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Microbial community and antibiotic resistance profiles of biomass and effluent are distinctly affected by antibiotics addition to an anaerobic membrane bioreactor

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Anaerobic membrane bioreactors (AnMBRs) have the potential to significantly improve upon the efficiency of conventional wastewater treatment. However, emerging microbial contaminants, such as antibiotic resistance genes, remain understudied in these systems. The present study shows that influent antibiotics can alter both microbial communities and ARG profiles of AnMBR effluents, which can affect downstream spread of ARGs to potentially pathogenic bacteria.

1 **Microbial community and antibiotic resistance profiles of biomass and effluent**
2 **are distinctly affected by antibiotics addition to an anaerobic membrane**
3 **bioreactor**

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24 **Abstract**

25 The transfer of antibiotic resistance to pathogenic bacteria is one of the most eminent human
26 health threats and a concern in water reuse schemes. Anaerobic membrane bioreactors
27 (AnMBRs) are an emerging wastewater treatment biotechnology that have significant
28 potential for mainstream wastewater treatment. However, AnMBR effluents remain largely
29 unexplored with respect to their microbial community composition and their antibiotic
30 resistance profiles. In this study, we operated a bench-scale AnMBR for the treatment of
31 domestic wastewater containing antibiotics (250 µg/L each of sulfamethoxazole, ampicillin,
32 and erythromycin) and evaluated microbial community structure and antibiotic resistance
33 gene (ARG) dynamics in both the biomass and effluent. Results showed that ARG abundances
34 in the biomass of the AnMBR consistently increased throughout the experiment, while the
35 effluent ARG abundances saw a sharp increase upon initial antibiotics exposure to the system
36 and then dropped immediately thereafter. Further, a vastly more variable microbial
37 community was observed in the AnMBR effluent as compared to the biomass. Several
38 potentially pathogenic genera in the effluent were strongly correlated with the abundance of
39 specific resistance genes (e.g., *sul1*), as well as a class 1 integrase gene (*intl1*). Overall, results
40 of this study provide useful insights into the association of ARGs with microbial community
41 dynamics in AnMBR, which is needed to devise operational and design strategies to lessen
42 dissemination of antibiotic resistance to the environment.

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45 Key words: antibiotic resistance; microbial community; pathogenic bacteria; anaerobic
46 membrane bioreactor

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48 **1. Introduction**

49 Antibiotic resistance is an issue of crucial concern as one of the most imminent human health
50 risks. In the US, antibiotic resistance is currently responsible for over \$20 billion in excess
51 health costs and 8 million additional hospital days.¹ The proliferation of antibiotic resistance
52 in the environment and clinical settings is dictated by the dissemination of antibiotic
53 resistance genes (ARGs).² Resistance spreads due to the selective pressures raised by
54 antimicrobial compounds and through vertical and horizontal gene transfer (VGT and HGT)
55 mechanisms among and across different bacterial populations, respectively. Recent studies
56 have revealed a considerable increase in the number of bacterial species capable of resisting
57 different antibiotic classes.^{1, 3, 4}

58 Wastewater treatment plants (WWTPs) are known reservoirs of ARGs. Studies have
59 revealed that variable concentrations of various ARGs are released daily from WWTPs to the
60 environment, sometimes at levels that can even exceed those observed in influent
61 wastewater.⁵ Although wastewater can be considered a resource of energy, water, and
62 nutrients, recovering these resources must be balanced with protecting the downstream
63 environment, including mitigating emerging threats such as antibiotic resistance.⁶ Anaerobic
64 membrane bioreactors (AnMBRs) are an emerging biotechnology that can recover energy
65 from wastewater via biogas production and reduce residuals production, while also
66 potentially playing a role in lessening antibiotic resistance dissemination.⁷⁻⁹ Given that the
67 biomass of WWTPs contains both a broader range and higher concentrations of ARGs than do
68 WWTP effluents,⁴ the vastly lower biomass production of AnMBRs as compared to
69 conventional aerobic WWTPs has the potential to significantly reduce the overall release of
70 ARGs to the environment. Further, microorganisms and larger mobile genetic elements

71 (MGEs) can be effectively retained in the reactor via the membrane barrier and its associated
72 biofilm layer, reducing dissemination of antibiotic resistance in the effluent.⁸

73 Despite these potential advantages, the fate of antibiotics and ARGs in AnMBRs
74 remains understudied. In a previous study, we demonstrated that individual influent
75 antibiotics at a range of concentrations (10 to 250 µg/L) can significantly alter the abundance
76 of both related and unrelated ARGs in the biomass and effluent of AnMBRs.⁸ It has been
77 reported elsewhere that the impact of a mixture of antibiotics on ARG profiles can be
78 considerably greater than an individual antibiotic.^{10, 11} Therefore, in the present study we
79 investigated the influence of a mixture of antibiotics on the ARG profile of both the biomass
80 and effluent of an AnMBR. Although no study to date has been able to verify the reasons for
81 variation of ARG profiles in the presence of antibiotics,^{12, 13} possible explanations could
82 include HGT,^{12, 14} changes in microbial community abundance,^{11, 15} or some combination of
83 the two. HGT is known to be a critical factor for dissemination of antibiotic resistance in the
84 environment¹⁴ and is of particular concern in WWTPs: namely, ARGs can spread via HGT from
85 non-harmful bacteria to more virulent pathogenic species, thus posing a serious human
86 health risk. In addition to HGT, the fate of ARGs during wastewater treatment is also
87 intricately connected to microbial community dynamics. Although previous studies have
88 demonstrated relationships between biomass microbial communities and their
89 accompanying ARG profiles during the anaerobic digestion of sewage sludge¹⁶, no studies to
90 date have investigated such associations in mainstream anaerobic wastewater treatment
91 systems like AnMBRs. Further, the effluent of AnMBRs may contain resistant pathogenic
92 bacteria, posing a serious health risk in water reuse schemes. Therefore, a comprehensive
93 investigation of associations between microbial community dynamics and ARGs in AnMBR
94 effluents is needed. Consequently, in the present study, both the microbial communities and

95 the ARG profiles of the biomass and effluent of an AnMBR were investigated before, during,
96 and after the addition of three antibiotics from different classes.

97 **2. Materials and methods**

98 **2.1. Configuration of bench-scale AnMBR**

99 Detailed information of the bench-scale AnMBR has been reported in our previous
100 study.⁸ Briefly, the AnMBR consisted of a continuously stirred-tank reactor (Chemglass Life
101 Science, Vineland, NJ) with a working volume of 5 L and three separate microfiltration silicon
102 carbide membrane modules (Cembrane, Denmark) submerged in the reactor. The effective
103 membrane area of each module was approximately 0.015 m² and the membrane pore size
104 was 0.1 μm. The AnMBR was seeded with sludge from a mesophilic anaerobic digester at the
105 Joint Water Pollution Control Plant (Carson, CA). The AnMBR was operated at 25°C and fed
106 with a synthetic wastewater representative of domestic wastewater in the US (ESI Table S2).¹⁷
107 The experiments of the present study commenced two month after the end of the previous
108 study. Membrane modules were chemically cleaned both after the end of the prior
109 experiment and before the commencement of the current study's experiment. Between
110 cleanings, the AnMBR was continuously operated with no antibiotics being added to the
111 influent. Steady-state performance of the AnMBR was reached by the end of this period
112 (defined as consistent COD removal of > 85%, stable biogas production and methane content
113 of > 60% over at least two weeks of operation). Five days after confirming steady-state
114 operation, three antibiotics that included sulfamethoxazole (SMX, a sulfonamide),
115 erythromycin (ERY, a macrolide), and ampicillin (AMP, a β-lactam), were simultaneously
116 added to the influent of the AnMBR at a concentration of 250 μg/L each for a period of one
117 month. Although the estimated total antibiotic concentration in domestic wastewater is
118 around 50 μg/L,¹⁸ a higher non-lethal concentration¹¹ was used in the present study to

119 emphasize the antibiotic selective pressure impact on microbial community. The present
120 study was divided into three periods: pre-antibiotics (defined as the time after steady
121 performance was reached and before antibiotic addition), antibiotics loading, and post-
122 antibiotics. To monitor the performance of the AnMBR, mixed liquor suspended solid (MLSS),
123 mixed liquor volatile suspended solid (MLVSS), chemical oxygen demand (COD), biogas
124 production, and methane content of biogas were measured continuously during the
125 experimental period, as described previously.⁸ Additional details on the system and
126 operational parameters were also described previously.⁸

127 **2.2. Quantification of ARGs by qPCR**

128 Quantitative polymerase chain reaction (qPCR) was performed to quantify targeted
129 ARGs using procedures described previously.⁸ For biomass ARG profiles, 2 mL of mixed liquor
130 was collected biweekly, centrifuged, and decanted. For effluent ARG profiles, 50 mL of
131 permeate was freeze-dried using a lyophilizer (FreeZone 2.5 Liter Freeze-Dryer, Labconco,
132 Kansas City, MO). Both biomass and effluent samples, were then stored at -80°C prior to DNA
133 extraction. DNA extraction was conducted using the Maxwell 16 Blood DNA Purification kit
134 (Promega, Madison, WI), recommended by the manufacturer for wastewater sludges,
135 according to manufacturer instructions. qPCR was performed using a LightCycler 96 (Roche,
136 Basel, Switzerland) targeting 8 ARGs commonly found in domestic wastewater^{19, 20} including
137 genes conferring resistance to sulfonamides (*sul1* and *sul2*), macrolides (*ermB* and *ermF*), β -
138 lactams (*ampC* and *oxa-1*), and tetracycline (*tetO* and *tetW*), as well as a class 1 integrons-
139 associated gene (*int1*). Due to variable operon copy numbers of 16S rRNA in bacteria, a single
140 copy molecular marker gene (*rpoB*) was selected for ARG normalization to avoid this bias²¹.
141 Details of thermal cycling and primers for each targeted gene were provided in ESI Table S3.

142 **2.3. Quantification of antibiotics by LC-MS**

143 Ten mL of influent and effluent samples were filtered through 0.2 μm PTFE syringe
144 filters (Whatman, GE Healthcare, UK) and then stored at 4°C for no more than 3 days prior to
145 analysis. Antibiotics quantification was performed using direct injection liquid
146 chromatography electrospray ionization tandem mass spectrometry (LC-ESI-MS) on a 6560
147 Ion Mobility Quadrupole Time-of-Flight (IM-QTOF) LC-MS system (Agilent Technologies, Santa
148 Clara, CA) using 1290 Infinity UHPLC, Dual Agilent Jet Stream (ASJ), and EclipsePlus C18
149 column (2.1 mm; 50 mm; 1.8 μm). Details on method development and calibration protocols
150 are provided in ESI.

151 **2.4. Microbial community analysis**

152 Biomass and effluent extracted DNA samples were sent for sequencing at the
153 Microbial Systems Molecular Biology Laboratory (University of Michigan, Ann Arbor, MI)
154 where library preparation and sequencing was performed on the Illumina MiSeq platform
155 using the MiSeq Reagent Kit V2 (2x250 bp reads). To amplify 16S rRNA gene targets, a
156 universal 16S rRNA gene primer set targeting the V4 region was used as described
157 previously.²² High-throughput sequencing results were then analyzed using Mothur v.1.42.1,
158 with Silva 132 reference database for alignment and Ribosomal Database Project (RDP)
159 reference taxonomy for classification (rarefaction curves are provided in ESI Fig. S1). A non-
160 metric multidimensional scaling (NMDS) plot was conducted to present the distance between
161 samples. To evaluate the richness and evenness of samples, the Inverse Simpson index was
162 calculated for each operational phase to serve as a diversity index.

163 **2.5. Data analysis**

164 Analysis of molecular variance (AMOVA)²³ was employed to determine the statistical
165 significance of temporal changes in the microbial community data set. Unweighted principal
166 component analysis (PCA) was conducted using XLSTAT to compare the distance of different

167 ARG profiles. To determine the significance of linear relationships between ARG abundances
168 and microbial community profiles, Spearman's correlation was conducted using MAXSTAT Pro
169 3.6 over a 95% confidence interval. Strong positive correlation was determined based on P-
170 value and Spearman's coefficients (ρ) where $p < 0.05$ and $\rho > 0.7$.

171 **3. Results and Discussion**

172 **3.1. AnMBR system performance was robust during antibiotics addition**

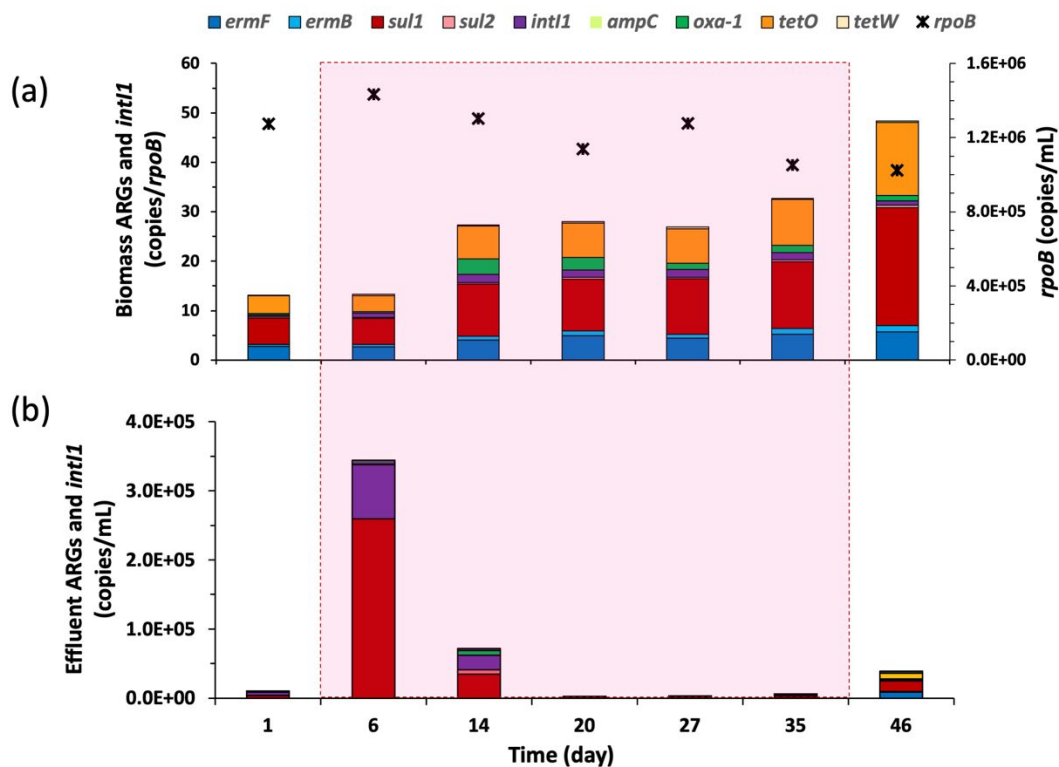
173 The AnMBR was operated at a hydraulic retention time (HRT) of 16 h and solids
174 retention time (SRT) of 300 d. Permeate flux was 7 L/m²h (LMH) and transmembrane pressure
175 (TMP) remained lower than 25 kPa throughout the experimental period. MLSS and MLVSS
176 were relatively constant at 10.6 ± 1.3 and 9.6 ± 0.3 g/L, respectively. Total COD in the influent
177 of the AnMBR averaged 453 ± 32 mg/L. COD removal was $90.0 \pm 1.8\%$ throughout operation,
178 with antibiotics addition having no significant effect on removal rate. Total biogas production
179 and methane content were also stable during the experimental period, averaging 736 ± 20
180 and 536 ± 14 mL/d, respectively (ESI Fig. S2a). Details of COD mass balance in the reactor
181 were provided in ESI Fig. S6.

182 Performance of the AnMBR with regards to antibiotics removal in the effluent (ESI Fig.
183 S2b) was similar to a previous study⁸ which explored separate addition of the same antibiotics
184 (AMP and SMX removal range of 89-98% and 69-78%, respectively). One exception was for
185 the case of ERY, which had a slightly lower removal rate in the present study than previously
186 observed (removal of 40-58% versus 67-88% in the previous study). In the previous study
187 antibiotics were added individually to the AnMBR and at incremental concentrations. Thus, it
188 is possible that the simultaneous addition of SMX, ERY, and AMP at higher concentrations
189 ($250 \mu\text{g/L}$) from the first day of the experiment in the present study resulted in less effective
190 ERY biotransformation in the system. This may have occurred due to the microbial community

191 requiring longer acclimation time under the elevated mixed antibiotics conditions.
192 Erythromycin has shown both higher variability and lower removal rates among mainstream
193 anaerobic treatment systems, in general, as compared to the other antibiotics used in this
194 study.²⁴

195 **3.2. AnMBR biomass was dominated by different ARGs than the AnMBR effluent**

196 Biomass ARGs were normalized against the *rpoB* gene (as gene copies/*rpoB*), while,
197 due to the potential significance of extracellular ARGs, effluents were normalized against
198 volume (as gene copies/mL). Despite differing normalization strategies, biomass and effluent
199 ARG profiles and temporal trends were markedly distinct (Fig. 1). The only exception was day
200 46 (post-antibiotics period), where the abundance of all effluent ARGs increased significantly.
201 The increase in effluent ARGs after the antibiotics loading period could be due to regrowth of
202 microorganisms harboring ARGs that were non-functional due to regulation and codon usage
203 bias, which can result in simultaneous ARG presence and antibiotic sensitivity.^{25, 26} SMX and
204 ERY are bacteriostatic antibiotics, halting growth but not necessarily lysing cells, and the only
205 two antibiotics present in significant concentrations in the effluent.²⁷ Growth of sensitive
206 ARG-harboring microorganisms may have been inhibited during the antibiotics phase,
207 resulting in low detection in the ARG profile. When the antibiotic selective pressure was
208 removed, the inhibited microorganisms may have increased growth rate, resulting in the
209 observed increase in ARG abundances.



210

211 **Fig. 1.** Abundance of targeted genes in the (a) biomass (copies/*rpoB*) and (b) effluent
 212 (copies/mL) of the AnMBR throughout the experimental period. The x-axis represents the
 213 days after steady performance of the AnMBR was reached. Day 1 represents pre-antibiotics
 214 period, days 6, 14, 20, 27 and 35 represent antibiotics loading period (area bordered by red
 215 dashed line), and day 46 represents post-antibiotics period. The markers in (a) represent
 216 abundance of *rpoB* (copies/mL; secondary y-axis).

217 *sul1*, *tetO*, and *ermF* were the most dominant ARGs in the biomass profile, each
 218 respectively accounting for $41.3 \pm 3.9\%$, $26.6 \pm 2.4\%$ and $16.7 \pm 3.1\%$ of targeted ARGs. Except
 219 for *sul1*, which was also the most abundant targeted ARG in the AnMBR effluent profiles
 220 (accounting for $53.6 \pm 12.5\%$), the dominant ARGs of the biomass (*tetO* and *ermF*) accounted
 221 for less than 1% of total targeted genes in the effluent (except for day 46, as described above).
 222 Alternatively, *intI1* was the second most abundant gene in the effluent profile, accounting for
 223 $31.0 \pm 14.6\%$ of targeted ARGs. Abundances of targeted ARGs throughout the experimental

224 period in the biomass and effluent of the AnMBR are provided in ESI Table S4 and Table S5,
225 respectively.

226 **3.2.1. ARG abundance in AnMBR biomass increased throughout operation**

227 After antibiotics addition on day 5, the abundance of class 1 integrons increased
228 significantly in the biomass profile. However, no significant changes were observed in ARG
229 abundances (Fig. 1a). Class 1 integrons are associated with HGT due to their presence on
230 mobile genetic elements (MGEs) such as plasmids and transposons.²⁸ Thus, the significant
231 increase in the biomass abundance of *int11*, one day after antibiotics addition, could indicate
232 a considerable rise in the presence of MGEs (and HGT). This rise may correspond to an
233 increase in plasmid-based resistance within the biomass microbial community. This, along
234 with an increase in the abundance of resistant microorganisms due to the antibiotic selective
235 pressure, could have subsequently resulted in the significantly higher total biomass ARG
236 abundances observed after day 6. Specifically, a marked increase in total ARG abundances
237 from 13.3 ± 1.1 on day 6 to 27.3 ± 1.7 copies/*rpoB* on day 14 was observed. Biomass ARGs
238 that increased on day 14 were *ermF*, *ermB*, *sul1*, *sul2*, *oxa-1*, and *tetO*. Increases in the
239 abundance of the aforementioned ARGs under selective pressure of high antibiotics has been
240 previously reported elsewhere.^{8, 11} After the initial increase, abundance of quantified biomass
241 ARGs remained approximately constant at 28.7 ± 2.6 copies/*rpoB* during the antibiotics
242 loading period. However, biomass ARGs further increased significantly to 48.3 ± 3.4
243 copies/*rpoB* ten days after ceasing addition of antibiotics (day 46). It should be noted that
244 with one sample for the post-antibiotics period, it is not possible to conclusively evaluate the
245 response of the ARG profiles to removing the antibiotics selective pressure. Ultimately, both
246 the initiation of antibiotics addition and its subsequent cessation at the final stage of the
247 experiment likely corresponded with marked increases in total biomass ARG levels.

248 **3.2.2. AnMBR effluent ARG abundance spiked upon initial antibiotics exposure**

249 Quantified effluent ARG abundances increased approximately 34-fold one day after
250 antibiotics addition, primarily due to increases in *sul1* gene abundance, along with the class 1
251 integrons-associated *int11* (Fig. 1B). Since, these two genes (*sul1* and *int11*) are commonly co-
252 located on conjugative plasmids, the cause of this phenomenon may relate to the fact that
253 plasmid/extracellular ARGs contribute significantly to the effluent ARG profiles in membrane-
254 based reactors. To explain this drastic increase in *sul1* and class 1 integrons, it can be inferred
255 that antibiotics addition induced a spike in the rate of HGT in the biomass that led to a
256 temporary increase in biomass extracellular plasmid DNA. This could have manifested as a
257 sharp increase in effluent extracellular plasmid DNA and, consequently, a significant rise in
258 abundance of *int11* and *sul1* genes in the effluent ARG profiles. We previously observed an
259 increase in *sul1* and *int11* abundance in AnMBR effluent during individual addition of
260 antibiotics (SMX, ERY and AMP).⁸ Further, it has been reported that antibiotic mixtures, as
261 opposed to individual antibiotics, result in more pronounced changes in ARG abundance.¹¹
262 Therefore, the drastic increase in *sul1* and *int11* abundance after simultaneous addition of
263 SMX, ERY, and AMP at a relatively high concentration of 250 µg/L is not unreasonable.
264 Another possible explanation for this drastic increase might be the considerable release of
265 MGEs due to the lysis of bacterial cells caused by AMP, which is bactericidal. Eight days after
266 the sharp spike, total abundance of effluent ARGs decreased by 5-fold, and then decreased
267 another 42-fold on day 20. Effluent ARG profiles remained approximately constant thereafter
268 for the rest of the antibiotics loading period. These reductions could be due to microbial loss
269 of MGEs such as plasmid DNA along with lower HGT rates after the initial antibiotics exposure.
270 Another possible explanation for the spike and subsequent reduction in total effluent ARG
271 abundances could be the development of the membrane fouling layer (biofilm) and

272 subsequent biofilm-based ARG removal.^{8, 29} However, given that the TMP was consistent
273 during the experimental period, it is possible that the influence of the biofilm on the effluent
274 ARG profile was negligible.

275 Comparing the ARG profiles in biomass with those of the AnMBR effluent, it is
276 noteworthy that both *tetO* and *ermF* genes (among the most abundant targeted ARGs in the
277 biomass) were hardly detected in the effluent. This likely implies that *tetO* and *ermF* gene
278 presence on extracellular MGEs was limited while bacteria harboring these ARGs were
279 concurrently retained in the AnMBR by membrane separation. However, this was not the case
280 for all genes. For instance, the significant increase in biomass abundance of the *oxa-1* gene
281 on day 14 corresponded with a considerable increase in its abundance in the effluent ARG
282 profile of the same day.

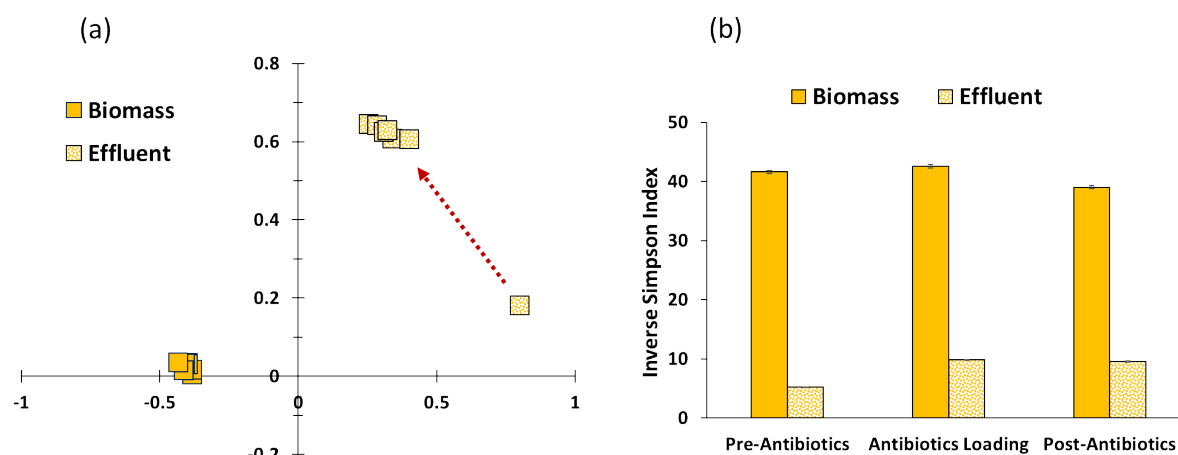
283 Following the same trend as the biomass ARG profile, effluent ARG abundances also
284 increased after cessation of the addition of antibiotics. Overall, ARG abundances in both
285 biomass and effluent were affected significantly by the addition of antibiotics to the influent.
286 However, the mechanisms that dictated the increases in effluent ARG abundances at the
287 stages of antibiotics addition and cessation, respectively, may have been vastly different from
288 each other. Specifically, the ARG increases seen at the beginning of antibiotics addition (days
289 6 and 14) were driven by different gene combinations (*sul1* and *int1*) than those that
290 dominated the gene profile after antibiotics addition ceased on day 46 (predominantly *tetO*,
291 *sul1*, and *ermF*). Interestingly, the effluent ARG profile of day 46 was also highly similar to that
292 of the biomass throughout the experiment (Fig. 1), this observation being confirmed by an
293 unweighted PCA performed on all of the ARG profiles (ESI Fig. S3). One possible explanation
294 for this phenomenon is that the effluent ARG increases on day 6 were caused by elevated
295 occurrence of HGT, while the increases on day 46 were the result of changes in the microbial

296 composition of the effluent after antibiotics cessation. Supporting this hypothesis, several
297 anaerobic biomass-associated microbial populations increased in their relative abundance in
298 the effluent line on day 46. These groups included *Syntrophomonas* and *Dechloromonas*
299 genera that showed strong correlation with biomass-dominating ARGs (including *ermF*, *ermB*
300 and *tetO*). Microbial dynamics and specific correlations are discussed in more detail in the
301 following sections.

302 **3.3. Microbial community analysis**

303 **3.3.1. Biomass relative abundances showed remarkable stability during antibiotics addition**

304 Similar to the ARG profiles, the microbial community structure in the biomass was
305 distinct from that of the effluent. The biomass community structure across the experimental
306 period was highly stable. This stability of the microbial community even after the addition of
307 multiple antibiotics at concentrations of 250 µg/L is further indication of the ability of AnMBRs
308 to sustain treatment of high antibiotic-containing wastewaters (e.g., hospital wastewaters).
309 NMDS and AMOVA were employed (Fig. 2a) to confirm similarity of the biomass microbial
310 community at the genus level throughout the experimental period. Results revealed that all
311 biomass community samples clustered closely together with no statistically significant
312 changes ($p = 0.081$). Based on the results of our previously published study,⁸ it was speculated
313 that variations in biomass ARG profiles during the addition of antibiotics to the influent could
314 have been primarily due to changes in the microbial community. However, given the high-
315 level of stability among community relative abundances before, during, and after antibiotics
316 addition observed in the present work, it is likely that these antibiotic-influenced changes to
317 the biomass ARG profile are primarily due to HGT and not microbial community alteration.
318 Occurrence of HGT in anaerobic digesters has been reported previously.^{12, 30}



319

320 **Fig. 2.** (a) Non-metric multidimensional scaling (NMDS) and (b) Inverse Simpson index for the

321 biomass and effluent of the AnMBR throughout the experimental period. The red arrow in

322 the NMDS plot indicates the significant shift of the effluent samples after antibiotics addition.

323 In the Inverse Simpson plot, the bars for antibiotic loading represent the average of the

324 diversity index in $n = 5$ samples during the loading period ($n = 1$ for the pre- and post-

325 antibiotics period). Error bars for the pre- and post-antibiotics period represent the standard

326 deviation calculated by Mothur for each sample, and for the antibiotics loading period

327 represents the standard deviations of the averages in $n = 5$.

328 Bacteroidetes ($28.6 \pm 1.96\%$), Chloroflexi ($19.7 \pm 1.2\%$), and Proteobacteria ($13.2 \pm$

329 1.0%) were the most abundant phyla in the biomass. At the family level (Fig. 3a), the microbial

330 community was comprised of $21.3 \pm 2.1\%$ unclassified Bacteroidetes and $11.0 \pm 0.8\%$

331 unclassified Chloroflexi. Only 25% of Bacteroidetes and 44% of Chloroflexi sequences were

332 classified at the family level. Anaerolineaceae ($8.59 \pm 0.50\%$), Syntrophaceae ($5.21 \pm 0.33\%$),

333 Ignavibacteriaceae ($4.50 \pm 0.27\%$), Methanoregulaceae ($3.70 \pm 0.20\%$), and

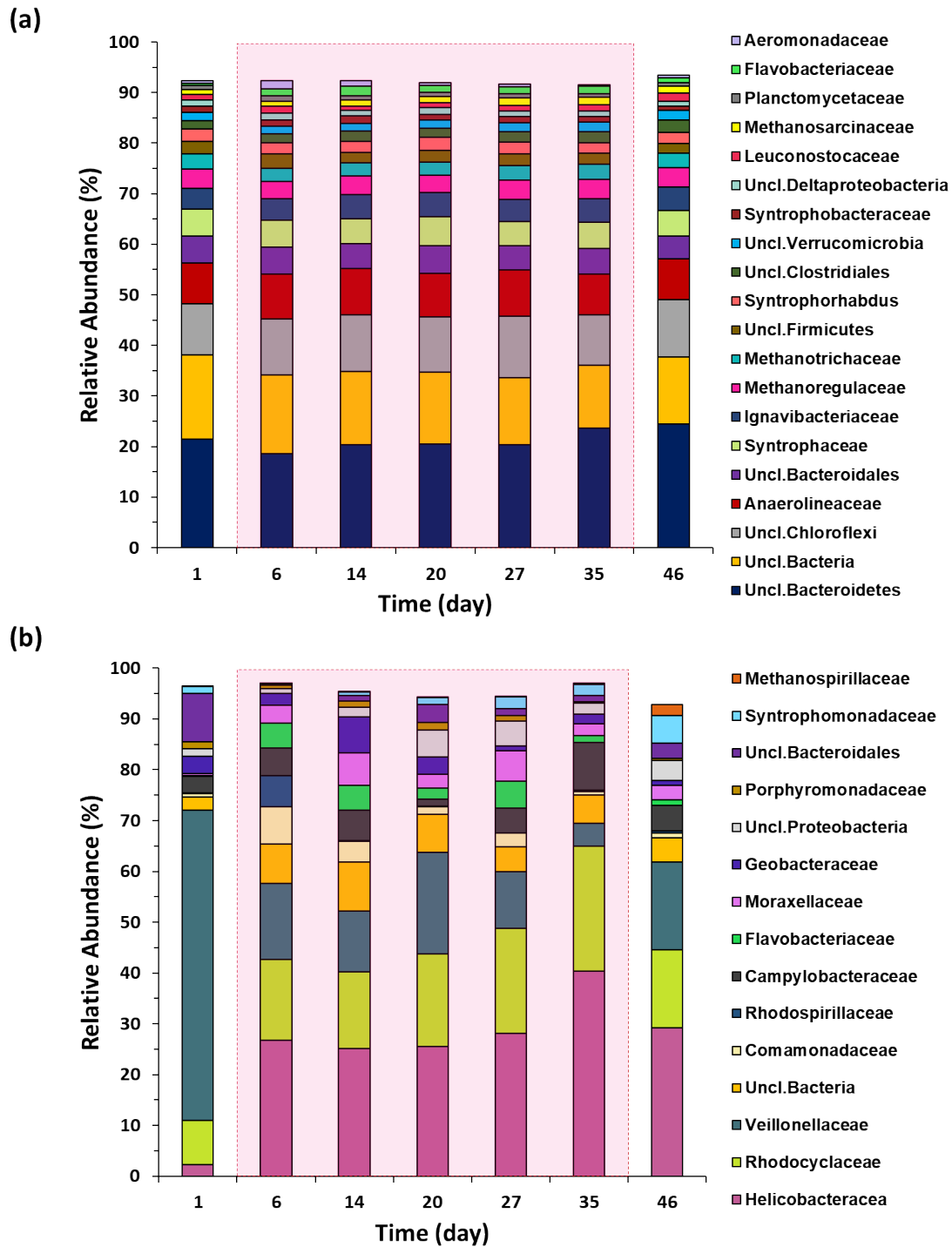
334 Methanotrichaceae ($2.77 \pm 0.18\%$) were the most abundant classified families in the biomass

335 community. Most genera associated with Anaerolineaceae are strict anaerobes that are

336 commonly found in anaerobic treatment system for domestic wastewater³¹ and anaerobic

337 digesters . The family Ignavibacteriaceae contains a single facultative anaerobic genus that is
338 capable of utilizing aromatic compounds³² and can also contribute to sulfide oxidation.³³

339 Relative abundances of both methanogens and syntrophic fatty-acid oxidizing bacteria
340 in the biomass remained stable throughout operation, averaging $8.68 \pm 0.45\%$ and $8.67 \pm$
341 0.55% relative abundance across the seven temporal samples (ESI Fig. S4). *Methanolinea*
342 ($3.73 \pm 0.19\%$) and *Methanosaeta* ($2.79 \pm 0.18\%$) were the most abundant methanogens,
343 indicating relatively comparable contribution of hydrogenotrophic (*Methanolinea*) and
344 acetoclastic (*Methanosaeta*) methanogenesis. The most abundant syntrophs were
345 unclassified Syntrophaceae and unclassified Syntrophorhabdaceae, with relative abundances
346 of $3.96 \pm 0.35\%$ and $2.30 \pm 0.20\%$, respectively. The family Syntrophaceae can oxidize long
347 chain fatty acids to produce acetate and hydrogen,³⁴ however, the family
348 Syntrophorhabdaceae mainly oxidizes aromatic compounds, such as benzoate.³⁵ Based on
349 AMOVA analysis, antibiotics addition did not significantly affect the relative abundance of
350 methanogens ($p = 0.108$) or their syntrophic counterparts ($p = 0.095$), indicating the
351 robustness of these keystone microbial populations regardless of influent antibiotics
352 concentration.



353

354 **Fig. 3.** Relative abundance of the (a) biomass and (b) effluent microbial community at the

355 family level throughout the experimental period. Day 1 represents pre-antibiotics period,

356 days 6, 14, 20, 27 and 35 represent antibiotics loading period (area bordered by red dashed

357 line), and day 46 represents post-antibiotics period.

358 **3.3.2. Dominant effluent microbial groups were affected significantly by antibiotics**
359 **exposure**

360 Effluent microbial communities were distinct from the biomass, as was clearly
361 elucidated in the NMDS plot and AMOVA analysis ($p = 0.001$) (Fig. 2a). This is not surprising,
362 owing to the fact that the 0.1 μm membranes used in this AnMBR system likely exclude
363 passage of nearly all microorganisms present in the biomass. However, the addition of
364 antibiotics to the influent did appear to significantly impact the effluent microbial community
365 structure (which was not the case for AnMBR biomass). Since transmembrane pressure was
366 consistent during the operational period at 19.3 ± 2.6 kPa, these changes in effluent
367 community structure were likely not due to membrane fouling. Firmicutes (63.1%),
368 Proteobacteria (22.5%), and Bacteroidetes (11.2%) were the most abundant phyla in the pre-
369 antibiotics periods, whereas the dominant phyla in the effluent during antibiotics addition
370 and in the post-antibiotics period were Proteobacteria ($69.4 \pm 7.7\%$), Firmicutes ($15.7 \pm 6.0\%$),
371 and Bacteroidetes ($6.47 \pm 2.02\%$). Since these effluent changes during antibiotics addition did
372 not concurrently change the biomass microbial community structure, further analysis of
373 microbial diversity was performed to compare the biomass and effluent communities. Results
374 revealed that the Inverse Simpson index of biomass sample in the pre-antibiotics period was
375 8 times higher than the effluent (Fig. 2b). This higher observed diversity indicates more
376 evenness and richness among the biomass microbial community structure as compared to
377 the effluent. The lower diversity of the effluent microbial community may have made this
378 community more susceptible to inhibition, resulting in the high variability observed upon
379 antibiotics addition.

380 At the family level (Fig. 3b), Veillonellaceae (61.07%) was the most abundant
381 population in the pre-antibiotics period. A significant decrease to $13.3 \pm 5.5\%$ relative

382 abundance during the antibiotics loading and post-antibiotics periods suggests that members
383 of this family are highly susceptible to inhibition from one or more of the introduced
384 antibiotics. In contrast, the relative abundance of Helicobacteraceae, a sulfur-oxidizing family,
385 significantly increased after addition of antibiotics from 2.24% in the pre-antibiotics period to
386 $29.2 \pm 5.7\%$ during the antibiotics loading and post-antibiotics periods. Selection of the
387 Helicobacteraceae family after antibiotics addition suggests a high likelihood of multi-drug
388 resistance. Rhodocyclaceae, another prominent member of the effluent community,
389 increased to $18.3 \pm 3.8\%$ after antibiotics addition. Within Rhodocyclaceae, *Zoogloea* is an
390 aerobic genus that has been commonly found in activated sludge systems.³⁶ The enrichment
391 of this genus in the effluent (ESI Fig. S5) might be an indication that aerobic bacteria likely
392 increased their presence over time in the effluent tubes (which discharged to the open air).
393 However, enrichment of *Zoogloea* in the effluent of anaerobic reactors has been reported
394 frequently.^{13, 37} Further, several effluent genera have been routinely isolated from biofilm
395 samples: *Novispirillum* (Rhodospirillaceae), *Arcobacter* (Campylobacteraceae), *Comamonas*
396 (*Comamonadaceae*), and *Aquabacterium* (*Comamonadaceae*).³⁸⁻⁴¹ These genera are mostly
397 facultative anaerobes or capable of growth under anaerobic conditions. Therefore, their
398 presence in the effluent may be due to seeding and regrowth as a result of membrane
399 permeation by even a relatively small number of bacteria.

400 **3.4. Correlations between ARGs and microbial community structure indicate a potential for** 401 **HGT**

402 To further investigate associations of ARGs with microbial communities in the AnMBR,
403 a Spearman's correlation analysis was performed between ARG abundances and relative
404 abundance of operational taxonomic units (OTUs) for the biomass and effluent samples (ESI
405 Table S6). We elected to use an out-based approach for correlation analysis due to the large

406 proportion of unclassified sequences at the genus level. Network analysis was used to
407 illustrate statistically significant strongly positive correlations (Fig. 4) due to the implications
408 that positive ARG-microbial correlations have on ARG association with specific bacterial
409 groups.⁴² The analysis revealed numerous strong positive correlations between certain ARGs
410 and microbial groups (OTUs) in the biomass. Although statistically significant positive
411 correlation between a microbial group and a particular ARG cannot be considered as evidence
412 of that group carrying antibiotic resistance, it can indicate the OTUs that are potential host
413 bacteria for ARGs and MGEs.⁴³ Several of the OTUs in the AnMBR biomass showed strong
414 correlations with multiple ARGs. Therefore, these groups may have a greater likelihood of
415 serving as potential multi-resistant host bacteria. Based on this and Table S6, OTU6
416 (Bacteroidetes), OTU32 (Clostridiales), OTU35 (Verrucomicrobia), OTU84 (Firmicutes),
417 OTU105 (Anaerolineaceae), and OTU134 (Bacteroidetes) in the biomass microbial community
418 were identified as potential multi-resistant host bacteria for *ermF*, *ermB*, *sul1*, *sul2*, *ampC*,
419 and *tetO* genes. Occurrence of multi-drug resistant bacteria in anaerobic environments, such
420 as anaerobic digesters, has been reported previously via strong correlations between
421 microbial community structure and ARG abundance.^{15, 44}

422 Effluent correlations bear strong implications to the bacterial types that are actually
423 entering the environment through effluent discharge and/or reuse, and therefore are of
424 particular interest. The correlation analysis resulted in the effluent microbial communities
425 being divided into five distinct groupings: OTUs with strong correlations to (1) *sul1* and *int11*,
426 (2) *sul2* and *oxa-1*, (3) *ermF*, *ermB* and *tetO*, (4) *tetW*, and (5) genera with no correlations to
427 ARGs. Based on Table S5, Group 1 included OTU25 (Bacteria), OTU27 (*Arcobacter*), OTU66
428 (*Novispirillum*), OTU78 (*Comamonas*), OTU89 (*Flavobacterium*), OTU159 (*Acetobacteroides*),
429 OTU247 (Bacteria), OTU264 (*Rhodococcus*), OTU352 (*Caulobacter*), OTU441 (*Sulfuricurvum*),

430 OTU501 (Caulobacteraceae), OTU550 (*Chryseobacterium*), and OTU648 (*Bdellovibrio*). It is
431 possible that these groups can serve as potential hosts for *sul1*-combining class 1 integrons,
432 thus also being implicated in HGT. *sul1* and class 1 integrons have been previously found on
433 the same gene cassettes in different *Arcobacter*⁴⁶ and *Comamonas*⁴⁷ species in wastewater
434 and soil samples. Strong correlation of *Flavobacterium* with these two genes has also been
435 reported in drinking water and soil samples.⁴⁸⁻⁵⁰ Perhaps most importantly, 7 OTUs of Group
436 1 are known to harbor pathogenic species. The presence of class 1 integrons in other
437 pathogenic bacterial strains, including *Escherichia coli*,^{51, 52} *Pseudomonas aeruginosa*,⁵³ and
438 *Salmonella* spp.,⁵⁴ is well established. Thus, the observed dominance of Group 1 by potentially
439 pathogenic groups might be an indication of the transferability of such MGEs to a broader
440 range of clinically significant bacterial strains. However, one of the limitations of the present
441 study was the inability to classify microbial groups at the species-level for identification of
442 pathogens.

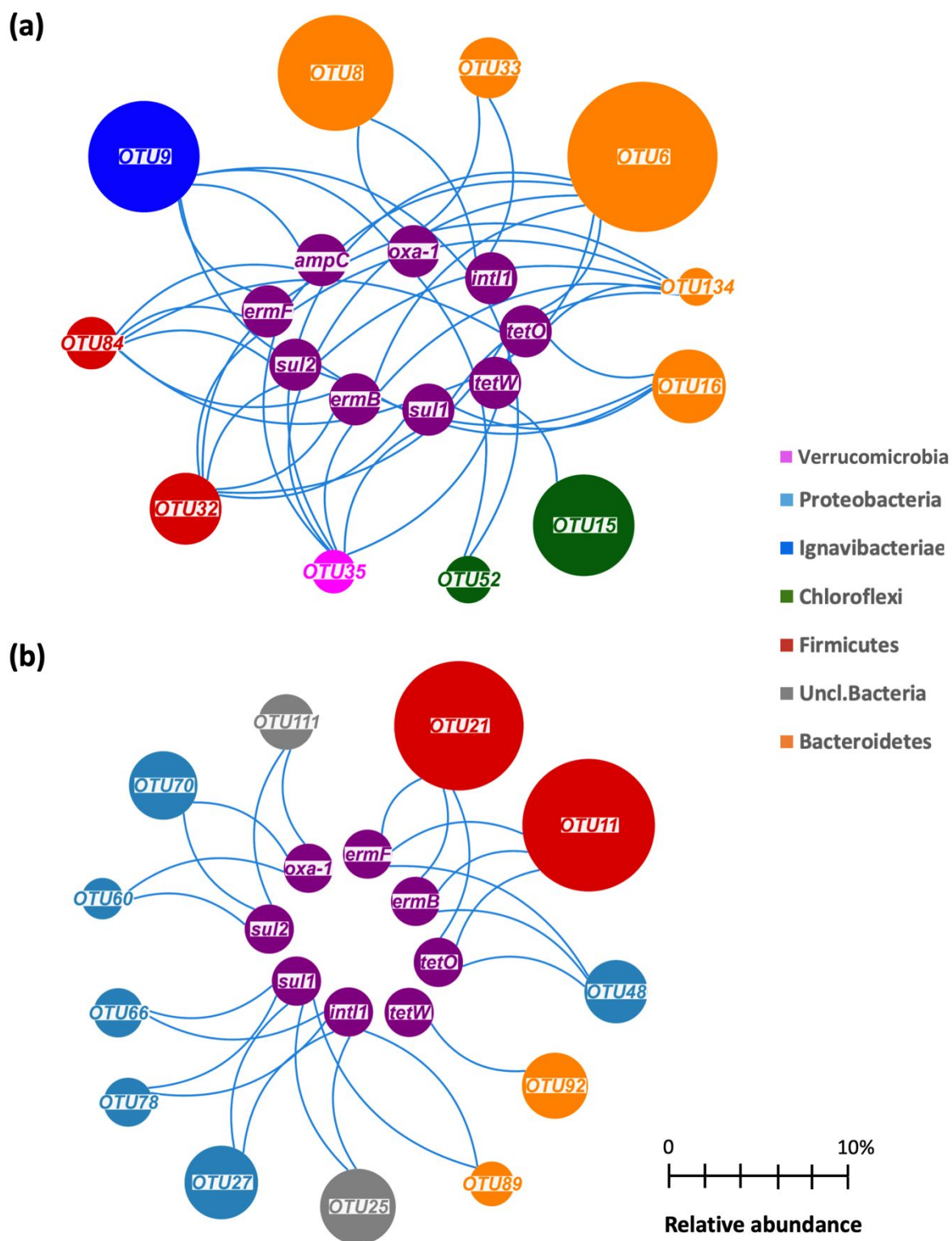
443 Group 2 (showing strong correlation to *sul2* and *oxa-1*) included OTU60
444 (Comamonadaceae), OTU70 (*Acinetobacter*), OTU111 (Bacteria), OTU163 (*Methylophilus*),
445 OTU209 (*Desulfobulbus*), OTU287 (*Desulfovibrio*), OTU360 (Cryomorphaceae), OTU384
446 (Spirochaetaceae), and OTU499 (*Bosea*). Among group 2, some species of *Acinetobacter*,
447 *Desulfovibrio*, Cryomorphaceae, Spirochaetaceae and *Bosea* might act as opportunistic
448 pathogens. *Acinetobacter* spp. are among the six most important multi-drug resistant
449 organisms in hospitals.⁵⁵ They are also frequently used as an antibiotic resistance indicator in
450 water and wastewater.⁵⁶ The presence of *sul2*⁵⁷ and *oxa-1*⁵⁸ genes in different isolates of
451 *Acinetobacter* spp. in wastewater samples has been reported previously. Although no
452 putative association between the other members of group 2 and *oxa-1* have been reported
453 in literature, in silico analysis of several Comamonadaceae species has revealed the presence

454 of open reading frames (ORFs) that correspond to OXA genes.⁵⁹ Further, strong correlation of
455 *sul2* genes with Comamonadaceae in treated wastewater,⁶⁰ *Methylophilus* in a soil microbial
456 fuel cell,⁶¹ and *Desulfobulbus* during anaerobic digestion of cattle manure⁶² have also been
457 reported. Given that the co-location of *sul2* and *oxa-1* genes on plasmids has also been
458 previously reported for multiple pathogenic strains,^{63, 64} the observed correlations of group 2
459 with these genes might suggest their potential transferability to such species in the effluent.

460 Group 3 (strongly correlated with *ermF*, *ermB* and *tetO*) genera included OTU11
461 (*Anaerosinus*), OTU21 (*Propionispira*), OTU48 (*Dechloromonas*), OTU178 (Rhodocyclaceae),
462 OTU184 (*Comamonas*), OTU299 (*Rhodopseudomonas*), and OTU389 (Bacteria). Strong
463 correlations of *ermF* and *ermB* genes with a different tetracycline resistance gene (*tetX*) have
464 previously been reported in multiple aquatic environments.^{44, 65, 66} Further, the co-location of
465 *ermB* and *tetO* genes in multiple *Enterococcus* spp. has been observed elsewhere.⁶⁷ Co-
466 occurrence of *ermF* and *ermB* genes with *Comamonas*,¹³ which is typically regarded as a
467 pathogenic resistant bacteria,⁶⁸ has also been reported previously.

468 The forefront of the threat of antimicrobial resistance is specifically related to its
469 ultimate transfer to pathogenic bacteria that can reach the environment. Thus, the dynamics
470 of ARGs and MGEs in wastewater effluent microbial communities that are released to
471 downstream water bodies and in reuse applications are of critical concern. This concern is
472 further exacerbated when effluents contain significant levels of multiple pathogenic
473 populations that can acquire these resistance elements through HGT. Correlation analysis of
474 ARGs with microbial communities in the AnMBR effluent revealed some possible associations
475 of potential pathogenic groups with at least one ARG. This is relevant, considering that several
476 of these genera were correlated with Class 1 integrons, which implies a strong potential for
477 HGT in the effluent environment. However, when putting these observations in the context

478 of AnMBR systems, specifically, it is important to note that previous studies have reported
479 significantly lower effluent abundances of pathogenic species in AnMBR effluents than in the
480 effluents of their aerobic MBR counterparts.⁶⁹ In the present study, a sharp decrease in
481 abundances of both ARGs and potentially pathogenic genera in the effluent were observed
482 after Day 6 of antibiotics addition and remained relatively low until the end of the
483 experimental period. This suggests that pathogenic and/or resistant bacteria in the effluent
484 were not sustained beyond the initial antibiotics exposure adjustment period. However, It is
485 noteworthy to state that since the community analysis of the present study was not classified
486 at the species level, no certain claims can be made about actual pathogenicity of identified
487 OTUs. Therefore, future research is needed to investigate AnMBR effluents with respect to
488 the association of antibiotic resistance with pathogenic species.



489

490 **Fig. 4.** Network analysis representing the positive correlations between ARGs (purple circles)491 and microbial structure (OTUs) with $\geq 0.5\%$ relative abundance in at least one sample in the

492 (a) biomass and (b) effluent of the AnMBR. A connection shows strong significant positive
493 correlation ($p < 0.05$; and $\rho > 0.7$). The bubble size is indicative of relative abundance.

494 **4. Conclusions**

495 AnMBRs are an emerging biotechnology for mainstream wastewater treatment with
496 the potential to enhance energy efficiency and effluent reuse, while also theoretically
497 lessening the spread of antibiotic resistance to the environment. In this study, a bench-scale
498 AnMBR was employed to treat domestic wastewater containing antibiotics and investigate
499 the association of microbial communities with ARG profiles in both biomass and effluent of
500 the AnMBR. The main conclusions of the experiment are as follows:

- 501 • Performance of the AnMBR regarding COD removal, biogas production, and methane
502 yield was robust during the simultaneous addition of three antibiotics at 250 $\mu\text{g/L}$.
- 503 • ARG profiles and temporal trends in the biomass of the AnMBR were markedly distinct
504 from those of the effluents.
- 505 • Effluent ARG abundance spiked upon initial antibiotics exposure, mostly due to the
506 significant increase in *sul1* and class 1 integrons. It then gradually decreased by around
507 167-folds and remained constant during the rest of antibiotics loading period.
- 508 • Biomass microbial community structure was unaffected by antibiotics addition and
509 was relatively uniform throughout the experimental period.
- 510 • Antibiotics addition significantly influenced effluent microbial community structure.
- 511 • Correlation analysis revealed the existence of potential multi-resistant host bacteria
512 in the biomass, while also showing that the effluent microbial community contained
513 distinct groups of bacteria with varied potential mechanisms of resistance.

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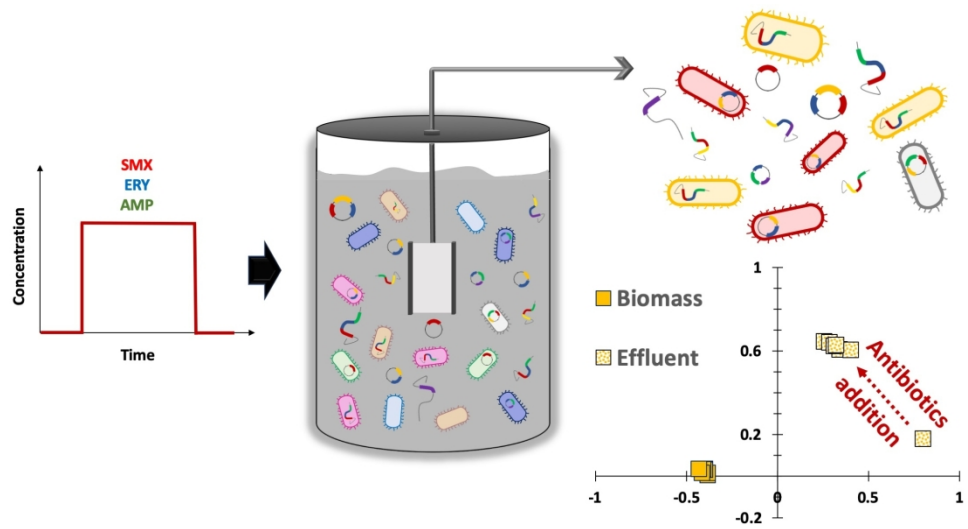
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