the percentage of the biomethanization achieved for each feedstock scenario.

### 183 **2.3. Sample collection for microbial ecology analysis**

184 Digestate samples were extracted from each AnMBR during the different pseudo-steady  
185 state periods achieved for the different combination of operational parameters applied to each  
186 AnMBR (Table 1). Since pseudo steady state was reached before each biomass collection point,  
187 samples can be considered biological replicates for each microalgae mono- and co-digestion  
188 scenario evaluated. Under each period, stabilized measures of digestate COD and TSS in the  
189 digestate, as well as the methane yield were determined in each AnMBR (Table 2).

190 All samples were frozen at  $-20^\circ\text{C}$  prior to the nucleic acid extraction. At least two samples  
191 were collected for each AnMBR experimental period regardless of the inoculum. In total, 13  
192 samples were collected from the digester, whereas 9 samples were extracted from the co-  
193 digester. Co-digester samples were duplicated as the AnMBR co-digester set-up (Figure S1b)  
194 included a reservoir tank where the digestate extracted to maintain the SRT in the main tank was  
195 stored also at  $35^\circ\text{C}$ . Additionally, 9 samples were extracted from the reservoir at the same  
196 collection points than the co-digester. Two extra samples were also stored from the reservoir  
197 at days 124 and 170. Hence, 33 samples were used in total in this study for microbial analysis.

### 198 **2.4. Nucleic acid extraction, 16S rRNA gene library preparation and amplicon** 199 **sequencing**

200 Following the procedures from Zamorano-López et al. (2019) the nucleic acids were  
201 extracted from each sample and frozen at  $-20^\circ\text{C}$  prior to their submission to the sequencing  
202 service of the *Fundación para el Fomento de la Investigación Sanitaria y Biomédica de la*  
203 *Comunitat Valenciana* (FISABIO, Valencia, Spain). Primers targeting the v3 to v4 region of  
204 the 16S rRNA gene were used for library preparation. The sequencing run was performed in a  
205 2x300 bp paired-end run using an Illumina Miseq sequencer and v3 reagent kit. The raw results  
206 can be found in the Sequence Reads Archive (SRA) repository from the NCBI platform:

207 bioproject PRJNA434206, accession numbers SAMN11567542-50 (co-digester),  
208 SAMN11567551-63 (digester) and SAMN11567566-76 (reservoir).

### 209 **2.5. Diversity analysis**

210 The sequences retrieved from the Illumina amplicon sequencing approach were analyzed  
211 as in previous studies (see Zamorano-López et al., 2019). Different Operational Taxonomic  
212 Units (OTU<sub>0.97</sub>) were defined at a 3% dissimilarity in an open-reference cluster step using  
213 QIIME. The weighted unifracs distance was estimated in all samples to explore the beta-  
214 diversity. The richness estimators chao1 and PD whole tree, jointly with the number of OTU<sub>0.97</sub>  
215 observed and the simpson evenness (simpson\_e) index were used to analyze the alpha-diversity  
216 of the bioreactor extracted samples. Biom resulting table from QIIME containing the OTU<sub>0.97</sub>  
217 composition and taxonomic assignments according to SILVA v128 release was exported to  
218 further analyze the microbial community.

### 219 **2.6. Biostatistics**

220 All biostatistics analysis were performed using R-studio (v.3.2) within vegan and  
221 mixomics packages. A principal co-ordinate analysis (PCoA) based on the weighted unifracs  
222 distances matrix was used to evaluate the beta-diversity of the different samples collected from  
223 both AnMBRs. Adonis test over the PCoA results were performed using 999 permutations for  
224 feedstock and digester categorical variable clusters. A Partial Least Square Discriminant  
225 Analysis (PLS-DA) was performed over all samples (digester, co-digester and reservoir) to  
226 explore the effect of the primary sludge addition over the AnMBRs populations. This statistical  
227 analysis allows to extract the most discriminant OTU<sub>0.97</sub> among a group of samples and their  
228 major association to any of the two AnMBR systems studied here.

## 229 **3. Results and discussion**

### 230 **3.1. 16S rRNA sequencing data analysis and alpha-diversity measurements**

231 The 16S rRNA gene amplicon sequencing approach resulted in a total of 1,431,467 raw  
232 sequences that after downstream analysis with high-quality settings resulted in an average of

233 57,409 clean sequences per sample. After rarefaction to the minimum value of clean sequences  
234 observed in the dataset (27,647) different alpha diversity estimators were extracted (Table 3).  
235 To compare these values, only samples taken under the same SRT in each AnMBR scenario  
236 were considered, since this parameter can strongly enhance species richness and diversity in  
237 anaerobic systems with high solids retention capacity such as the AnMBR.

238 The highest diversity was found in the samples taken during *Chlorella* digestion: 4150  
239 observed OTUs. This scenario also presented the highest diversity in terms of non-detected  
240 OTUs, which are estimated through the chao1 index (7075). On the contrary, the *Scenedesmus*  
241 scenario presumably had the minimum diversity observed with 3358 OTUs and an estimated  
242 6023 chao1 index value. This could be related to the development of a more specific  
243 community for *Scenedesmus* digestion than for *Chlorella digestion*. As it has been reported in  
244 the literature, *Scenedesmus* is among the hardest Chlorophyta member for direct disruption  
245 using microbial communities due to the presence of algaenan (Fernandez et al., 2018).  
246 Although *Chlorella* cell walls are also composed of recalcitrant compounds similar to chitin  
247 (Baudeflet et al., 2017), the n-alkaenan composition of algaenan could have a stronger selective  
248 pressure effect over microbial communities and therefore decrease AnMBR alpha diversity.

249 Phylogenetic similarity of each sample can be measured through the PD\_whole\_tree  
250 estimator (Table 3). The higher number of phylogenetic tree branches, the higher value of  
251 PD\_whole\_tree estimator and thus, this value reveals the existence of more diverse and distant  
252 species in each sample. The highest PD\_whole\_tree values were observed during *Chlorella*  
253 digestion, again suggesting that this was the more diverse feedstock scenario of the four  
254 studied. Between the two co-digestion scenarios, slight differences were observed in the three  
255 indexes (observed\_otus, chao1 and PD\_whole\_tree). This could be related to the higher  
256 presence of microbial groups with wider metabolic capacities in the digester when both  
257 substrates were present than when it was only fed with microalgae.

258 The evenness measurement retrieved for each scenario (see simpson evenness index in  
259 Table 3) reflected that the changes in the relative abundance patterns of the observed OTUs  
260 were more dynamic in the co-digestion scenarios than when only microalgae was digested. It  
261 should be noticed that from an ecological perspective, the addition of a co-substrate which has  
262 a certain microbial diversity can enhance richness and evenness diversity due to the presence  
263 of minor and rare groups that might not be active in the anaerobic system but are though  
264 retained. Related to this, Chen et al. (2019) observed higher diversity in the primary sludge  
265 than in the anaerobic digester samples. Thus, primary sludge could also have enhanced  
266 evenness in the AnMBR co-digester in this work. Interestingly, Greses et al. (2017) pointed  
267 out that despite a shared bacterial diversity of 32% between microalgae feedstock and  
268 anaerobic digester samples, the resulting communities established in the microalgae digester  
269 were significantly different from the influent. Consequently, the influence of diversity-rich  
270 feedstock, especially in presence of anaerobic microorganisms (like it occurs in the primary  
271 sludge), over anaerobic digestion communities should be carefully explored in bioreactor  
272 configurations such as the AnMBR. In this system, the use of ultrafiltration membranes  
273 enhances the retention of niche and biofouling-related microorganisms (Robles et al., 2018;  
274 Skouteris et al., 2012). Furthermore, high solids retention capacity enhances microbial  
275 persistence resulting in microbial communities with high diversity and richness, according to  
276 16S rRNA/rRNA gene sequencing results (Mansfeldt et al., 2019).

### 277 ***3.2. Beta diversity analysis reveals different structures of microalgae-degrading*** 278 ***communities in the AnMBRs***

279 According to the beta-diversity analysis performed through PCoA over the weighted  
280 unfrac distance matrix, there are different structures among samples depending on the  
281 microalgae biomass used as feedstock and the addition or not of a co-substrate (*e.g.* primary  
282 sludge). The first component of the PCoA explains the 38% of the differences between the  
283 samples that were collected from the digester when the primary sludge was added or not added.

284 The second component explains 27% of the variability between samples and specially remarks  
285 a change in the structure of co-digester samples (Figure 1).

286 As shown in Figure 1, samples were categorized according to the digester and the  
287 feedstock. For the first categorical variable, two clusters were revealed by Adonis test (digester,  
288  $p < 0.001$ ). Hence, the microbial structure of the co-digester and its reservoir was consistent in  
289 between but differed from the microbial structure of the digester samples. The second  
290 categorical variable used in the Adonis test revealed the existence of three clusters (feedstock,  
291  $p < 0.011$ ), although four feedstock scenarios were analyzed in the present study. Thus, the  
292 differences in the microbial community structures of both AnMBRs should be attributed to the  
293 addition or not of a co-substrate and not to the species of microalgae fed to the reactor. In fact,  
294 microbial structure in the digester did not shift significantly when feeding *Scenedesmus* or  
295 *Chlorella*. The change in the microalgae did not either disturb the microbial structure of the  
296 co-digester, since the co-digester early stages samples are grouped with the *Scenedesmus* and  
297 primary sludge scenario samples (see top left corner samples in Figure 1). Finally, the  
298 differences among the digester samples were related to the effect of the SRT over the microbial  
299 population and the acclimation trend of the biomass, as previously mentioned.

300 The proximity between the samples collected when digesting *Scenedesmus* or *Chlorella*  
301 observed through the PCoA (Figure 1) suggest the potential use of the same anaerobic biomass  
302 to degrade these two algae. This is a remarkable fact and highlights the potential use of this  
303 acclimated biomass in microalgae-based bioenergy recovery processes. This concept which is  
304 based on a circular economy requires low-cost stages of microalgae disruption. An attractive  
305 strategy is to use these acclimated microbial communities as hydrolytic biomass sources and  
306 convert microalgae into biomethane through anaerobic digestion. Both Chlorophyta belonging  
307 genera are commonly found in fresh water and spontaneously grow over sewage-treated  
308 effluents (Garrido-Cardenas et al., 2018). Hence, the findings here reported support the use of

309 this biological strategy in a loop-system combining microalgae cultivation using anaerobically  
310 treated sewage-effluents, biomass harvesting and their further conversion into energy.

### 311 ***3.3. Combining feedstock acclimation and high SRT operation to promote microalgae*** 312 ***degrading microorganisms***

313  
314 During SRT acclimation from 50 days up to 100 days in the digester AnMBR, slow-  
315 growing hydrolytic microorganisms were selected allowing the degradation of raw microalgae  
316 with remarkable methane yields (Table 2). Figure 2 shows the relative abundances calculated  
317 at phylum level from the OTU<sub>0.97</sub> compositions among samples of the AnMBR digester. The  
318 changes in the patterns reveals the effect of SRT over microbial composition.

319 During the operation at the lowest SRT (50 days) the dominant phyla observed were:  
320 23.5% Chloroflexi, 16.6% Proteobacteria, 11.1% Planctomycetes and 9.7% Firmicutes (Table  
321 S4, samples xx). These four groups were present during the whole experimental period and are  
322 common groups of anaerobic digesters, as shown in different studies of full-scale systems  
323 (Calusinska et al., 2018; De Vrieze et al., 2018) and also in microalgae digesters (Córdova et  
324 al., 2018; Greses et al., 2018, 2017; Klassen et al., 2017; Sanz et al., 2017) or co-digesters (Li  
325 et al., 2017). However, their relative abundances changed under different SRT operation as  
326 other microbial groups like Bacteroidetes, Cloacimonetes, Spirochaetes, Aminicenantes and  
327 Candidatus Dojkabacteria (WS6 phylum) thrived in the system and co-existed with the  
328 previous phyla. The operation at 70 days SRT with an HRT of 50 days was characterized by  
329 the remarkable presence of the Ca. Dojkabacteria (14.8%, see Table S4). This novel group is  
330 poorly described and none of the belonging members has been isolated yet. Their early  
331 identification by Dojka et al. (2000) using culture-independent approaches (16S rRNA gene  
332 cloning) suggested their importance in organic-rich environmental anaerobic niches. Up to the  
333 present date and to the knowledge of the authors of this manuscript, no other studies have  
334 clearly described their function in anaerobic digesters for microalgae conversion into energy.  
335 Interestingly, Qiao et al. (2013) observed Ca. Dojkabacteria during anaerobic digestion of corn

336 straw. According to more recent metagenomic findings *Ca. Dojkabacteria* related OTUs have  
337 xylan disruption capacity (Solden et al., 2016). This sugar is commonly observed in  
338 Chlorophyta cell walls (Baudeflet et al., 2017; Domozych, 2014) and hence, the role of this  
339 phylum in microalgae degradation could be suggested from these findings.

340 However, *Ca. Dojkabacteria* presence decreased in the AnMBR after changing the HRT  
341 from 50 to 15 days. An antagonist response was observed for Firmicutes phylum, which was  
342 favoured during increased SRT operation at 70 days, reaching relative abundance values up to  
343 12.5% (sample 211, Figure 2). Different members of Firmicutes are commonly reported in  
344 complex polysaccharide anaerobic degradation, since they can release enzymes to the  
345 environment and disrupt complex molecules (Calusinska et al., 2018; Cheng et al., 2014). The  
346 Firmicutes phylum decreased in terms of relative abundance after the HRT reduction from 50  
347 to 15 days, suggesting that other microbial groups have higher affinity for the substrates and  
348 thrived in the AnMBR. Despite the maintenance of a SRT of 70 days, the reduction of the HRT  
349 increases the organic loading rate of the system and reduces the contact time in between the  
350 soluble phase and the microorganisms. Thus, lower HRT can affect the mass transference of  
351 the system and enhance microbial groups with lower specific rates of substrate utilization.

352 During operation at high 15 days HRT and high SRT the relative abundances of  
353 Bacteroidetes and Aminicenantes phyla increased at 100 days SRT. Both groups remained in  
354 the AnMBR digester during operation at 70 days SRT, although their relative abundance values  
355 were lower over time and especially at the end of *Scenedesmus* digestion (samples 483 and  
356 624, Figure 2). Then, for *Chlorella* digestion also at high SRT of 100 days and low HRT of 15  
357 days, changes in the phyla profiles were observed. Consequently, the relative abundances of  
358 both Bacteroidetes and Aminicenantes were lower for *Chlorella* digestion scenario than for  
359 *Scenedesmus*. Both phyla have been related to the core of wastewater anaerobic digestion  
360 systems in a recent study performed over twenty years targeting the 16S rRNA gene

361 (Calusinska et al., 2018). The role of Bacteroidetes in the present work could be more  
362 heterogeneous, as different members related to this phylum are involved in both polysaccharide  
363 and peptide degradation. Indeed, Bacteroidetes has been remarked as a key phylum continuous  
364 raw microalgae digestion for methane production (Klassen et al., 2017). On the other hand,  
365 Farag et al. (2014) early suggested the wide potential metabolic implication of Aminicenantes  
366 in anaerobic environments. However, little is known about this group as none of the  
367 representative members of this has been isolated yet in a pure culture, but recent findings  
368 suggest their importance in hydrogen and acetate production after saccharolytic degradation  
369 (Kadnikov et al., 2019). Hence, they could play an important role during microalgae  
370 degradation at high SRT as methanogenic substrate donors.

371 According to these results, a robust long-time acclimation of the mesophilic inoculum used  
372 in the digester resulted in an enrichment of potential microalgae degraders from the  
373 Chloroflexi, Proteobacteria, Bacteroidetes and Aminicenantes phyla that were retained in the  
374 system through membrane operation even under different SRT conditions. Hence, this  
375 acclimated community could be inoculated in another anaerobic system coupled to future  
376 WRRFs to produce bioenergy from sewage in an anaerobic-microalgae loop technology.

#### 377 ***3.4. A microbial core with similar biomethanization pathways from microalgae and*** 378 ***primary sludge***

379 In the present work another AnMBR was run in parallel using the same microalgae biomass  
380 plus primary sludge collected from the WRRF primary settler. Similar communities might be  
381 established when treating the same feedstock sources, as a result of the stabilization of a  
382 microbial core in biogas reactors (Zuopeng et al., 2019). In fact, a microbial core for microalgae  
383 biomethanization was elucidated in this study, as the Venn diagram shows (Figure 3). A total  
384 number of 578 OTU<sub>0.97</sub> were shared between the AnMBR operated under the different  
385 scenarios.



386 Also, the venn diagram revealed the presence of unique OTU<sub>0.97</sub> in the four scenarios. The  
387 scenario with the highest number of unique members was Chlorella and Primary Sludge (131  
388 OTU<sub>0.97</sub>). The remaining scenarios had 103, 109 and 92 specific OTU<sub>0.97</sub> (Scenedesmus,  
389 Scenedesmus and Primary Sludge and Chlorella, respectively) (Figure 3). The small difference  
390 between the digestion and co-digestion scenario for Scenedesmus contrasts with the high  
391 difference in terms of unique OTU<sub>0.97</sub> of Chlorella digestion and co-digestion, which showed  
392 the lowest and highest value of unique members. These findings are similar to the alpha  
393 diversity analysis results, since Scenedesmus scenarios had higher specificity than Chlorella  
394 scenarios. However, the unique OTU<sub>0.97</sub> were not present in relative abundance over 0.7% in  
395 any sample. Hence, the presence of specific members in each different scenario might not be  
396 as important as the persistence of a microbial core of 578 OTU<sub>0.97</sub> that are shared in between  
397 the four scenarios.

398 The most abundant OTU<sub>0.97</sub> found in the microbial core were related to uncultured members  
399 of Anaerolineaceae family (phylum Chloroflexi), Synergistaceae (Synergistetes) and the  
400 candidate phylum Cloacamonas; besides Smithella and Methanosaeta (Figure 4). *Smithella*  
401 genus (order Syntrophobacterales) was predominantly observed in the co-digester, coinciding  
402 its highest values within the highest presence detected of *Methanosaeta* (order  
403 Methanosarcinales). As reported by Leng et al. (2018), both genera are commonly found in  
404 anaerobic digestion processes and play an important role during methane production after fatty  
405 acid conversion into a more reduced form *i.e.* acetate. Although no other omics approach rather  
406 than DNA amplicon sequencing was performed in this work, the findings of both microbial  
407 members suggest that methane was mainly produced by *Methanosaeta* through the *Smithella*  
408 pathway. Indeed, the comparison of the consensus sequence for each OTU<sub>0.97</sub> reported a 99%  
409 and 96% identity with *Methanosaeta concilii* and *Smithella propionica*. This would have been  
410 promoted through the addition of the co-substrate to balance the C:N ratio of the influent. This

411 strategy can also have a positive impact on enzymatic synthesis stimulation improving  
412 microalgae anaerobic digestion yields (Sialve et al., 2009) and consequently, methanogenic  
413 population. In contrast, the microalgae digester presented a lower abundance of *Methanosaeta*  
414 (1.3% when digesting *Scenedesmus* and 0.6% with *Chlorella*) and *Smithella* was detected at  
415 very low values (4.8% and 0.6%, respectively). This could be related to the less balanced  
416 scenario for methane production as the C:N ratio in the digester was lower than in the co-  
417 digester. In fact, as shown in Figure 4 higher presence of syntrophic members classified under  
418 the Syntrophobacterales family were observed for the two co-digestion scenarios.

419 The still poorly characterized phylum Chloroflexi showed up a very high presence in all  
420 samples. Summarizing, three OTU<sub>0.97</sub> were observed in the digester when the co-substrate was  
421 added as well as when only microalgae was digested. However, more relative abundance of  
422 Anaerolineaceae clusters I and II were observed in the co-digestion scenarios, compared to the  
423 digestion scenarios (Figure 4). According to the review from McIlroy et al. (2017) all isolated  
424 members of this family are donors of acetate after fermentation of carbohydrates. Also, this  
425 family has been proposed as biological disrupters of microalgae (Greses et al., 2017; Sanz et  
426 al., 2017) besides macroalgae (Zou et al., 2018) and would be involved in the production of  
427 other fermentation products such as lactate, hydrogen and formate. Interestingly, the cells of  
428 the microorganisms belonging to this group are filamentous type. A recent study from Bovio  
429 et al. (2019) supports their importance in granule generation in Up-flow Anaerobic Sludge  
430 Blanket (UASB) systems. This is a key capacity also during biofouling and cake formation  
431 processes in AnMBR systems that could explain the dominance of Anaerolineaceae in the  
432 present study. Moreover, as reported by McIlroy et al. (2017), Anaerolineaceae and  
433 *Methanosaeta* are commonly associated forming a complex filamentous network. If this group  
434 was major donors of acetate to *Methanosaeta* in this work, the association in a “spaghetti-like”

435 structure of Anaerolineaceae and the aceticlastic methanogen could have promoted the  
436 metabolites transfer flux between both groups, resulting in high methane production rates.

437 In summary, the ecology of the microbial core forming members suggest a relevant role  
438 of propionate production and further reduction during the digestion of microalgae with and  
439 without co-substrate. The higher detection of potential syntrophs during codigestion might be  
440 related to the favorable effect of the addition of an extra carbon source to the AnMBR. Since  
441 microalgae composition is less heterogeneous than primary sludge, metabolic pathways might  
442 tend to be more specific with higher reliance on fermentation of sugars into acetate or on amino  
443 acid fermentation after protein lysis. Besides, the synergies promoted by the addition of  
444 primary sludge would be reflected in the diversity of intermediate steps before methane  
445 production such as the propionate-dependending *Smithella* pathway.

446 This microbial core has been defined in terms of dominant relative abundances (see Tables  
447 S1-S4 for further details). However, further research is needed in order to develop future  
448 methodologies for monitoring the dynamics of these groups in anaerobic systems producing  
449 energy from microalgae. Since microalgae degradation is complex from a metabolic  
450 perspective due to the heterogeneous composition of microalgae cell walls (Baudelet et al.,  
451 2017), targeting the members of the microalgae-degrading microbial core could be an effective  
452 strategy to monitor microalgae digesters. A future necessary step would be the design of  
453 specific probes or oligonucleotides that can target the rRNA and provide the activity levels of  
454 these groups. Besides, qPCR approaches or 16S rRNA gene sequencing coupled to flow-  
455 cytometry sorting systems (Rinke, 2018; Wang et al., 2010) or including a spike control  
456 (Stämmeler et al., 2016) could provide absolute measurements of these relevant microorganisms  
457 for bioenergy production. Towards the development of future microbial-based models of  
458 anaerobic digestion of complex feedstocks that are produced in WRRF this effort should be  
459 considered, since microbial communities cannot be longer overstated (Widder et al., 2016).

460 **3.5.PLS-DA analysis to find differences between microalgae digestion and co-digestion**  
461 **with primary sludge from relative abundance magnitude**  
462

463 All OTU<sub>0.97</sub> including minor and rare groups were considered for PLS-DA model  
464 construction. As detailed in the methods section, those groups at very low relative abundances  
465 are removed during downstream sequencing analysis. However, there are several groups that  
466 are in relative abundances values below 1% but might play an important functional and  
467 ecological role in complex microbial networks (Rivière et al., 2009). Interestingly, these groups  
468 might be the most discriminants of each microbial structure observed during microalgae  
469 digestion and co-digestion with primary sludge due to their presence or absence.

470 Figure 5 shows the results from the fitted PLS-DA regression model. As can be seen in this  
471 figure the samples from the microalgae digestion are well separated from the samples from co-  
472 digestion. To elucidate the most discriminant groups between both digestion substrates, the  
473 variable importance in the projection (VIP) was calculated. The first 30 microbial members  
474 sorted by the highest VIP value retrieved from PLS-DA are shown in Figure S2. Genera  
475 belonging to Actinobacteria, Atribacteria, Chloroflexi, Cloacimonetes, Firmicutes,  
476 Proteobacteria, Spirochaetae, Verrucomicrobia (Bacteria) and WSA2 (Archaea) were among  
477 the most discriminant ones. Some of them are classified inside of the dominant phyla observed  
478 in both AnMBRs (Figure 2). However, others like *Ca. Caldatribacterium* (phylum  
479 Atribacteria) are detected at very low abundances (1-2%) but were highly discriminating  
480 between samples. According to Dodsworth et al. (2014) this group is able to perform  
481 saccharolytic fermentation from cellulosic as well as hemi-cellulosic substrates. Since cellulose  
482 is present in common WTP primary sludge stream in about 30-50% of the influent suspended  
483 solids (Crutchik et al., 2018), the thrive of this bacteria group during co-digestion but not when  
484 only microalgae was digested could be related to the higher presence of this complex  
485 polysaccharide in the feedstock. *Treponema*, a Spirochaetae member, was also among the most  
486 discriminant and found only in the samples from co-digestion. The presence of this group was

487 associated in a co-digestion study of sewage sludge and food waste (Cheng et al., 2014).  
488 Besides, the saccharolytic capacity of *Treponema* might explain their presence in this work and  
489 other microalgae degrading bioreactors (Klassen et al., 2016; Sanz et al., 2017). Nevertheless,  
490 future analysis with complementary approaches to amplicon sequencing such as proteomics  
491 would be needed in order to understand the complete metabolic implication of these groups  
492 and elucidate their link to primary sludge digestion or to microalgae degradation.

493 **3.6. Ecological implications of complex and diversity richness during raw feedstock**  
494 ***anaerobic digestion and future research needs***

495 The use of microbial-rich biomass sources as co-substrate might present a drawback when  
496 using biological strategies to convert microalgae into biogas. The primary sludge strongly  
497 shaped the microbial communities in the co-digester as shown in the PCoA (Figure 1). From a  
498 microbial ecology perspective, this could also be partially related to the accumulation of co-  
499 substrate incoming microorganisms and groups entering the system might be viable during  
500 microalgae co-digestion. Primary sludge has a high species richness. Although its diversity has  
501 not been evaluated on its own in the present study and is rarely evaluated in similar studies, Ju  
502 et al. (2017) observed 3424 OTU<sub>0.97</sub> in the primary sludge seed used for their anaerobic  
503 digestion trials. However, this study only relied on the microbial characterization through the  
504 biomarker 16S rRNA gene and could not therefore evaluate the survival of these potential  
505 microbial groups present in the influent. Further research using transcriptomic approaches  
506 might help to elucidate the activity levels of the microorganisms observed. Since some of  
507 microorganisms are anaerobic and might be acclimated to cellulolytic components present in  
508 the primary sludge, they could improve the later digestion of microalgae during the co-  
509 digestion.

510 The present work has demonstrated that a core representing the 57% of the microbial  
511 diversity is maintained over time in bioreactors treating microalgae. The maintenance of a core  
512 microbiome in anaerobic reactors was reported to be extremely relevant in order to maintain

513 the functional status (Rivière et al., 2009). Peces et al. (2018) reported a convergent diversity  
514 after 120 days of continuous operation of four different anaerobic digesters, inoculated with  
515 different sources but identically operated to produce biogas from a cellulose:casein feedstock.  
516 According to these authors, the microbial core contained a 78% of the anaerobic digesters  
517 diversity. The neutral theory predicts that populations are driven by deterministic factors such  
518 as SRT, HRT and OLR, as it has been demonstrated using different inocula to anaerobically  
519 degrade cellulose (Vanwonterghem et al., 2014). Up to date, most of the microbial core focused  
520 studies have only used the target 16S rRNA gene. Therefore, further research is needed in order  
521 to elucidate the active microbial core, as minor groups might have a relevant role during  
522 microalgae digestion. This has been suggested in the present study through application of PLS-  
523 DA that remarks the importance of the presence or absence of certain groups to shape microbial  
524 structures, despite of their low relative abundances. On this basis, RNA-based sequencing (De  
525 Vrieze et al., 2018) could facilitate a better profile of key microorganisms during microalgae  
526 digestion and especially during co-digestion. Functional profiling of anaerobic communities is  
527 a necessary step towards the development of new probes to monitor the wealth of anaerobic  
528 digesters from a microbiologist perspective. Also, to retrieve more accurate information in  
529 future microbial ecology studies of anaerobic digesters, efforts in targeting the active cells like  
530 active cell sorting in flow cytometers and later sequencing (Nakamura et al., 2016) or RNA-  
531 based sequencing (De Vrieze et al., 2017) would be required.

## 532 **Conclusions**

533 A microbial core has been elucidated in this study from four different scenarios for raw  
534 microalgae conversion into biogas. The high presence of several Anaerolineaceae members  
535 highlights the importance of saccharolytic and peptidic hydrolysis and fermentation. The  
536 dominance of *Smithella* and *Methanosaeta* suggest the relevant role of syntrophic and  
537 methanogenic pathways for bioenergy production from raw microalgae. This association was

538 more important during co-digestion than when only microalgae was digested, probably because  
539 of the composition of primary sludge. Nonetheless, no significant change in the acclimated  
540 communities was observed during microalgae shift from *Scenedesmus* to *Chlorella*. Instead,  
541 the microbial core was maintained over time in both AnMBR.

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#### 552 **Authors contributions**

553 NZL: conception and design, analysis and interpretation of the data, drafting of the article,  
554 collection and assembly of data. DA: statistical expertise. DA and LB: critical revision of the  
555 article for important intellectual content, analysis and interpretation of the data. AS: provision  
556 of study materials or patients and obtaining of funding. All authors: final approval of the article

#### 557 **Supplementary data**

558 Supplementary data associated to the present study can be found in the digital version of this  
559 manuscript

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