

1 **Targeted manipulation of abundant and rare taxa in the *Daphnia magna* microbiota with**  
2 **antibiotics impacts host fitness differentially**

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4 Running title: Targeted microbiota manipulation impacts fitness

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24 **Abstract**

25 Host-associated microbes contribute to host fitness, but it is unclear whether these contributions  
26 are from rare keystone taxa, numerically abundant taxa, or interactions among community  
27 members. Experimental perturbation of the microbiota can highlight functionally important taxa;  
28 however, this approach is primarily applied in systems with complex communities where the  
29 perturbation affects hundreds of taxa, making it difficult to pinpoint contributions of key  
30 community members. Here, we use the ecological model organism *Daphnia magna* to examine  
31 the importance of rare and abundant taxa by perturbing its relatively simple microbiota with  
32 targeted antibiotics. We used sublethal antibiotic doses to target either rare or abundant members  
33 across two temperatures, then measured key host life history metrics and shifts in microbial  
34 community composition. We find that removal of abundant taxa had greater impacts on host  
35 fitness than did removal of rare taxa and that the abundances of non-target taxa were impacted by  
36 antibiotic treatment, suggesting no rare keystone taxa exist in the *Daphnia magna* microbiota but  
37 microbe-microbe interactions may play a role in host fitness. We also find that microbial  
38 community composition was impacted by antibiotics differently across temperatures, indicating  
39 ecological context shapes within-host microbial responses and effects on host fitness.

40 **Importance**

41 Understanding the contributions of rare and abundant taxa to host fitness is an outstanding  
42 question in host microbial ecology. In this study, we use the model zooplankton *Daphnia magna*  
43 and its relatively simple cohort of bacterial taxa to disentangle the roles of distinct taxa on host  
44 life history metrics, using a suite of antibiotics to selectively reduce the abundance of  
45 functionally important taxa. We also examine how environmental context shapes the importance  
46 of these bacterial taxa on host fitness.

47

## 48 **Introduction**

49 The microbes in and on host organism tissue, collectively referred to as the microbiome, are  
50 recognized as having important beneficial impacts for the host. Many functions have been tied to  
51 bacterial species in the microbiota, including nutrient acquisition for the host<sup>1</sup> and immune  
52 system priming<sup>2</sup>. As most species in host-associated microbiota are difficult to culture,  
53 experimental perturbation of the microbiota and subsequent sequencing combined with host  
54 fitness metric measurement is a commonly used set of methods to understand functional  
55 contributions of microbial taxa to host fitness<sup>3,4,5</sup>. To understand the impacts of individual taxa  
56 on host fitness, antibiotics can be chosen to selectively perturb taxa and fitness outcomes can be  
57 measured<sup>6</sup>. However, this approach is primarily used in systems with highly complex  
58 microbiomes, often with hundreds of interacting taxa impacted by these antibiotics<sup>7,8</sup>. While  
59 large-scale perturbations are necessary for understanding overall microbiome structure and  
60 broad-level interactions, fundamental questions about host-microbiome interactions can be  
61 addressed readily in systems with simpler microbial communities. For example, determining  
62 whether host fitness is affected more by overall microbiome diversity (number of distinct taxa)  
63 or functional diversity (taxa with distinct functions) is tractable in systems with fewer microbial  
64 taxa. Identifying the contribution of numerically abundant taxa to host function is also possible  
65 in these systems, as specificity of antibiotic targeting can be greater.

66

67 To better understand the relationship between specific taxa in the microbiota and host fitness we  
68 applied an antibiotic suppression technique in *Daphnia magna*, a widely used model organism in  
69 ecotoxicology<sup>9</sup>, population genomics<sup>10</sup>, and host-parasite dynamics<sup>11</sup>. *Daphnia magna*'s

70 microbiome is relatively simple, with only 10-15 amplicon sequence variants (ASVs)  
71 constituting >70% of relative abundance<sup>12,13</sup>. In particular,  $\beta$ -proteobacteria,  $\gamma$ -proteobacteria,  
72 and Sphingobacteriia are bacterial Classes consistently identified in the *Daphnia magna*  
73 microbiome across environments and genotypes<sup>14,15,16</sup>. Important contributions to host fitness  
74 may be directly linked to taxa in these Classes, as removal of the microbiota with broad-  
75 spectrum antibiotics has been directly linked to decreases in *Daphnia* growth, survival, and  
76 fecundity<sup>17,13,18</sup>. In particular, *Limnohabitans*, a highly abundant genus of  $\beta$ -proteobacteria, has  
77 been shown to benefit host fecundity<sup>17</sup>. However, no host fitness benefits have been directly  
78 linked to the other bacterial Classes prevalent in the *Daphnia magna* microbiome.  
79  
80 Functions provided by the microbiota to the host can be dependent on biotic and abiotic factors.  
81 Environmental factors like temperature<sup>19</sup>, pH<sup>20</sup>, and food availability and diet<sup>21</sup> alter microbiome  
82 composition and gene expression profiles of present taxa. Intrinsic tolerance differences among  
83 taxa or host-mediated selection for tolerant taxa may drive changes in community composition,  
84 which in turn could influence host fitness. We aimed to investigate this environmental factor-  
85 microbiome-host fitness interaction using temperature, because *Daphnia magna* live at a wide  
86 range of temperatures<sup>22</sup> and because temperature influences the *Daphnia magna* microbiota<sup>12,23</sup>.  
87 Here, we sought to understand which taxa were affected by environmental change using a cold,  
88 environmentally relevant temperature similar to that found in late fall, and whether impacted taxa  
89 contributed to host fitness.  
90  
91 To identify taxa in the *Daphnia* microbiota associated with specific host life history traits, we  
92 selected antibiotics that each suppressed two of the three major bacterial Classes. We used

93 aztreonam to suppress  $\beta$ -proteobacteria and  $\gamma$ -proteobacteria<sup>24</sup>; erythromycin to suppress  $\gamma$ -  
94 proteobacteria and Sphingobacteriia (specifically Bacteroidetes)<sup>25</sup>; and sulfamethoxazole to  
95 suppress  $\beta$ -proteobacteria and Sphingobacteriia<sup>26</sup>. We aimed to understand how antibiotic-  
96 induced changes to the microbiome impacted host fitness, linking changes in relative abundance  
97 to host fitness outcomes. Because the  $\beta$ -proteobacteria *Limnohabitans* impacts host fitness<sup>17</sup> and  
98 is present in high relative abundance in the *Daphnia magna* microbiota, we hypothesized that  
99 more abundant taxa contributed a greater share of functions impacting host fitness; as such,  
100 suppression of these more abundant taxa would reduce host fitness, specifically in reduced  
101 fecundity, survival, and growth. We also aimed to understand how the microbiome and its  
102 associated functions changed depending on environmental context. To do this, we raised  
103 *Daphnia* in cold temperatures. We hypothesized the differing environment would induce shifts in  
104 microbiota composition, as different microbial functions may be necessary to respond to the  
105 change. Finally, we combined environmental change with antibiotic treatments to see if  
106 reduction of abundant taxa caused differential shifts in host fitness across environments. Here,  
107 we hypothesized that reduction of taxa abundant in warmer temperatures with antibiotics would  
108 not cause as severe changes in host fitness in the colder temperature treatment, as taxa providing  
109 beneficial functions in a different environment would not be targeted by these perturbations and  
110 would have reduced competition from the now-suppressed taxa.

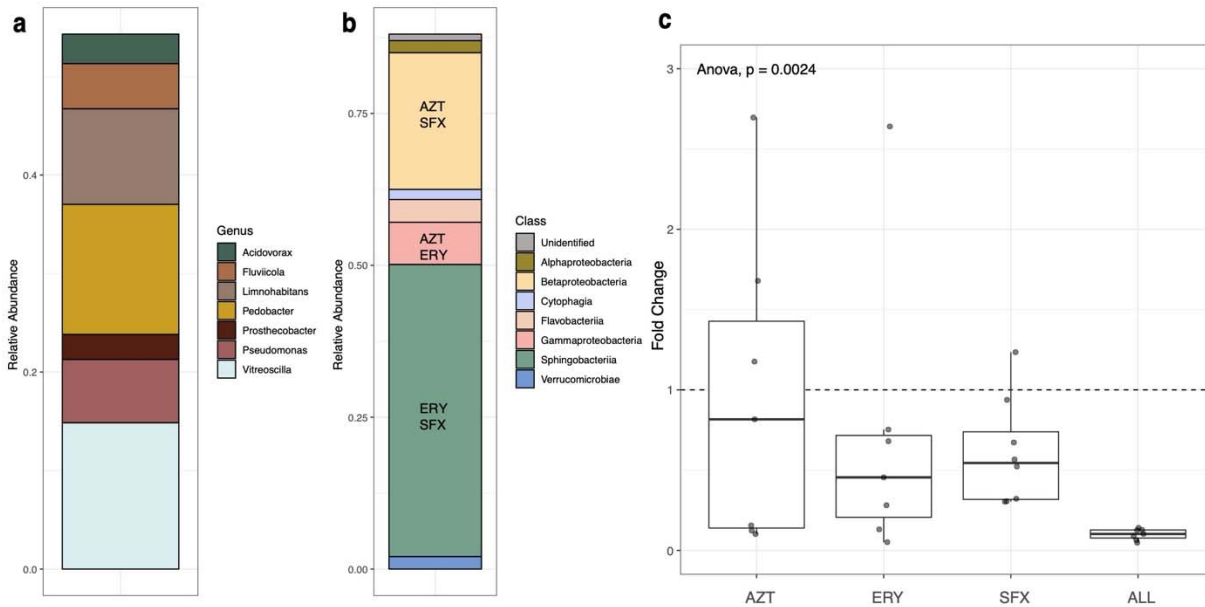
111

## 112 **Results**

113 The microbiota of antibiotic-free *Daphnia magna* was dominated by few taxa, with only 10  
114 unique ASVs comprising approximately 60% of total abundance (**Figure 1a**). Primarily, the  
115 most abundant ASVs belonged to the *Limnohabitans*, *Pedobacter*, and *Vitreoscilla* genera. In

116 total, the *Daphnia magna* microbiota was relatively simple, with only 8 bacterial Classes  
117 identified (**Figure 1b**). Of these, Sphingobacteriia and  $\beta$ -proteobacteria were most common  
118 (48% and 26%, respectively).

119



120

121 Figure 1: 16S rRNA sequencing results from adult *Daphnia magna* in control conditions. (a)

122 Identified genera of the most abundant 10 ASVs in the *Daphnia magna* microbiota and their

123 relative abundances. (b) Relative abundances of bacterial Classes identified in the *Daphnia*

124 *magna* microbiota. Antibiotics targeting the three bacterial Classes are identified within the

125 columns (AZT = aztreonam, ERY = erythromycin, SFX = sulfamethoxazole). (c) Boxplot of

126 bacterial relative abundance fold change across antibiotic treatments as compared to the control

127 treatment, measured by amplification of the 16S rRNA V4 hypervariable region using qPCR.

128

129 Relative abundance of bacteria in *Daphnia magna* was significantly impacted by antibiotic

130 treatment ( $F_{4,32} = 5.197$ ,  $p = 0.0024$ , **Figure 1c**). Mean fold change in *Daphnia* treated with the

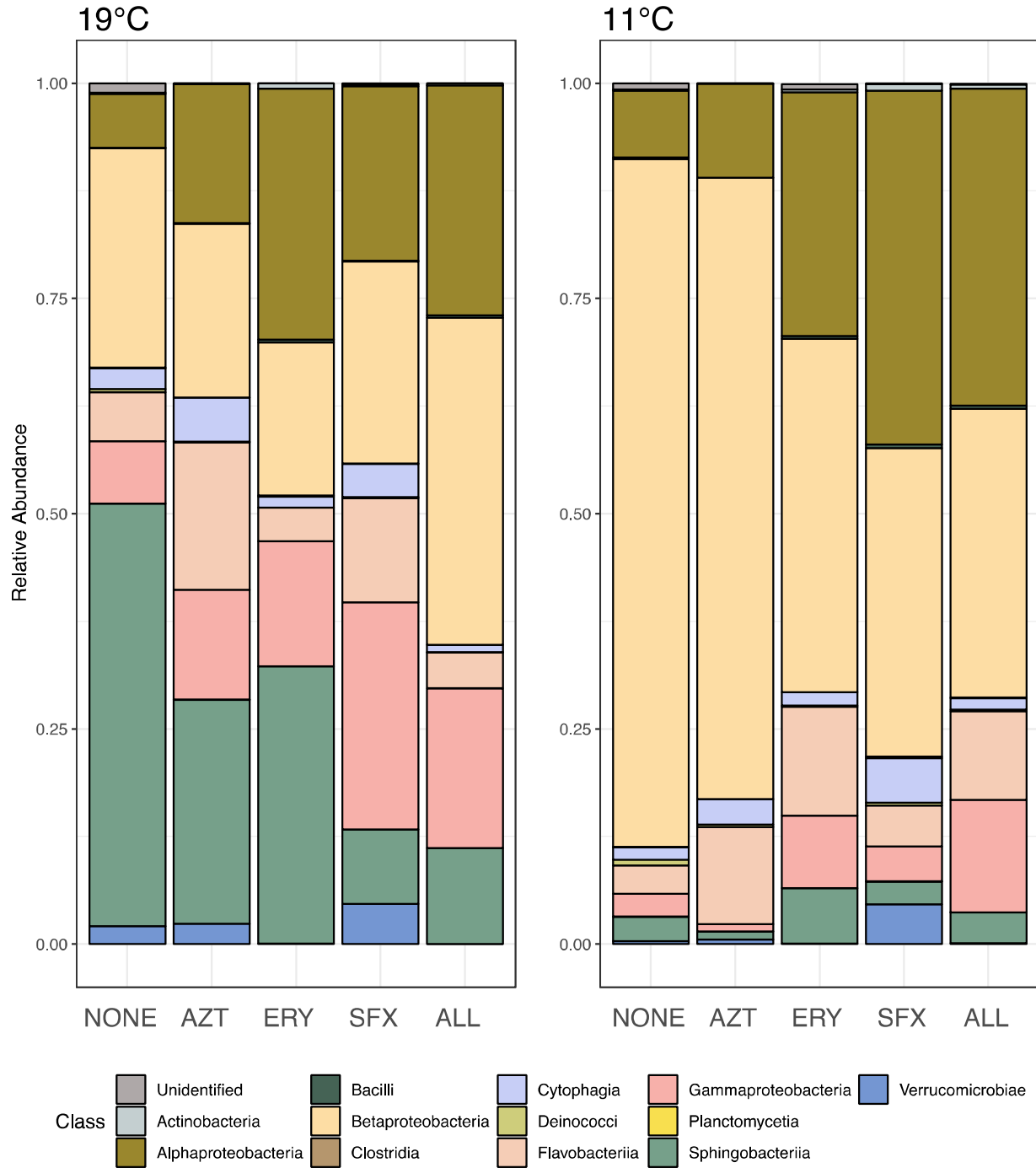
131 antibiotic trio was 0.1-fold that of the control treatment, while treatment with each of the three

132 antibiotics individually slightly reduced relative abundance (AZT 0.96 fold change; ERY 0.71  
133 fold change, SFX 0.61 fold change). The microbiota was significantly impacted by antibiotic  
134 treatments at the class rank (PERMANOVA, pseudo- $F_{4,41} = 3.977$ ,  $R^2 = 0.26$ ,  $p = 0.01$ , **Figure**  
135 **2**). Pairwise PERMANOVA comparisons of antibiotic treatments to the no antibiotic control  
136 indicated significant impacts of erythromycin (which suppresses  $\gamma$ -proteobacteria and  
137 Sphingobacteriia;  $p = 0.01$ ), sulfamethoxazole (which suppresses  $\beta$ -proteobacteria and  
138 Sphingobacteriia;  $p = 0.001$ ), and the antibiotic trio ( $p = 0.001$ ) on microbiota composition  
139 (**Supplementary Table 1**). We found multiple ASVs that were differentially expressed in each  
140 antibiotic treatment (**Figure 3, Supplementary Table 2**). Though aztreonam-treated samples did  
141 not have significantly different overall composition than the control, there were differences in the  
142 relative abundance of 8 ASVs, including decreases in *Pseudomonas* ( $2^{-26.5}$ , or  $10^{-8}$  fewer  
143 *Pseudomonas* than in the no antibiotic control) and Sphingomonas ( $2^{-24.3}$ ), and increases in  
144 *Microvirga* ( $2^{7.04}$ ). Erythromycin had 16 differentially abundant ASVs, with an unidentified  
145 Sphingobacteriia genus and *Emticicia* experiencing the greatest fold abundance changes ( $2^{20.2}$   
146 and  $2^{-9.43}$ , respectively). Treatment with sulfamethoxazole induced changes for 8 ASVs,  
147 primarily increasing the abundance of an unidentified Sphingobacteriia ( $2^{27.4}$ ) and decreasing the  
148 abundance of *Pedobacter* ( $2^{-9.19}$ ). Treatment with the antibiotic trio impacted the abundances of  
149 19 ASVs, contributing to fold increases of an unidentified Sphingobacteriia ( $2^{19.24}$ ) and  
150 decreasing the abundance of *Caulobacter* ( $2^{-20.18}$ ). In summary, aztreonam, meant to target  $\beta$ -  
151 proteobacteria and  $\gamma$ -proteobacteria, reduced the relative abundances of  $\gamma$ -proteobacteria and  
152 some  $\alpha$ -proteobacteria; erythromycin, meant to target  $\gamma$ -proteobacteria and Sphingobacteriia,  
153 increased the relative abundance of  $\alpha$ -proteobacteria while decreasing the relative abundances of  
154 multiple other Classes including  $\gamma$ -proteobacteria and Sphingobacteriia; sulfamethoxazole, meant

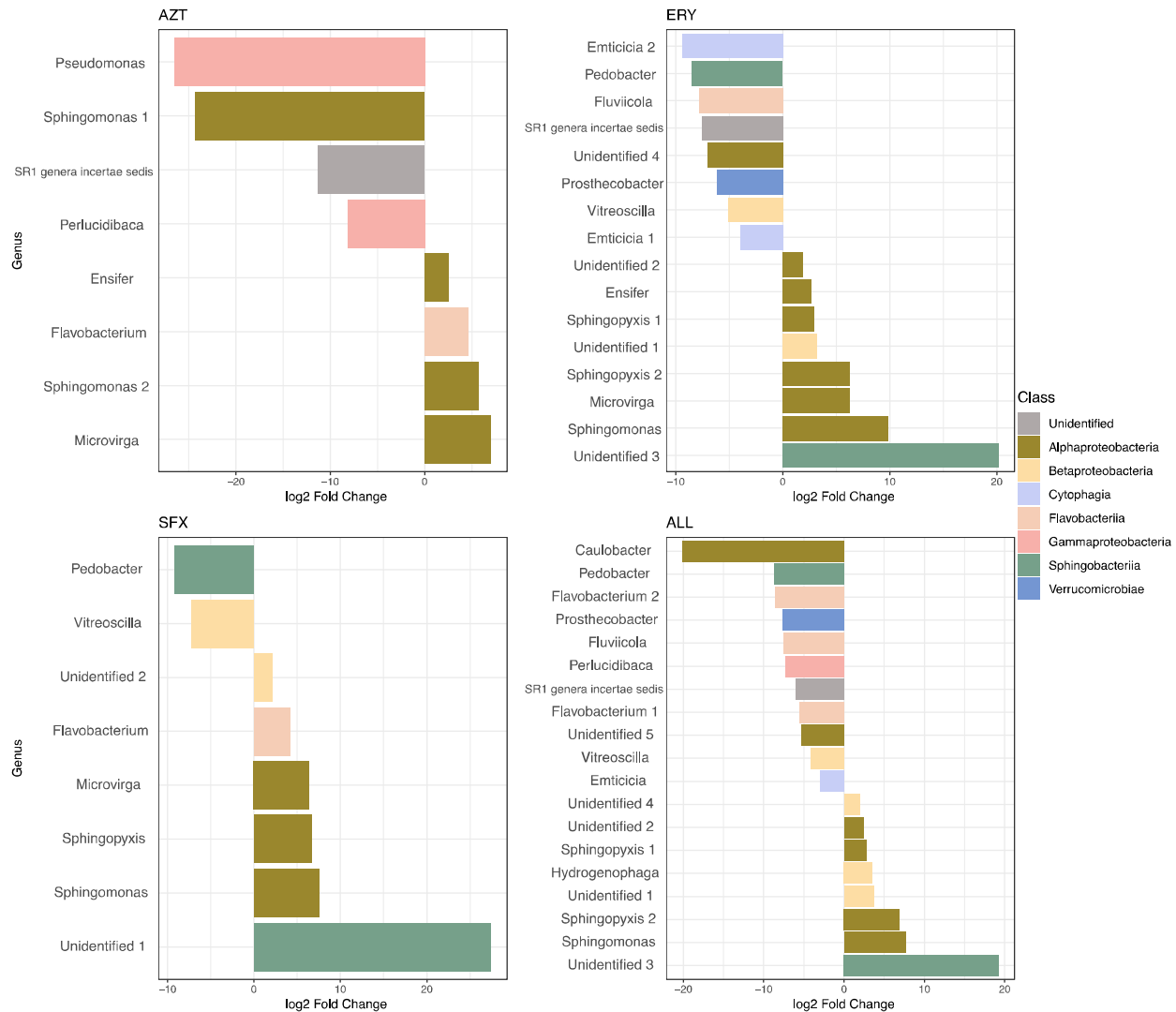
155 to target  $\beta$ -proteobacteria and Sphingobacteriia, decreased the relative abundance of  $\beta$ -  
156 proteobacteria and increased  $\alpha$ -proteobacteria. The antibiotic trio had the most wide-ranging  
157 effects on the microbiota, increasing the relative abundance of some Sphingobacteriia but  
158 primarily decreasing relative abundances across multiple Classes.

159





161 Figure 2: Microbiota composition at the Class level in *Daphnia magna* across antibiotic  
 162 treatments and across temperature treatments (NONE = no antibiotics, AZT = aztreonam, ERY =  
 163 erythromycin, SFX = sulfamethoxazole, ALL = AZT, ERY, and SFX). Taxa are conglomerated  
 164 at the Class rank to show differences in relative abundance of taxa among antibiotic treatments.  
 165



166  
 167 Figure 3: Differentially abundant ASVs in each antibiotic treatment as compared to the no  
 168 antibiotic control. Each bar represents a single ASV identified to the genus level, with genus

169 name indicated on the left. Bar color indicates the bacterial Class of each ASV, and bar length  
170 indicates the fold change in abundance of each ASV.

171

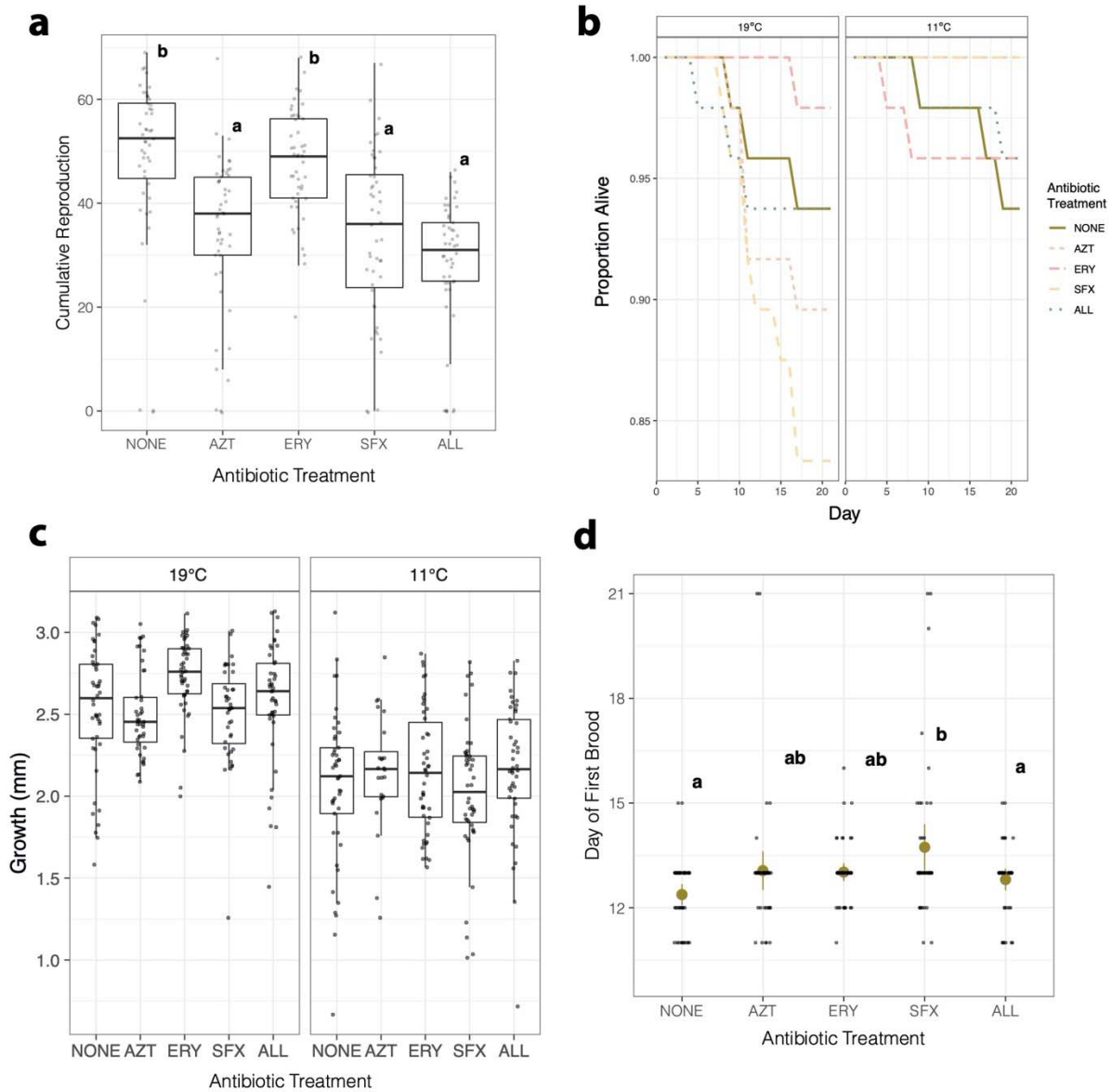
172 The *Daphnia magna* microbiota was significantly impacted by temperature. Composition was  
173 significantly shifted at the Class taxonomic rank across temperatures (pseudo- $F_{1,41} = 7.178$ ,  $R^2 =$   
174  $0.12$ ,  $p = 0.01$ ) and across the interaction between antibiotic treatment and temperature (pseudo-  
175  $F_{4,41} = 1.623$ ,  $R^2 = 0.11$ ,  $p = 0.02$ , **Figure 2**). In all antibiotic treatments except aztreonam, there  
176 were significant differences in microbiota composition between temperatures when compared to  
177 the control (Pairwise PERMANOVAs, erythromycin pseudo- $F_{1,9} = 4.001$ ,  $R^2 = 0.31$ ,  $p = 0.006$ ;  
178 sulfamethoxazole pseudo- $F_{1,9} = 2.605$ ,  $R^2 = 0.30$ ,  $p = 0.045$ ; trio pseudo- $F_{1,6} = 5.142$ ,  $R^2 = 0.46$ ,  $p$   
179  $= 0.034$ ). Multiple taxa were found to be differentially abundant even in treatments not found to  
180 have significantly different community compositions overall (**Supplementary Figure 1**,  
181 **Supplementary Table 2**). The no antibiotic control had a single ASV (Sphingobacteriia genus,  
182  $2^{-13.68}$ ) that was differentially abundant in cold temperatures. In aztreonam, 6 ASVs were reduced  
183 in relative abundance in cold temperatures, with a Sphingobacteriia (unidentifiable beyond  
184 Class) and an  $\alpha$ -proteobacteria genus experiencing the greatest reductions ( $2^{-11.98}$  and  $2^{-11.4}$ ,  
185 respectively). In erythromycin, 12 ASVs were differentially abundant; of those, the same  
186 Sphingobacteriia as in the aztreonam treatment was reduced ( $2^{-27.54}$ ) and a *Pseudomonas* ASV  
187 was more abundant ( $2^{6.34}$ ). Sulfamethoxazole impacted 13 ASVs, including *Limnohabitans* ( $2^{6.61}$ )  
188 and the same Sphingobacteriia ( $2^{-25.84}$ ). The antibiotic trio treatment only had one  
189 differentially abundant ASV (*Nubsella*,  $2^{7.89}$ ). Generally, differentially abundant Classes were  
190 limited to the  $\alpha$ - and  $\beta$ -proteobacteria, Flavobacteriia, and Sphingobacteriia. Sphingobacteriia  
191 exhibited the greatest changes in abundance across antibiotic treatments, with a Sphingobacteriia

192 genus experiencing between  $2^{-5}$  and  $2^{-27.5}$  reduced abundance in no antibiotics, aztreonam,  
193 erythromycin, and sulfamethoxazole at low temperatures.

194

195 Host fitness was significantly impacted by antibiotics. In particular, cumulative host  
196 reproduction over the course of the experiment was reduced by antibiotics ( $F_{4,470} = 49.59$ ,  $p <$   
197  $0.001$ ). Post-hoc Tukey tests revealed that this reduction was most significant in the aztreonam,  
198 sulfamethoxazole, and the antibiotic trio treatments (all  $p < 0.001$ ) (**Figure 4a**). A complete list  
199 of post-hoc comparisons for cumulative reproduction can be found in **Supplementary Table 3**.  
200 Though cumulative host reproduction was reduced in these treatments and many amplicon  
201 sequence variants experienced shifts in relative abundance across treatments, there was no single  
202 bacterial genus that shifted in relative abundance predictably across treatments, suggesting that  
203 there is no genus that is uniquely important to *D. magna* reproductive fitness. Reproductive  
204 timing was also impacted ( $F_{4,222} = 4.797$ ,  $p < 0.001$ ), where *Daphnia magna* treated with  
205 sulfamethoxazole experienced a later age at first reproduction than those treated with other  
206 antibiotics (Tukey HSD,  $p = 0.03$ , **Figure 4d, Supplementary Table 4**). *Daphnia magna*  
207 exposed to antibiotics experienced a significant overall reduction in growth (ANOVA,  $F_{4,419} =$   
208  $2.08$ ,  $p = 0.004$ , **Supplementary Table 5**); the main contributor to this was a significant  
209 reduction in growth in sulfamethoxazole as compared to erythromycin (Tukey HSD,  $p = 0.003$ ,  
210 **Supplementary Table 5**). Exposure to any antibiotic had no impact on *Daphnia* survival  
211 (**Figure 4b, Supplementary Figure 2**). Host fitness was also impacted by temperature.  
212 Cumulative reproduction was reduced almost completely in the cooler 11°C treatment; no  
213 reproduction was observed in the control and only one adult individual reproduced across all of  
214 the antibiotic treatments (ANOVA,  $F_{1,470} = 1751.08$ ,  $p < 0.001$ ). There was also an effect of the

215 interaction between antibiotics and temperature (ANOVA,  $F_{4,470} = 19.84$ ,  $p < 0.001$ ), to the  
216 extent that the impacts of antibiotics on reproduction in low temperatures was not noticeable due  
217 to the strong effect of temperature on reproduction. *Daphnia* growth was limited in the cold  
218 temperature treatment ( $F_{1,419} = 176.01$ ,  $p < 0.001$ ) (**Figure 4c**). Temperature impacted survival,  
219 with *Daphnia* in colder temperatures surviving significantly more than those in control  
220 temperatures (HR = -1.087,  $p = 0.013$ ) (**Figure 4b, Supplementary Figure 2**).  
221



222

223 Figure 4: Summary of *Daphnia magna* fitness. (a) Boxplot of cumulative reproduction over the  
224 21-day experiment in *Daphnia magna* across antibiotic treatments (NONE = no antibiotics, AZT  
225 = aztreonam, ERY = erythromycin, SFX = sulfamethoxazole, ALL = all three antibiotics) in the  
226 19°C temperature treatment. Points show cumulative reproduction of each individual *Daphnia*  
227 over the 21-day experiment. Letters denote significant differences among treatments. (b)  
228 Survival curves of *Daphnia magna* in antibiotic treatments across temperature treatments. Line

229 color and pattern denote antibiotic treatment. (c) *Daphnia magna* growth in millimeters over the  
230 21-day experiment across antibiotic and temperature treatments for individuals who survived the  
231 entire time course. Boxplots denote median and first and third quartiles, and jittered points show  
232 growth of each individual within a treatment. (d) Time to reproductive maturity of *Daphnia*  
233 *magna* in the 19°C temperature treatment across antibiotic treatments. Jittered points denote  
234 individuals within each antibiotic treatment. Letters denote significant differences among  
235 treatments.

236

## 237 **Discussion**

238 In this study, we manipulated the microbiome of *Daphnia magna* using low doses of targeted  
239 antibiotics to examine the impacts of selective suppression on host fitness. We found that  
240 aztreonam and sulfamethoxazole, antibiotics targeting abundant bacterial Classes in the *Daphnia*  
241 *magna* microbiome, had the largest impacts on host fitness in both fecundity and growth.  
242 However, our results contrast with those of Sison-Mangus *et al.* (2015)<sup>18</sup> and Callens *et al.*  
243 (2016)<sup>14</sup> in that survival was not impacted by antibiotics here. This is likely due to our use of  
244 targeted antibiotics, rather than the broad-spectrum antibiotics used to completely suppress the  
245 microbiota in those studies. We found that in all of the antibiotic treatments, relative abundance  
246 of bacteria was lower than in the control treatment. However, relative abundance in treatments  
247 did not fully correspond with the relative abundances of the intended target groups, suggesting  
248 that non-target groups may have also been affected. Alternatively, the *Daphnia magna*  
249 microbiota may have a carrying capacity, where non-target groups may replace taxa negatively  
250 affected by the antibiotic perturbation. Our 16S rRNA sequencing shows that non-target Classes  
251 were impacted in unexpected ways, potentially due to microbial interactions or off-target

252 antibiotic effects. We also found that shifts in microbiome composition were dependent on  
253 environmental conditions, with *Daphnia magna* exposed to the same antibiotics at low  
254 temperatures not experiencing the same shifts in microbiome composition and differentially  
255 affecting host fitness.

256

257 Our characterization of the *Daphnia magna* microbiome in standard conditions (19°C , no  
258 antibiotics) yields a composition that is similar to that of healthy adult *Daphnia magna* in other  
259 studies, though this genotype has been isolated in laboratory culture for >3 years<sup>15,34,37</sup>. This  
260 suggests that our *Daphnia magna* maintained in culture have retained *Daphnia*-specific microbes  
261 present during initial field collection, as all culture media is autoclaved prior to use and algae is  
262 grown axenically. We did find that the microbiota of untreated adult *Daphnia magna* in these  
263 cultures exhibited a higher relative abundance of Sphingobacteriia than in other studies  
264 (approximately 48%, compared to 4%-20%), but other abundant bacterial Classes were similar,  
265 including  $\alpha$ -proteobacteria,  $\beta$ -proteobacteria, and  $\gamma$ -proteobacteria<sup>15,34</sup>. In particular, we found  
266 that species in the  $\beta$ -proteobacteria genus *Limnohabitans* are highly abundant in the *Daphnia*  
267 microbiome, consistent with prior work<sup>15,27,28,29,30</sup>. *Limnohabitans* has important effects on host  
268 fitness in *Daphnia magna*, with monoassociations with *Limnohabitans* strains increasing  
269 *Daphnia* fecundity<sup>17</sup>. Though species in the *Limnohabitans* genus are prevalent free-living  
270 freshwater bacterioplankton<sup>31</sup>, metagenome-assembled *Limnohabitans* genomes from the  
271 *Daphnia magna* microbiota have <90% average nucleotide identity shared with their closest  
272 sequenced free-living relatives<sup>32</sup>, indicating these *Limnohabitans* species are likely specific to  
273 *Daphnia magna*. These *Daphnia*-associated *Limnohabitans* species encode for biosynthesis and  
274 export of amino acids essential for *Daphnia magna*<sup>33,34</sup>, potentially providing these when hosts



275 are unable to acquire them from their diet. Other genera identified here and in other studies  
276 include *Pedobacter*, *Emticicia*, and *Acidovorax*<sup>35,15</sup>, among others.

277

278 Antibiotics suppressed the microbiota of *Daphnia magna* but were not as targeted as expected.

279 Treatments of aztreonam and sulfamethoxazole reduced the relative abundance of  $\beta$ -

280 proteobacteria as expected; however, erythromycin treatment unexpectedly reduced  $\beta$ -

281 proteobacteria relative abundance as well. Treatment with erythromycin also increased the

282 relative abundance of some Sphingobacteriia ASVs, though it reduced the abundance of

283 Bacteroidetes as expected. This suggests two possibilities: narrow-spectrum antibiotics may be

284 affecting different taxa than expected here, or suppression of the targeted taxa impacts non-

285 targeted taxa that may be interacting with those that were suppressed.

286

287 Host fecundity was impacted by treatment with aztreonam, sulfamethoxazole, and the antibiotic

288 trio, and host growth was reduced when treated with the trio as well. Sulfamethoxazole (targeting

289  $\beta$ -proteobacteria and Sphingobacteriia) also delayed age at first reproduction, supporting our

290 hypothesis that suppression of more abundant bacterial Classes would have larger impacts on

291 host fitness. While it is possible that the antibiotics used here have direct effects on host fitness<sup>30</sup>,

292 germ-free *Daphnia magna* treated with the same antibiotic trio had the same fitness metrics as

293 the *Daphnia* treated here (Cooper, unpublished data), suggesting that differences in host fitness

294 are mediated through the microbiome. The reduction in host fecundity without a consistent

295 associated decrease in abundance of any particular microbial taxon across treatments suggests

296 that multiple taxa are involved in the functions that benefit host fecundity, supported by potential

297 cross-species interactions identified through shotgun sequencing<sup>32</sup>. Specifically, metagenome

298 sequencing indicates that a *Pedobacter* species uniquely encodes for chitin degradation and sialic  
299 acid cleavage and that other species (primarily *Limnohabitans*) may be able to utilize those  
300 cleaved sialic acids for amino acid biosynthesis<sup>33</sup>. This delay in reproductive maturity and  
301 associated reduction in *Pedobacter* in the sulfamethoxazole treatment may indicate that microbe-  
302 microbe interactions are affected by targeted antibiotic treatment.

303

304 Interestingly,  $\gamma$ -proteobacteria are found in high relative abundances in the *Daphnia magna* gut  
305 and could play a role in nutrient acquisition or pathogen protection<sup>14,15</sup>, yet their suppression  
306 with erythromycin did not affect host fitness. We hypothesize that this may be due to functional  
307 redundancy of taxa found in the *Daphnia* gut, as *Daphnia magna*'s indiscriminate filter feeding  
308 exposes gut microbes to a wide array of nutrients<sup>14,36</sup>. Indeed, the abundances and identities of  
309 taxa in the *Daphnia* gut vary substantially across studies<sup>23,18,16,30</sup>; this variation may allow  
310 different taxa not targeted by erythromycin to retain the necessary functions for nutrient  
311 acquisition.

312

313 Temperature dramatically shifted the microbiome and the fitness of *Daphnia magna*. These  
314 changes have been documented across host genotypes and warmer temperatures<sup>23,12</sup>, but to our  
315 knowledge this is the first to examine the effects of this cold of a temperature (11°C) on the  
316 microbiota of this keystone species. *Daphnia magna* raised in cold temperatures survived more,  
317 grew less, and had almost no offspring, a well-studied physiological mechanism for actively  
318 surviving winter in aquatic ecosystems<sup>37</sup>. Correspondingly, the microbiome shifts during this  
319 time. In 11°C,  $\beta$ -proteobacteria became even more abundant, comprising >80% of relative  
320 abundance. Sphingobacteriia, Flavobacteriia, and  $\gamma$ -proteobacteria were reduced to <5% relative

321 abundance each. This may be due to cold-induced changes to host metabolic processes like fat  
322 storage and processing, which have been shown to shape microbiota composition<sup>38</sup>. *Daphnia*  
323 *magna* reduce stearic acid formation at low temperatures but increase monounsaturated fatty acid  
324 formation<sup>39</sup>, which could select for taxa able to utilize these types of fatty acids. Alternatively,  $\beta$ -  
325 proteobacteria may be so important for host fecundity<sup>17</sup> that they must remain in high abundance  
326 to ensure they remain for the post-winter reproductive cycle. Though microbiota composition  
327 shifted in cold temperatures, it is unlikely that host fitness is mediated by microbiota change.  
328 Even with a greater relative abundance of reproductive fitness-promoting  $\beta$ -proteobacteria,  
329 *Daphnia magna* in cold temperatures had significantly reduced reproductive fitness, suggesting  
330 that temperature directly impacts fitness.

331

332 Antibiotics did not affect microbiome composition in cold temperatures as they did in standard  
333 conditions. Treatment with aztreonam at 11°C resulted in a microbiome composition nearly  
334 identical to that of *Daphnia magna* not treated with antibiotics in 11°C. Though erythromycin  
335 does not target  $\beta$ -proteobacteria, this Class was reduced in the cold temperature-erythromycin  
336 treatment, suggesting that taxa in the *Daphnia* microbiota may be differentially susceptible to  
337 antibiotics depending on environmental factors and based on host physiological responses to the  
338 environment. Horizontal gene transfer could play a role in this differential response, as species in  
339 the *Daphnia magna* microbiome do encode for antibiotic resistance and efflux<sup>33</sup> and this  
340 experiment was conducted over a time period long enough to allow antibiotic resistance to  
341 establish within species<sup>40</sup>. Furthermore, cold temperatures have been shown to increase the  
342 abundance of antibiotic resistance genes<sup>41</sup>.

343

344 Our results suggest that more abundant Classes in the microbiome (in this case, Sphingobacteriia  
345 and  $\beta$ -proteobacteria) have larger impacts on host fitness than rarer taxa. Within-host microbial  
346 communities generally have a skewed abundance pattern, where a few species constitute the  
347 majority of total abundance but many species are found in low abundances. Some work indicates  
348 that abundant taxa contribute to host fitness<sup>42</sup>, while others indicate that rare, keystone taxa have  
349 disproportionate impacts on host fitness<sup>43</sup>. However, a general relationship between abundance  
350 in the microbiota and benefit to host fitness is hard to untangle in complex systems, and much  
351 research in model systems focuses on hosts with single microbial taxa that have significant  
352 impacts on host fitness<sup>44</sup>. Utilizing animal models like *Daphnia magna* with more than one taxon  
353 contributing to host fitness but a relatively simple overall microbial community allows for a  
354 greater understanding of the interplay between the microbiota and host. In *Daphnia magna*, more  
355 abundant taxa (e.g., *Limnohabitans*) confer greater benefits to host fitness, primarily through  
356 functions that contribute to increases in host fecundity and growth, whereas the loss of rare  
357 species had little effect on fitness. At the same time, abiotic conditions can have a much larger  
358 effect on host fitness than the microbiome, as demonstrated by the changes in *Daphnia* fitness  
359 with temperature not directly mirrored by changes in the microbiota. Our results show that  
360 multiple members of the *Daphnia magna* microbiota impact host fitness in different ways, and  
361 that these impacts must be understood in the broader context of external factors known to  
362 directly affect host fitness.

363

## 364 **Methods**

365 *Daphnia*. This experiment was conducted using *Daphnia magna* clone 8A, taken from Kaimes  
366 Farm, Leitholm, Scottish Borders<sup>45</sup>. Stock cultures of *D. magna* clone 8A were maintained in

367 19°C controlled chambers with a 16h light, 8h dark light cycle in 400 mL jars with phosphorous-  
368 and nitrogen-depleted COMBO medium<sup>46</sup> for multiple generations. Cultures were fed a  
369 standardized 0.25 mg C/mL/day using green algae *Chlamydomonas reinhardtii* (CPCC 243). *C.*  
370 *reinhardtii* was cultured in COMBO medium. The volume necessary to provide *D. magna* with  
371 adequate carbon was calculated using the Biotek Epoch Microplate Spectrophotometer.

372

373 *Experimental design.* Prior to the experiment, 72 *D. magna* were moved to 35 mL glass vials  
374 with COMBO medium and allowed to mature under controlled conditions. Neonates from the  
375 third brood of each adult were pooled within 24h of birth and randomly assigned to experimental  
376 treatments (n=48 per experimental treatment). The experimental treatments consisted of five  
377 antibiotic treatments crossed with two temperature treatments, 19°C and 11°C. Antibiotic  
378 treatments were as follows: a control treatment with no antibiotics; 500 ug/L aztreonam; 400  
379 ug/L erythromycin; 250 ug/L sulfamethoxazole; and an antibiotic trio consisting of all three  
380 antibiotics together at the concentrations listed above. These antibiotic concentrations were  
381 chosen on the basis of a pilot experiment showing no short-term toxicity effects on *D. magna*  
382 survival but a significant reduction in bacterial abundance (revealed using qPCR with universal  
383 bacterial 16S primers). Body size of each *D. magna* was measured from eyespot to beginning of  
384 apical spine before placement into the experimental vials. Experimental *D. magna* were raised in  
385 35 mL glass vials with COMBO for 21 days. Each vial was checked for survival and fed daily  
386 with 0.25 mg C/ml/day of the diet treatment. Vials were also checked daily for offspring, which  
387 were counted and removed if present. At the conclusion of day 21 or upon death, *D. magna* were  
388 collected from the treatments. Body size was again measured from eyespot to beginning of apical

389 spine to determine growth. *D. magna* from each treatment were pooled in sets of 10 in 1.5 mL  
390 microcentrifuge tubes for DNA extraction and processing (n=4 per treatment).  
391  
392 *DNA extraction, library preparation, and sequencing.* DNA was extracted from all pooled  
393 samples using the Qiagen DNEasy Blood & Tissue Kit using the manufacturer's spin-column  
394 protocol of total DNA from animal tissues (Qiagen, Hilden, Germany). Whole *D. magna* were  
395 digested with Proteinase K for 24h to ensure cells within the carapace were lysed but the  
396 carapace was not<sup>47</sup>. Following extraction, PCR amplification of the V4 region of the 16S rRNA  
397 gene was performed using the 515f (5'-GTGCCAGCMGCCGCGGTAA-3') and 806r (5'-  
398 GGACTACHVGGGTWTCTAAT-3') universal 16S primer pair<sup>48</sup>. Amplification consisted of  
399 denaturation at 95°C for 3min, followed by 35 cycles of 95°C for 45sec, 58°C for 30sec, and  
400 72°C for 45sec, and finished with a 72°C for 5min extension step. Simultaneously, a subset of  
401 samples from each of the antibiotic treatments was prepared for qPCR using the FastStart SYBR  
402 Green Master Mix to verify antibiotic treatments were reducing overall bacterial abundance.  
403 Each sample was run in triplicate to ensure amplification was achieved in each sample. All  
404 samples were checked for successful amplification using a 1% agarose electrophoresis gel.  
405 Samples were then normalized with the SequalPrep Normalization Plate Kit. Prior to sample  
406 pooling, sample quality was checked using the Agilent High Sensitivity DNA Kit on the Agilent  
407 TapeStation and via qPCR with the KAPA Library Quantification Kit. Samples were then pooled  
408 and spiked with PhiX DNA. The pooled libraries were then sequenced using the Illumina MiSeq  
409 Reagent Kit v2 (300-cycles) on an Illumina MiSeq. Sequencing was carried out at the Nebraska  
410 Food for Health Center (Lincoln, Nebraska, USA).  
411

412 *Sequencing data processing.* Following sequencing, reads were demultiplexed using Illumina's  
413 built-in MiSeq Reporter software. All reads were then analyzed using DADA2<sup>49</sup> in R. In  
414 DADA2, our pipeline consisted of low-quality (<Q30) read trimming, estimation of read error,  
415 dereplication of reads within samples, and chimera removal. Remaining reads were considered  
416 amplicon sequence variants (ASVs), then were assigned taxonomy to the genus level using the  
417 Refseq-RDP database<sup>50</sup>. All visualization of ASVs was performed with Phyloseq<sup>51</sup> in R, where  
418 reads without a taxonomic assignment at the phylum level and those assigned to "Chloroplast"  
419 were removed for visual clarity. All scripts for read processing and visualization are available on  
420 GitHub (<https://github.com/reillyowencooper/ab-targeting-daphnia>).

421

422 *Statistical analysis.* All statistical tests were performed in R. Host *D. magna* life history traits  
423 measured as indicators of fitness outcomes included growth, survival, and reproduction. Growth  
424 was quantified as the difference between size measurements at the beginning and end of the  
425 experiment. Differences in growth among treatments, including the interactions between  
426 antibiotics and temperature, were analyzed using an ANOVA. We also used ANOVAs to test for  
427 effects of antibiotics and temperature on reproduction, which was measured as number of  
428 juveniles per brood and day of first reproductive event (production of the first brood for each  
429 individual). Tukey's HSD post-hoc tests were conducted to determine which treatments  
430 significantly differed from the control treatment. Survival rates among treatments were analyzed  
431 using the Cox proportional hazards model. Individuals alive at experiment conclusion were  
432 coded as censored. We used the  $\Delta\Delta C_t$  method to calculate log fold change in abundance of the  
433 16S rRNA gene among antibiotic treatments, normalizing against the *Daphnia magna* actin gene  
434 (forward primer: 5'-CCACACTGTCCCCATTTATGAA-3', reverse primer: 5'-

435 CGCGACCAGCCAAATCC-3') and against the control treatment. A PERMANOVA was  
436 conducted among treatments on the calculated unweighted UniFrac distances to test the effects  
437 of antibiotics and temperature on microbiota composition, then pairwise comparisons of  
438 antibiotic treatments in each temperature were conducted to find treatments with significantly  
439 different overall community composition. DESeq2 was used to find differentially abundant taxa  
440 among treatments.

441

442

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447

#### 448 **Author Contributions**

449 ROC and CEC designed the study. ROC and JMV performed the experiments. ROC and CEC  
450 wrote the manuscript, and all authors approved the manuscript in its final form.

451

452

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