

1 **Evolutionary History Impacts Phyllosphere Community Assembly on Forage Grasses**

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16

17 **ABSTRACT**

18 Benefits leaf bacterial communities provide to plant hosts are reduced by external stress.

19 Understanding how plant hosts impact phyllosphere community assembly, how microbes

20 influence plant traits, and how this interaction changes under stress will advance our insight into

21 the evolutionary relationship between plants and their microbial communities. We investigated

22 phyllosphere community assembly change over time, between host species, and under drought

23 stress on three native temperate grasses and three non-native tropical grasses. By growing them

24 together, effects of host geography and differences in environmental variables were eliminated
25 allowing us to test evolutionary history on community assembly. We found evidence of
26 phyllosymbiosis which increased significantly under drought stress, indicating phyllosphere
27 communities and their response to stress relate to grass species phylogeny. We also show native
28 temperate grasses displayed stronger cophylogenetic relationships between grass hosts and their
29 microbial communities and had increased selection by host species over time compared to non-
30 native tropical hosts. Interestingly, the functional marker gene *nifH*, though differentially present
31 on all host species was not susceptible to drought. The evidence of shared evolutionary history,
32 presence of functionally important bacteria, and responses to drought suggest that microbial
33 communities are important plant traits that coevolve alongside their plant hosts.

34

35

36 INTRODUCTION

37 As one of the largest terrestrial habitats, grasslands make up nearly 70% of global agricultural
38 land and contribute important ecosystem services including impacting water quality, erosion
39 prevention, and climate regulation through carbon sequestration and greenhouse gas mitigation
40 [1]. The important agricultural and ecological functions grasslands provide are threatened due to
41 projected decreases in water availability as drought frequency and severity continue to increase
42 [2–4]. This will have drastic effects on grassland productivity, ultimately reducing global food
43 security and increasing climate change [5].

44

45 Grass leaf surfaces harbor diverse microbial communities, termed the phyllosphere, which
46 provide important functions to their host including disease prevention, stress tolerance,

47 ecosystem productivity, and nutrient cycling through processes such as nitrogen fixation [6–9].
48 In return, plants provide nutrients to the bacteria creating a symbiotic relationship, but what
49 drives these relationships is not completely understood. Previous studies show that while plant
50 host identity plays an important role in microbial community assembly, phyllosphere
51 communities are broadly dominated by similar taxa including *Proteobacteria*, *Bacteroidetes*, and
52 *Actinobacteria* [10–13].
53
54 Common theories to explain phyllosphere community assembly include the existence of a
55 functional core community, in which phyllosphere community members provide consistent
56 functional traits across host species [11, 14], and the hologenome theory of evolution, which
57 postulates evolution occurs between hosts and microbes together [15]. Many core functions
58 support epiphytic bacterial growth under harsh conditions indicating microbial adaptation to the
59 phyllosphere. These include pigmentation and DNA repair systems to protect from UV radiation,
60 production of extracellular polysaccharides to promote biofilm formation which protects against
61 osmotic stress, and motility-related proteins for movement towards nutrients [8, 11, 16, 17].
62 Additional functional traits are important for plant health and ecosystem functioning, by
63 promoting global carbon and nitrogen cycles, photosynthetic strategies, resource acquisition, and
64 plant defense [11, 14, 18]. Nitrogen fixation by bacteria called diazotrophs, frequently associated
65 with the rhizosphere, occurs in the phyllosphere contributing to total nitrogen input in an
66 ecosystem [9, 11, 19]. The observed differences in relative abundance of functional genes and
67 taxonomic identity despite low variability between host species [13, 14], suggest the functional
68 core exists within the hologenome theory. For example, phyllosphere bacteria can have
69 rhodopsins which provide energy and protection from UV damage. These pigments absorb

70 different wavelengths of light than their plant host allowing for optimal utilization of resources
71 thus indicating shared evolutionary history [20–22].
72
73 How functional profiles and plant-microbe relationships change under stress conditions is still
74 unknown. Therefore, we do not understand if response to stress is a stochastic process dependent
75 largely on atmospheric conditions, a response to changes in plant physiology, or a response
76 characterizing joint plant-microbe interactions. One method to explore host-microbe
77 relationships is phylosymbiosis, which determines if significant associations between microbial
78 communities and the phylogeny of their host species exist [23–25]. Phylosymbiosis can be
79 determined using a Mantel test to compare a host phylogenetic distance matrix to a microbial
80 community distance matrix. When phylosymbiosis occurs, phylogenetically related host species
81 have more similar microbial communities than less phylogenetically related hosts.
82 Phylosymbiosis can result from coevolution, which occurs when plant-microbe systems act as
83 reciprocal selective forces on each other [25, 26]. However, it can also result from differences in
84 host geography, host traits, or codiversification, which occurs when hosts and microbes exhibit
85 parallel divergence during continued associations [27]. A second method used to understand how
86 host phylogeny relates to microbial communities is cophylogeny, which tests the concordance of
87 the host phylogeny with the phylogeny of the associated microbial community [28, 29].
88 Cophylogenetic occurrences indicate shared evolutionary history between hosts and microbial
89 groups [29, 30]. While cophylogeny can result from processes such as biogeographical distance,
90 presence of cophylogeny is consistent with host-microbe coevolution [29, 31, 32]. Previous work
91 suggests that cophylogenetic associations are more likely to exhibit microbe-to-host interactions

92 [14, 33, 34]. Therefore, identifying these associations can help identify evolutionarily important
93 and ecologically active plant-microbe relationships.

94

95 To understand plant-microbe interactions we need to understand rules of assembly and functional
96 processes. Our objective was to investigate if phyllosphere communities are an adapted plant
97 trait. To address this objective, we explored the questions: (i) How does host phylogeny
98 influence microbial community assembly? (ii) How does host identity or phylogeny influence
99 microbial community response to drought stress? (iii) How is diazotroph abundance related to
100 microbial community structure and response to stress? To answer these questions, we
101 investigated how microbial community assembly changed over time, between host species, and
102 under drought stress. We chose three species of grasses commonly used in temperate forage
103 systems and three species commonly used in tropical forage systems. By growing all species in
104 the same common garden experiment, we eliminated effects of host geography and differences in
105 environmental variables on community assembly. Additionally, by growing native temperate and
106 foreign tropical species, we tested the influence of evolutionary history on community assembly.
107 Comparing the evolutionary history of phyllosphere communities to that of their hosts and
108 determining how communities change under drought stress, allowed us to understand if
109 phyllosphere microbes are a plant trait and begin to understand how to leverage microbes to
110 promote plant growth and stress tolerance.

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115 MATERIALS AND METHODS

116 *Study system*

117 Seeds for three non-native tropical grasses, *Brachiaria brizantha* (CIAT 26564), *Brachiaria*
118 *decumbens* (CIAT 6370), and *Brachiaria* hybrid (CIAT 1794), were acquired from CIAT (Cali,
119 Columbia). Native temperate grass species seeds, *Festuca arundinacea* (endophyte free Tall
120 Fescue), *Dactylis glomerata* (Orchardgrass), and *Lolium perenne* (Ryegrass), were acquired from
121 Albert Lea Seed Company (Albert Lea, MN, USA). Seeds were germinated in Pro-mix
122 commercial potting medium (Quakertown, PA, USA) in 2018 and grown in the College of
123 Natural Sciences Research and Education Greenhouse at the University of Massachusetts-
124 Amherst. In June 2019, individual plants were transplanted into 15x30cm pots filled with soil
125 collected from natural grass fields in Amherst, MA (Supplementary Methods, Supplementary
126 Table 1). Pots were moved outside, organized in a randomized block design, and allowed to re-
127 establish. Ten plant replicates of each temperate species were divided between ‘control’ and
128 ‘drought’ treatments. Drought treatment plants were placed under a 10 ft high rain shelter made
129 of greenhouse plastic allowing maximal airflow and high UV light penetration (Supplementary
130 Figure 1). Drought conditions were imposed over 38 days (21 AUG - 27 SEPT 2019). Plants in
131 the control group were given supplemental water to maintain soil moisture above 80% field
132 capacity. Plants in the drought group were given supplemental water when necessary to maintain
133 an even dry-down rate, determined from soil-moisture readings measured twice weekly using a
134 MiniTrase TDR with Buriable probe (Soilmoisture Equipment Corp., Goleta, CA, USA).

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136

137

138 *Plant Health Measurements*

139 Plant health measurements were taken on days 1, 19, 26, 33, and 38 to understand the effect of
140 drought on the plant host. Plant measurements taken were leaf relative water content (RWC),
141 chlorophyll concentration, and leaf cellular membrane stability determined by measuring
142 electrolyte leakage [35–37]. At the end of the drought period, above ground biomass was
143 measured by dividing plant material into five categories: stems, flowers, dead, mature, and
144 young leaves. After determining fresh mass, samples were dried in an incubator at 70°C for 5
145 days and dry mass was measured.

146

147 *Bacteria Community Sampling*

148 At each timestep, bacterial community DNA was extracted using the Nucleospin Plant II
149 Extraction Kit (Machery-Nagel, Düren, Germany) following a modified protocol. Five whole
150 ryegrass leaves or three whole leaves of each other species were aseptically removed from the
151 plant host and placed into a 15 ml conical tube with 1.5 ml of NucleoSpin Type-B beads and 4X
152 volume of Buffer PL1. Tubes were vortexed horizontally for 5 min at room temperature. The
153 lysate was incubated for 60 min at 65°C, placed in a NucleoSpin Filter tube, and centrifuged for
154 2 min at 11,000xg. The filtrate was added to 4X Buffer PC and extraction continued following
155 the manual. Aydogan et al. found that vortexing whole leaf samples in tubes with lysis buffer and
156 beads extracted important community members from biofilms with minimal plant DNA co-
157 extraction [12]. Extracted DNA samples underwent a two-step PCR amplification to attach
158 Illumina adaptor sequences and barcodes (Supplementary Methods). The first PCR step used
159 chloroplast excluding primers 799F and 1115R targeting the V5-V6 region of the 16S rRNA
160 gene [10] with linker sequences to attach Access Array Barcodes (Fluidigm, San Francisco, CA,

161 USA) [38]. Amplicons were pooled and sequenced on Illumina MiSeq Platform, with 251 bp
162 paired-end sequencing chemistry at the Genomics Resource Laboratory (University of
163 Massachusetts-Amherst). The abundance of nitrogen-fixing bacteria was determined using qPCR
164 quantifying the *nifH* gene using the PolF and PolR primers [39].

165

166 *Sequence Analysis*

167 Using the QIIME2 [40] pipeline, paired-end reads were demultiplexed, merged, trimmed to 315
168 base pairs, and binned inferring amplicon sequence variants (ASVs). Taxonomic identities were
169 assigned using the naïve Bayes sklearn classifier trained with the 799F/1115R region of the
170 Greengenes 13_8 database.

171 The data contained 9,207 ASVs from 280 samples containing a total of 15,218,029 reads.

172 Samples were rarefied to 4,000 reads, resulting in a loss of 16 samples. Alpha diversity was
173 calculated using Shannon Diversity Index and beta diversity using Weighted UniFrac and Bray-
174 Curtis distance metrics.

175

176 *Machine learning*

177 We used the mikropml R package to conduct machine learning (ML) analyses [41–43]. For each
178 model, we used random forest classification with 75% of the test data used to train the model and
179 the remaining 25% to test the model. ML was used on data collected the last day of drought to
180 predict if: (1) communities are from control or drought treated plant hosts regardless of host
181 species, and (2) bacterial communities came from tropical or temperate grass hosts regardless of
182 treatment. Model performance was evaluated using the area under the operating characteristic
183 curve (AUC) value. Models yielding AUC values above 0.6 were determined to have good

184 predictive power. Additionally, the mikropml pipeline enables determination of bacterial features
185 important for prediction and how much they contribute to AUC values.

186

187 *Phylosymbiosis and Cophylogeny*

188 Phylosymbiosis was determined using a Mantel test with matrices of grass species' phylogenetic
189 distances and microbial community beta diversities calculated using Bray-Curtis and weighted
190 UniFrac distances. Grass host phylogenetic distances were calculated using MEGAX [44].
191 Sequences of the chloroplast gene for each species were retrieved from NCBI [45] and aligned
192 using MUSCLE [46]. A phylogenetic tree was constructed using the maximum likelihood
193 method. A Mantel test was performed with the Spearman's Rank correlation with 9999
194 permutations using the Vegan package in R [47].

195

196 We tested for cophylogeny to understand if coevolution between microbial communities and
197 their plant host exists. Two separate global fit methods were employed: ParaFit as carried out in
198 the ape package [48] and PACo using the R package paco [30]. Microbial data used to test for
199 cophylogeny were filtered to only include data collected on the last sampling day with at least
200 100 reads across all samples resulting in 359 ASVs. Both methods were performed with host
201 phylogeny, microbial 16S rRNA phylogeny, and a presence/absence matrix for each host and
202 ASV. Additionally, both methods used the Cailleux correction method to account for negative
203 eigenvalues. PACo analysis was performed with 1000 permutations using the most conservative
204 quasiswap method, which is used when it is uncertain if the host is tracking symbiont evolution
205 or symbionts are tracking host evolution. ParaFit was performed using 999 permutations.
206 Significant associations were plotted using the cophyloplot function in the ape package.

207

208 *Statistical methods*

209 Separate generalized linear mixed models (GLMMs) were created to assess changes in alpha
210 diversity and *nifH* abundance using gamma distributions with a log link using the lme4 R
211 package [49]. Drought treatment, host species, and time were fixed effects and sample ID a
212 random effect to account for sampling over time. Effects of each variable was determined using
213 Tukey tests for comparison using lsmeans [50]. Effects of host species, drought treatment, time,
214 and their interactions on microbial community structure were determined using permutational
215 analysis of variance (PERMANOVA) and analysis of multivariate homogeneity of group
216 dispersions (PERMDISP2) with weighted UniFrac distances. Results were visualized using non-
217 metric multidimensional scaling (NMDS). PERMANOVA, PERMDISP2, and NMDS were
218 conducted using the vegan package and visualized using ggplot2 [51].

219

220

221 RESULTS

222 Phyllosphere communities varied between host species, over time, and as a result of drought.
223 Across all sample days and host species, *Alphaproteobacteria* was the dominant class in both
224 control (34.2%) and drought samples (34.6%), but community dynamics over time and as a
225 result of drought were different between host species (Figure 1A). At the start of the experiment,
226 *Alphaproteobacteria* and *Gammaproteobacteria* were the dominant groups, but by the end of the
227 experiment *Cytophagia* was the dominant class under control conditions. While *Cytophagia*
228 increased in relative abundance under drought conditions, *Alphaproteobacteria* remained the
229 dominant group on drought stressed hosts.

230

231 *Alphaproteobacteria* was dominated by *Sphingomonas* and *Methylobacterium* for each species,
232 but trends in relative abundance over time and as a result of drought were different between the
233 host species (Figure 1B). Genera from the class *Gammaproteobacteria* were more diverse and
234 variable between treatments, host species, and over time, but *Pseudomonas* was consistently
235 present across samples (Figure 1C). The increase in *Cytophagia* was accounted for almost
236 exclusively by the genus *Hymenobacter* (Figure 1D).

237

238 *Host species, time, and drought drive changes in community structure*

239 To evaluate the role plant species and drought had on microbial community diversity, we
240 modeled how alpha diversity changed over time, as a result of drought, and based on host species
241 (Supplementary Figure S2). Alpha diversity was not affected by drought treatment but was
242 significantly different based on host species identity.

243

244 Phyllosphere community structures were impacted by time, host species, and drought.

245 Additionally, the degree microbial communities changed as a result of drought related to known
246 drought tolerances of their host species. The strongest driver of phyllosphere community
247 structure was plant host species ($R^2=0.19$, $p=0.00$; PERMANOVA on weighted UniFrac
248 distances) (Table 1). Sample day ($R^2=0.14$, $p=0.001$) and watering condition ($R^2=0.02$, $p=0.001$)
249 were also significant drivers. All three two-way interactions were significant, with the strongest
250 interaction between sampling day and host species ($R^2=0.10$, $p=0.001$). However, the three-way
251 interaction was not significant ($R^2=0.04$, $p=0.0572$). PERMDISP2 was conducted to ensure
252 significant PERMANOVA results were caused by shifts in community structure instead of

253 differences in dispersion within treatments. PERMDISP2 analyses were not significant ($p=0.07$),
254 indicating that significant results from the PERMANOVA analyses are important factors for
255 community structure. Because of significant two-way interactions, we conducted individual
256 analyses on host species and sampling day to understand how microbial communities from each
257 host changed over time and were impacted by drought. Overall community response was first
258 detected 33 days into the experimental period. Additionally, host species effect on community
259 structure increased over time (Figure 2). Separate PERMANOVAs run on control samples from
260 Day 1 ($R^2=0.38$, $p=0.01$) and Day 38 ($R^2=0.57$, $p=0.001$) show increased effect of host species
261 on community assembly under non-stressed conditions. Additionally, influence of host species
262 within the temperate ($R^2=0.34$, $p=0.039$) and tropical ($R^2=0.31$, $p=0.094$) groups was similar at
263 the start of the experiment, but temperate species ($R^2=0.72$, $p=0.001$) explained greater
264 variability by the end of the experiment than the tropical species ($R^2=0.39$, $p=0.024$) (Table 2).

265 To further understand changes in community structure, average community distance over time
266 was modelled. On temperate grasses, average distance significantly decreased over time in both
267 control and drought treatments ($p_{\text{adj}} < 0.001$, TukeyHSD posthoc analyses). However,
268 community distance on tropical grasses remained stable over time. Additionally, significant
269 differences were not observed between temperate and tropical control groups but were observed
270 between drought samples ($p_{\text{adj}}=0.04$) (Figure 3).

271 We analyzed individual host species to understand how susceptible to drought each bacterial
272 community was based on host species. Ryegrass microbial communities were the first to show
273 changes between control and drought treatments; significant differences were first observed on
274 day 26 ($R^2=0.37$, $P=0.012$) (Supplementary Figure S3). *B. brizantha* ($R^2=0.47$, $P=0.031$), Tall
275 Fescue ($R^2=0.27$, $P=0.035$), and Orchardgrass ($R^2=0.25$, $P=0.033$) first showed significant

276 differences on day 33; *Brachiaria* hybrid ($R^2 = 0.30$, $P = 0.006$) on day 38; and *B. decumbens*
277 never displayed significant differences.

278 *Machine Learning Allows Accurate Prediction of Microbial Communities in Drought*

279 Machine learning (ML) allows detection of trends missed by traditional methods such as
280 PERMANOVA [43], and allows identification of features that enable its predicative power. We
281 used ML to test for a common response to drought among host species despite plant host
282 selection on microbial communities. ML revealed high predictive power in determining if
283 microbial communities were from the control or drought treatment (AUC=0.87) (Supplementary
284 Figure S4) and that the top eight ASVs contributed 0.07 to our AUC value (Figure 4,
285 Supplementary Table S2).

286

287 Additionally, ML had high predictive power in determining if communities were from temperate
288 or tropical hosts (AUC=0.89) at the end of the experiment regardless of drought treatment. The
289 model identified that 2 features, *Sphingomonas mali* and *Methylobacterium organophilum*,
290 contributed 0.107 to our AUC values, indicating their presence was important in model
291 performance (Supplementary Figure S5). Since tropical grasses are more related to each other
292 than to the temperate grasses, this analysis helped determine if community assembly is stochastic
293 or deterministic and identifies features associated with the two grass types.

294

295 *Grass Host Phylogeny Influences Phyllosphere Communities*

296 Host species impact on community assembly and response to drought was tested using
297 phyllosymbiosis, which occurs when significant association between host species phylogeny and
298 associated microbial communities occur [23]. Mantel tests on Bray-Curtis dissimilarities showed

299 more closely related host species had more similar microbial communities (Mantel $r=0.117$,
300 $p=0.0001$). Additionally, microbial communities were more related to host phylogeny during
301 drought stress (Mantel $r=0.202$, $p=0.007$) than under control conditions (Mantel $r=0.158$, $p=0.02$)
302 at the end of the drought period. Tests of phyllosymbiosis using Weighted UniFrac measures
303 showed similar trends but with weaker associations (All: Mantel $r=0.064$, $p=0.0002$; Drought:
304 Mantel $r=0.114$, $p=0.05$; Control: Mantel $r=0.057$, $p=0.19$). Because weighted UniFrac
305 incorporates phylogenetic information, it reduces nuanced variations at the tips of the bacterial
306 phylogenetic trees [25].

307

308 To further explore evolutionary relationships between host phylogeny and bacterial
309 communities, cophylogeny was tested with two separate global-fit methods. Global-fit methods
310 test congruence between host phylogenetic trees and the corresponding microbial phylogeny and
311 allow for identification of significant associations. PACo (Procrustes Approach to Cophylogeny)
312 uses Procrustes analyses to test the dependency of one phylogeny on the other [29, 30]. ParaFit
313 compares two distance matrices constructed from host and microbial phylogenetic distances and
314 tests for random associations between the groups [52]. Positive correlations can indicate host-
315 microbe coevolution [32, 53]. Tests for cophylogeny conducted on all samples collected on day
316 38 regardless of treatment using ParaFit (ParaFitGlobal=1.6024, $p=0.001$, permutations=999),
317 and PACo (PACo=0.999, $p=0.003$) revealed significant global-fit cophylogenetic relationships.

318

319 The influence of drought stress on cophylogenetic signal was determined to understand if
320 microbial response to drought was a stochastic process, and to look for evidence of a joint plant-
321 microbial response. Results using Parafit from both control (ParaFitGlobal=1.136, $p=0.001$) and

322 drought (ParaFitGlobal=1.296, $p=0.001$) showed evidence of cophylogeny. In the control
323 treatment there were 414 significant associations between bacteria and plant hosts and 340
324 significant associations in drought treatment samples. Tanglegrams displaying significant
325 associations between host and microbe phylogenies were created for control and drought
326 treatments (Figure 5). Evidence of cophylogeny at the end of the experimental period was
327 additionally detected using PACo for control (PACo=0.998, $p=0.001$) and drought
328 (PACo=0.999, $p=0.002$) treatments.

329

330

331 *nifH* gene abundance varies over time and by host species

332 No significant trend in *nifH* gene abundance was observed as a result of drought treatment, but
333 significant differences in abundance were observed between host species and over time (Figure
334 6). The temperate grasses displayed a decrease in abundance over time to varying degrees, but
335 the tropical grasses did not. Ryegrass control samples were temporally stable, but drought
336 samples significantly decreased between day 1 and day 38 ($p_{\text{adj}}=0.004$). Tall Fescue
337 ($p_{\text{adj}}=0.02$) and Orchardgrass ($p_{\text{adj}}=0.006$) significantly decreased between sample day 1 and
338 38 regardless of treatment. Control samples of *B. brizantha* showed no significant changes over
339 time, but *nifH* copy number significantly increased over time in drought samples ($p_{\text{adj}}=0.001$).
340 *Brachiaria* hyb. showed no differences as a result of drought but significantly varied across days
341 (Day1 compared to 26 ($p_{\text{adj}}=0.001$) and day 33 ($p_{\text{adj}}=0.03$)). *B. decumbens* significantly
342 increased over time in both control and drought conditions ($p_{\text{adj}}>0.001$). The trends over time
343 for *nifH* abundance closely matched the trends observed in average UniFrac distance over time
344 for each host species (Supplementary Figure S6).

345 DISCUSSION

346 Evolutionary history impacts grass phyllosphere communities and their response to drought.
347 Consistent with previous studies across multiple plant species, we found that host species was
348 the most important factor influencing community assembly and that *Alphaproteobacteria*
349 dominated communities [10, 13, 54, 55]. Additionally, our study revealed that communities
350 changed over time and as a result of drought. We observed strong temporal patterns in which
351 *Gammaproteobacteria* were replaced over time by *Cytophagia*, similar to studies in switchgrass
352 that found *Gammaproteobacteria* were replaced throughout the growing season by
353 *Alphaproteobacteria* [56]. While temporal replacement occurred on each host species, degree of
354 replacement varied widely and between treatments. By the end of the experimental period,
355 *Cytophagia* was the dominant class on control plants while *Alphaproteobacteria* dominated
356 drought plants. The persistent presence of *Alphaproteobacteria*, in particular *Sphingomonas* and
357 *Methylobacterium*, throughout the experiment on control and drought stressed plants likely
358 resulted from niche partitioning and their complementary metabolisms uniquely suited to the
359 phyllosphere [11, 56]. In the phyllosphere, *Sphingomonas* survive on a wide range of substrates
360 due to high abundance of TonB receptors, while *Methylobacterium* can grow on one-carbon
361 compounds such as methanol, a byproduct of host cell-wall metabolism [11, 57]. Additionally,
362 their flexible metabolisms allow for adaptation to changing nutrient availability as leaf
363 conditions change. Not only are they able to survive the harsh phyllosphere environment, they
364 can promote plant growth and stress tolerance. Inoculation of *Sphingomonas* onto soybean plants
365 resulted in increased tolerance of drought conditions and *Methylobacterium* on leaf surfaces are
366 able to fix nitrogen and increase plant biomass production [58, 59]. The observed persistence
367 under stress conditions in combination with their functional benefits, could indicate

368 coevolutionary adaptation to life in the phyllosphere. Furthermore, *Sphingomonas* and
369 *Methylobacterium* should be explored as biofertilizers because of their widespread presence and
370 observed drought tolerance.

371

372 Host species effect on community assembly increased over time. On day one of the experiment,
373 host species accounted for 38% of community variability in the control samples compared to
374 57% on the last day. This likely results from host selection on community assembly; host species
375 selection increases over time as communities successfully establish, as more bacteria land on the
376 leaf surface through dispersal, and as communities change in relation to plant development [56,
377 60, 61]. Interestingly, the effect of host species overtime was different between the native
378 temperate grasses and the non-native tropical grasses. On the first day of the experiment, host
379 species exhibited similar influence on microbial communities from temperate and tropical
380 grasses. However, by the end of the experimental period, species explained 72% of the
381 variability on temperate grass hosts but only 39% on tropical grass hosts. The difference in effect
382 over time between the tropical and temperate grasses likely results from host-microbe
383 evolutionary relationships that exist for the native temperate species but not for the non-native
384 tropical species.

385

386 To understand if phyllosphere communities from temperate grasses experienced increased
387 selection compared to tropical grasses, we determined how ecological distance changed over
388 time for each host species. Since temperate grasses were grown in their native environment, we
389 expected increased host selection compared to tropical grasses. While change over time
390 accounted for similar amounts of variability in the tropical (27%) and temperate grasses (23%),

391 the average distance of communities found on each host species decreased in temperate grasses
392 but remained stable in tropical grasses. These significant differences suggest deterministic
393 assembly in the phyllosphere. In the temperate grasses, decreased distance could result from
394 increased selection caused by coevolved plant-microbe relationships. However, since tropical
395 grasses were not grown in an environment with their native microbiota, changes over time and
396 between species were more likely a result of host physiology.

397

398 Presence of phylosymbiosis under non-stressed conditions indicates host-species influences
399 community assembly and that bacterial communities are more similar to each other on plant
400 hosts that are more phylogenetically similar [62]. While phylosymbiosis could result from
401 coevolution or cospeciation, it can also result from differences in host ecological niche,
402 geographic locations, or host filtering in which related hosts have many shared traits [23, 63, 64].
403 Therefore, phylosymbiosis does not determine a specific mechanism. Presence of
404 phylosymbiosis demonstrates that phyllosphere community assembly is a deterministic process
405 but what is driving it is still not fully understood. By growing plants in the same environment,
406 we eliminated some of the confounding factors that might otherwise contribute to this
407 relationship such as differences in soil, weather patterns, or biogeographic separation. Previous
408 work across animal species concluded that when related hosts grown under identical conditions
409 maintain distinct microbial communities, it is analogous to microbial markers of host
410 evolutionary relationships [26].

411

412 Unsurprisingly, the cophylogenetic analysis revealed strong evidence of cophylogeny in the
413 temperate grass species with hundreds of significant correlations compared to only dozens

414 observed in the tropical grasses. Thus, cophylogenetic signal was stronger in the native
415 temperate grasses than in non-native tropical grasses. The differences between tropical and
416 temperate hosts further supports the idea that the host-species effect seen in the temperate grasses
417 is a result of coevolution as microbial members have adapted alongside their host.

418

419 **Microbial Community Response: An Adaptation to Drought**

420 Understanding how phyllosphere communities respond to drought in relation to their plant host
421 is important for understanding how we can use bacteria as biofertilizers to promote plant health
422 in the future. Interestingly, no difference in alpha diversity as a result of drought was observed
423 even though drought caused changes in bacterial community structures. Previous work found
424 that phyllosphere community diversity but not composition was related to plant community
425 productivity [6]. Therefore, shifts we are seeing in community structure but not in alpha diversity
426 could indicate microbial communities act as a stress response trait.

427

428 The forage grass species used in this experiment have varying degrees of drought tolerance
429 resulting from the different strategies used under drought stress. Drought tolerance is well
430 documented for the temperate grasses used in this study. Ryegrass is the most drought
431 susceptible and under field conditions tall fescue is the most drought tolerant [65, 66]. However,
432 when grown in pots, orchardgrass exhibits higher drought tolerance due to its abilities to take up
433 water in low soil moisture conditions, promote membrane stabilization, and protect its meristem
434 from dehydration [66, 67]. In the field, tall fescue has higher drought tolerance due to its ability
435 to form deep root networks, which were limited by the depth of pots in which they were grown.
436 The C4 tropical plants have greater water use efficiency due to their ability to maintain higher

437 photosynthetic rates under decreased water stress compared to the C3 temperate species [68, 69].
438 Additionally, they can form extensive root networks that enable high water uptake efficiency
439 from soil [70]. These levels of known drought tolerance correlate with the changes we saw in
440 microbial community structures. Ryegrass, the most drought susceptible, was the first to show
441 signs of change, followed by orchardgrass and tall fescue. The more drought tolerant tropical
442 grasses showed changes in microbial communities later than the temperate grasses. Even though
443 previous studies found *B. brizantha* and *B. decumbens* to have similar drought tolerances, we
444 observed changes in *B. brizantha* communities on day 33 but no significant changes as a result of
445 drought in *B. decumbens* [70, 71]. What remains unclear is, if changes in microbial community
446 structure are in direct response to drought or in response to changes in host physiology.

447

448 If changes in communities were in direct response to drought, we may expect to see a decrease in
449 phylosymbiosis as selection imposed by host species decreased and communities became more
450 similar to each other. Instead, we observed an increase in phylosymbiosis under drought stress
451 indicating that selection on microbial communities is increasing. Conversely, total
452 cophylogenetic associations decreased as a result of drought with less overall connections
453 between host and microbe phylogenies. However, not all host species showed similar trends,
454 indicating that microbial community response to drought, much like community assembly, is a
455 deterministic process facilitated by changes in plant host physiology. Because strong evidence of
456 phylosymbiosis and cophylogeny remain in spite of shifts in community structure, we propose
457 that phyllosphere communities are a plant stress response trait that has coevolved alongside its
458 plant host.

459

460 Despite the strong host species effect on microbial communities during drought, we used
461 machine learning to determine if there was a common response to drought across our host
462 species. Determining if any microbes invariably survive under drought conditions across our
463 range of hosts and have the potential to promote plant growth is important for determining
464 prospective bacteria to test as biofertilizers. Our ML pipeline accurately predicted if a sample
465 came from a drought stressed or control plant 87% of the time, confirming that there is a
466 common response to drought despite divergent communities. No single bacterial ASV was
467 responsible for model prediction, rather several ASVs provided minor predictive power. Of the
468 top 8 predictors, 5 from the order Actinomycetales were slightly elevated in drought samples
469 including *Microbacterium*. In the rhizosphere, *Microbacterium* can produce volatile compounds
470 that promote plant health and growth [72] and help regulate plant response to drought stress by
471 altering the metabolite profile to promote osmoregulation [73]. Thus, even though microbial
472 communities are host specific, core functions exist in phyllosphere communities across plant
473 hosts that enable microbial survival under harsh conditions while also offering functional support
474 to their plant host.

475

476 **Nitrogen Fixation: A Core Function**

477 We used nitrogen fixation as one example of an important function microbes provide plants.
478 Recent studies found phyllosphere communities input nitrogen into their ecosystems [9, 59, 74].
479 When we assessed our communities for nitrogen fixation potential, we found stable diazotroph
480 presence on every host species. However, *nifH* abundance was not negatively impacted by
481 drought. The occurrence of temporally stable diazotrophs on every host species indicates its

482 likely an important part of the functional core community, while the differential abundance
483 between host species points to the evolutionary relationships between plants and their microbes.

484

485 When under drought stress, the most drought tolerant host species, *B. decumbens*, exhibited
486 increased *nifH* abundance and a strong cophylogenetic relationship with the bacterial family
487 Oxalobacteraceae accounting for 14 of the 47 significant relationships. Additionally, relative
488 abundance of Oxalobacteraceae from the class Betaproteobacteria increased on *B. decumbens*
489 under drought. Oxalobacteraceae are adapted to oligotrophic conditions and some genera are
490 nitrogen-fixers (Supplementary Figure S7) [75]. Because of the sustained presence of *nifH* across
491 time and treatments in combination with their correlation to community structure, we propose
492 nitrogen-fixation as a keystone function of phyllosphere communities. This further supports our
493 hypothesis that microbial communities are a plant trait which help promote plant stress tolerance.

494

495 **Conclusion**

496 This study revealed phyllosphere community assembly is related to host evolutionary history.
497 The strong evidence of phyllosymbiosis in combination with increased selection and cophylogeny
498 in the native temperate grasses compared to the non-native tropical grasses, suggests that
499 microbial communities are a plant trait that coevolve alongside their plant hosts. The conserved
500 presence but differential abundance of important bacteria such as *Sphingomonas* and
501 *Methylobacterium*, and the functional potential of nitrogen fixation during drought stress further
502 support the idea that microbial communities are plant traits that evolve to promote plant growth
503 and stress tolerance. Future studies should look at the effect of inoculating plants with the
504 taxonomically and functionally important bacteria identified in this study that were also

505 temporally and drought stable. Creating biofertilizers with ecologically important and
506 evolutionarily selected microbes could promote plant health and tolerance to a changing climate.

507

508

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515

516

517 COMPLIANCE WITH ETHICAL STANDARDS

518 Conflict of interest.

519 The authors declare that they have no conflict of interest.

520

521

522 SUPPLEMENTARY INFORMATION

523 Supplementary information is available online only.

524 Single PDF file, 5.1 Mb of Supplementary information containing table of contents, methods,
525 tables, figures, and references.

526

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- 712
- 713

714 FIGURE LEGENDS

715

716 **Figure 1. Average relative abundance of bacteria from 57 plants (27 control and 30**

717 **drought) sampled at 5 separate time points over 38 days. (A)** The most dominant bacterial

718 classes changed over time, between host species, and as a result of drought. To understand the

719 composition of these classes, the average relative abundance of the genera from the three most

720 abundant classes were plotted. Genera included were present in greater than 0.25% average

721 relative abundance. At the end of 38 days when drought effect was strongest, we observed

722 significant differences as a result of drought in *Actinobacteria* ($P < 0.001$), *Bacilli* ($P = 0.006$), and

723 *Cytophagia* ($P = 0.001$) (calculated using TukeyHSD). Additionally, strong differences were

724 observed between host species with significant differences observed in *Actinobacteria*

725 ($P < 0.001$), *Alphaproteobacteria* ($P < 0.001$), *Bacilli* ($P < 0.001$), *Betaproteobacteria* ($P < 0.001$),

726 *Cytophagia* ($P < 0.001$), *Deltaproteobacteria* ($P = 0.001$), and *Gammaproteobacteria* ($P = 0.008$)

727 (B) The class Alphaproteobacteria was dominated by the genera *Sphingomonas* and

728 *Methylobacterium*, (C) Gammaproteobacteria was not consistently dominated by any individual

729 genera, and (D) the class Cytophagia was dominated by the genus *Hymenobacter*.

730

731

732 **Figure 2. Bacterial communities from each host species became more distinct over time and**

733 **were significantly impacted by drought stress.** NMDS ordination was plotted for each

734 sampling day using weighted UniFrac distances. PERMANOVA was conducted for each

735 corresponding day to determine how communities were changing over time and when drought

736 stress altered community structure.

737

738 **Figure 3. Phyllosphere communities became more similar over time regardless of treatment**

739 **on the temperate host species but stayed the same on the tropical host species.** Average

740 distance between samples from the same host species were calculated for each sampling day

741 using weighted UniFrac distance. Average distance within each host species significantly

742 decreased in the communities from temperate hosts but did not change in the communities from

743 the tropical hosts.

744

745

746 **Figure 4. The top eight ASVs important for predicting if samples were from control or**

747 **drought stressed plant hosts.** Average relative abundance of each of the eight ASVs is given

748 for control (n=27) and drought stressed plants (n=30) on day 38 of the experiment. ASV

749 identities provided in Supplementary Table S1.

750

751 **Figure 5. Cophylogenetic relationship analysis was conducted for (A) control plants (n=27)**

752 **and (B) drought stressed plants (n=30).** Blue lines in this tanglegram represent significant

753 associations between phyllosphere bacteria on the left and their plant hosts on the right measured

754 using ParaFitGlobal, which were determined if either of the ParaFit F statistics were below 0.05.

755 Numbers under host species identity indicate the number of significant associations that a host

756 species has with the bacterial phylogenetic tree. The bacterial phylogenetic tree was constructed

757 in QIIME2 using FastTree which infers approximately-maximum-likelihood phylogenetic trees.

758 The maximum-likelihood tree for the grass host phylogeny was constructed in MEGA. Only five

759 grass species were included because host sequence information was not available for the
760 *Brachiaria* hybrid.

761

762

763 **Figure 6. Abundance of the *nifH* gene was significantly different between host species and**
764 **changed over time.** However, it was not significantly impacted by drought stress. *nifH*
765 abundance was measured using qPCR and standardized to number of copies per gram of leaf
766 material.

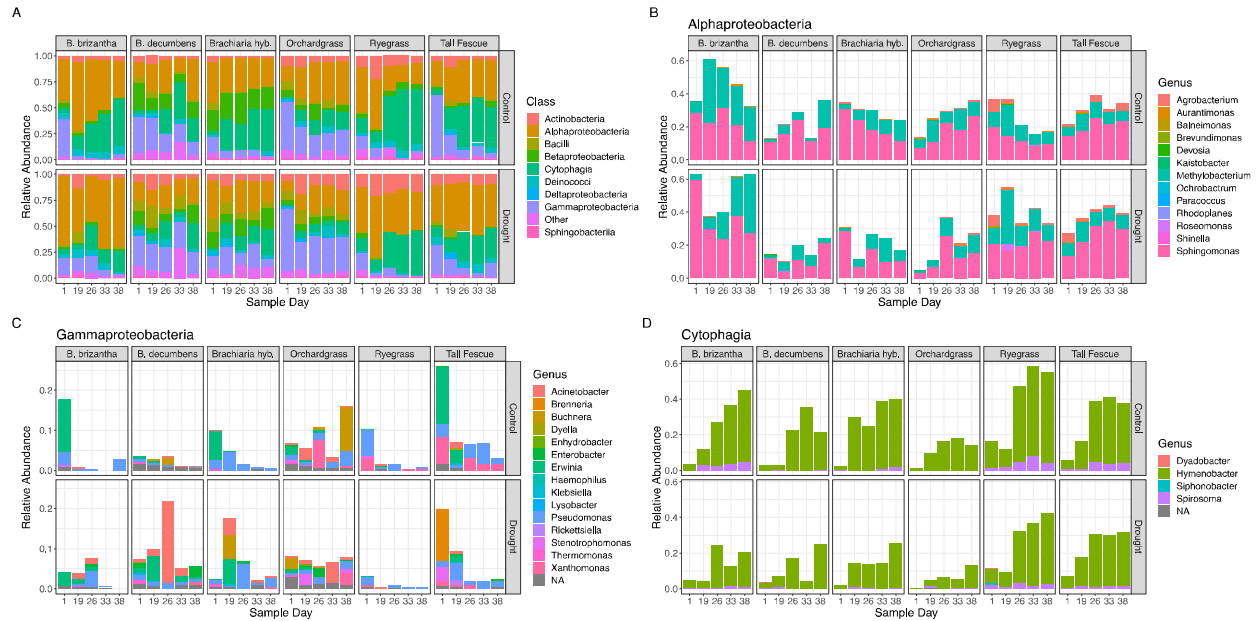
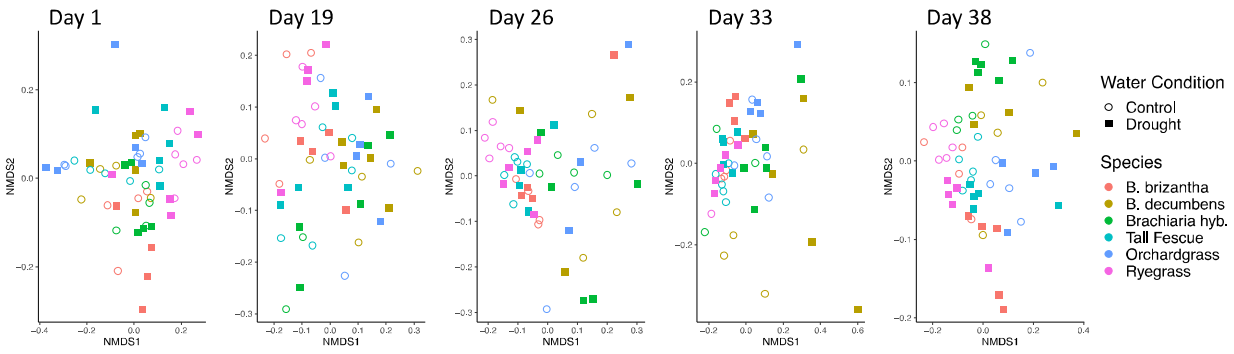


Figure 1. Average relative abundance of bacteria from 57 plants (27 control and 30 drought) sampled at 5 separate time points over 38 days. (A) The most dominant bacterial classes changed over time, between host species, and as a result of drought. To understand the composition of these classes, the average relative abundance of the genera from the three most abundant classes were plotted. Genera included were present in greater than 0.25% average relative abundance. At the end of 38 days when drought effect was strongest, we observed significant differences as a result of drought in *Actinobacteria* ($P < 0.001$), *Bacilli* ($P = 0.006$), and *Cytophagia* ($P = 0.001$) (calculated using TukeyHSD). Additionally, strong differences were observed between host species with significant differences observed in *Actinobacteria* ($P < 0.001$), *Alphaproteobacteria* ($P < 0.001$), *Bacilli* ($P < 0.001$), *Betaproteobacteria* ($P < 0.001$), *Cytophagia* ($P < 0.001$), *Deltaproteobacteria* ($P = 0.001$), and *Gammaproteobacteria* ($P = 0.008$) (B) The class Alphaproteobacteria was dominated by the genera *Shingomonas* and *Methylobacterium*, (C) Gammaproteobacteria was not consistently dominated by any individual genera, and (D) the class Cytophagia was dominated by the genus *Hymenobacter*.



Weighted Unifrac										
	Day 1		Day 19		Day 26		Day 33		Day 38	
	R ²	P-value	R ²	P-value	R ²	P-value	R ²	P-value	R ²	P-value
Treatment	0.016	0.269	0.022	0.142	0.029	0.067	0.063	0.003**	0.047	0.003**
Species	0.346	0.001***	0.282	0.001***	0.312	0.001***	0.381	0.001***	0.390	0.001***
Species* Treatment	0.069	0.377	0.109	0.056	0.067	0.586	0.074	0.267	0.106	0.005**

Figure 2. Bacterial communities from each host species became more distinct over time and were significantly impacted by drought stress. NMDS ordination was plotted for each sampling day using weighted UniFrac distances. PERMANOVA was conducted for each corresponding day to determine how communities were changing over time and when drought stress altered community structure.

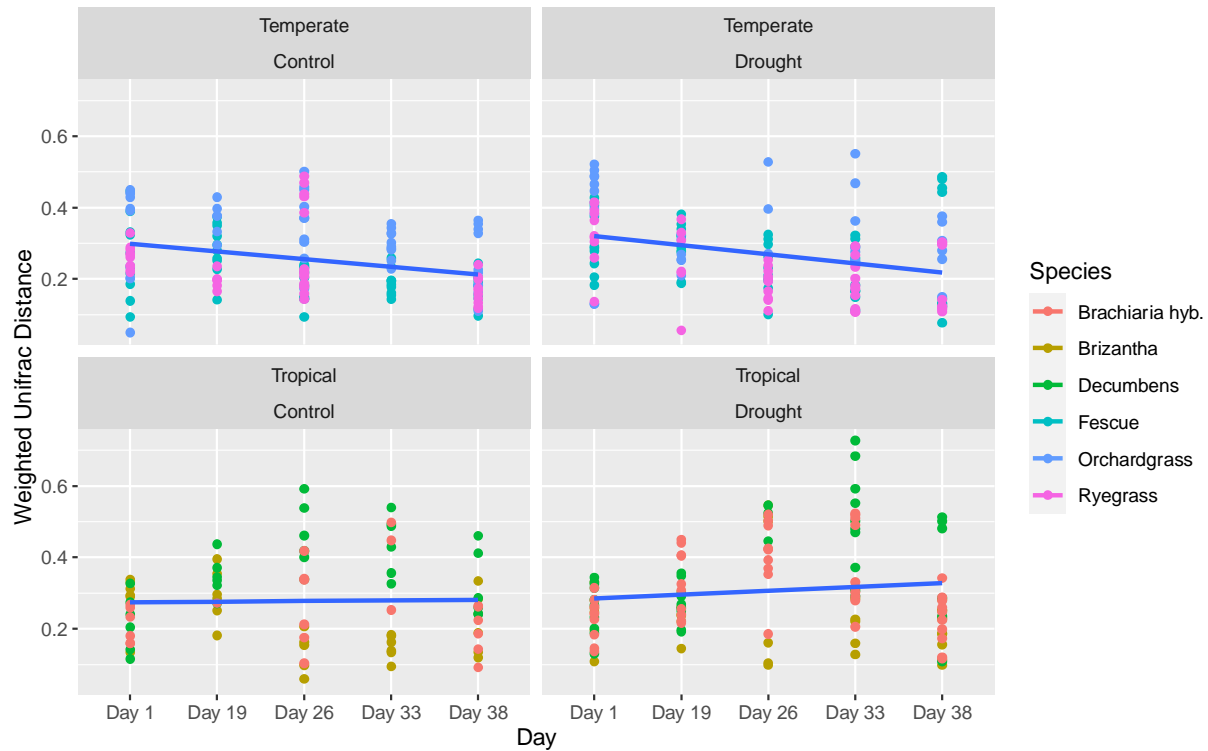


Figure 3. Phyllosphere communities became more similar over time regardless of treatment on the temperate host species but stayed the same on the tropical host species. Average distance between samples from the same host species were calculated for each sampling day using weighted UniFrac distance. Average distance within each host species significantly decreased in the communities from temperate hosts but did not change in the communities from the tropical hosts.

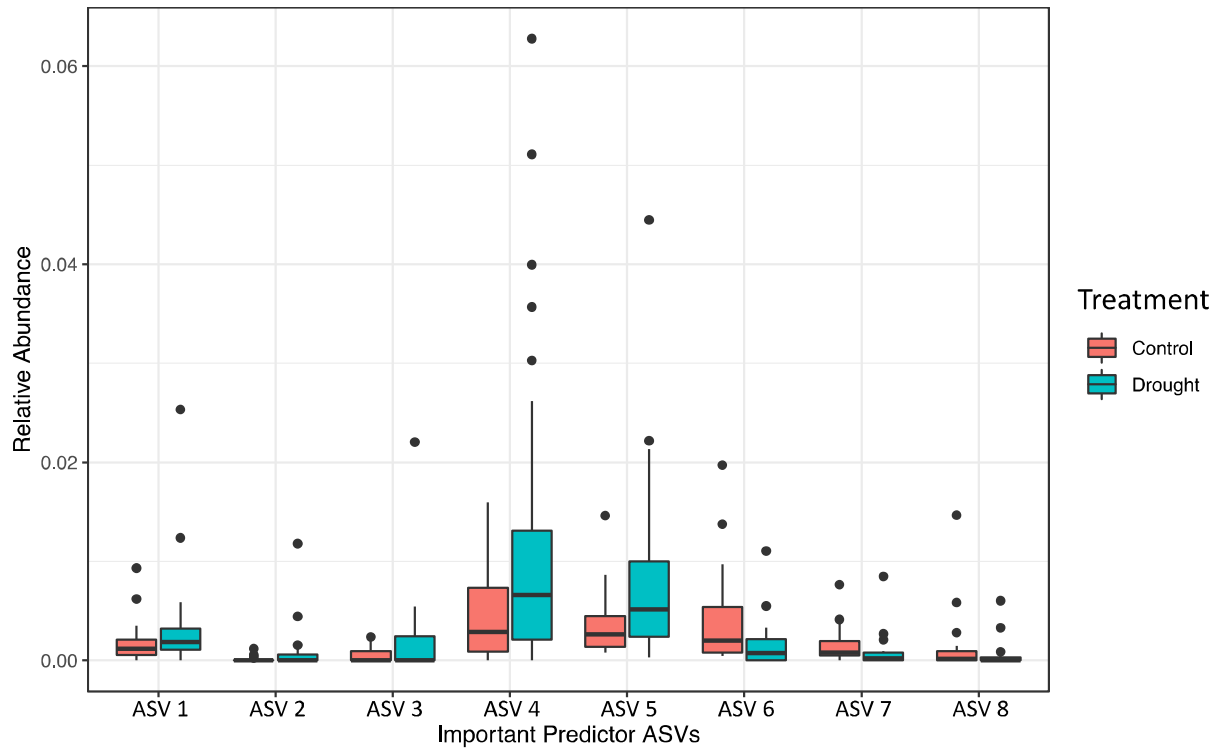


Figure 4. The top eight ASVs important for predicting if samples were from control or drought stressed plant hosts. Average relative abundance of each of the eight ASVs is given for control (n=27) and drought stressed plants (n=30) on day 38 of the experiment. ASV identities provided in Supplementary Table S1.

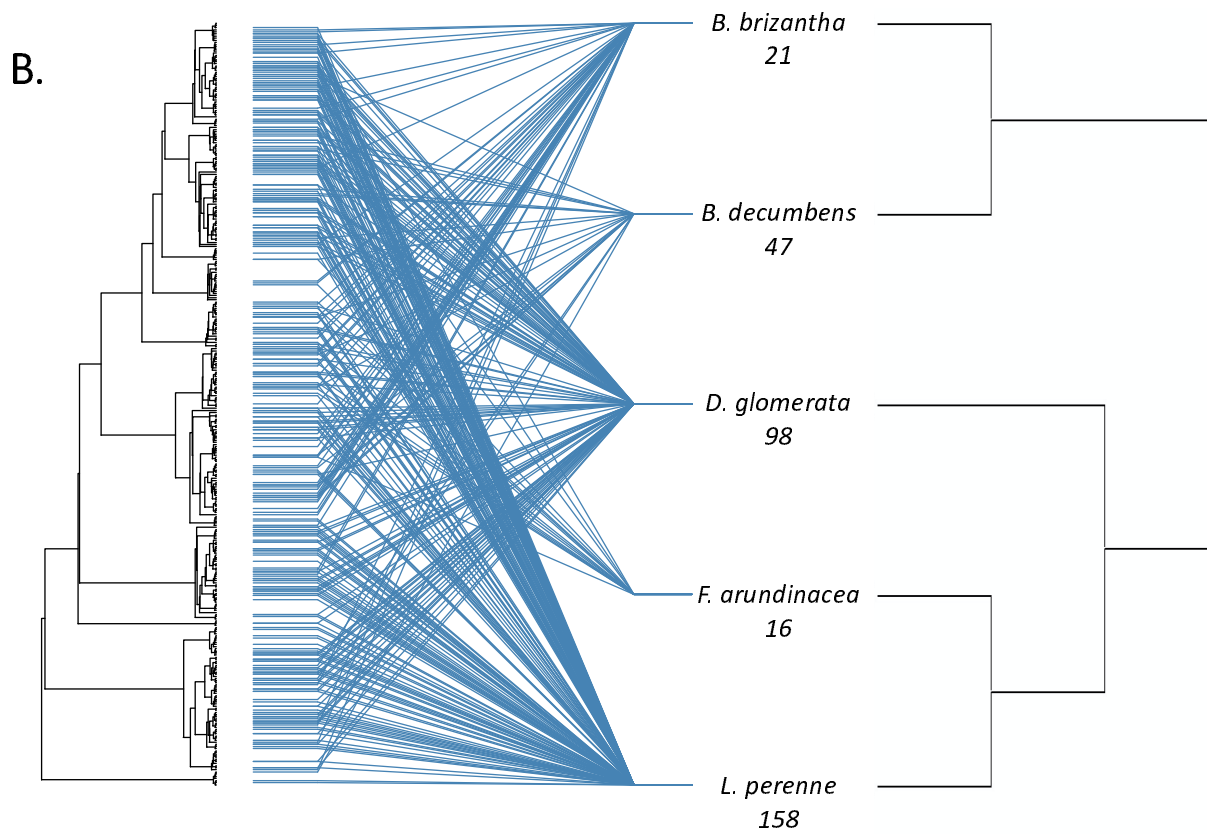
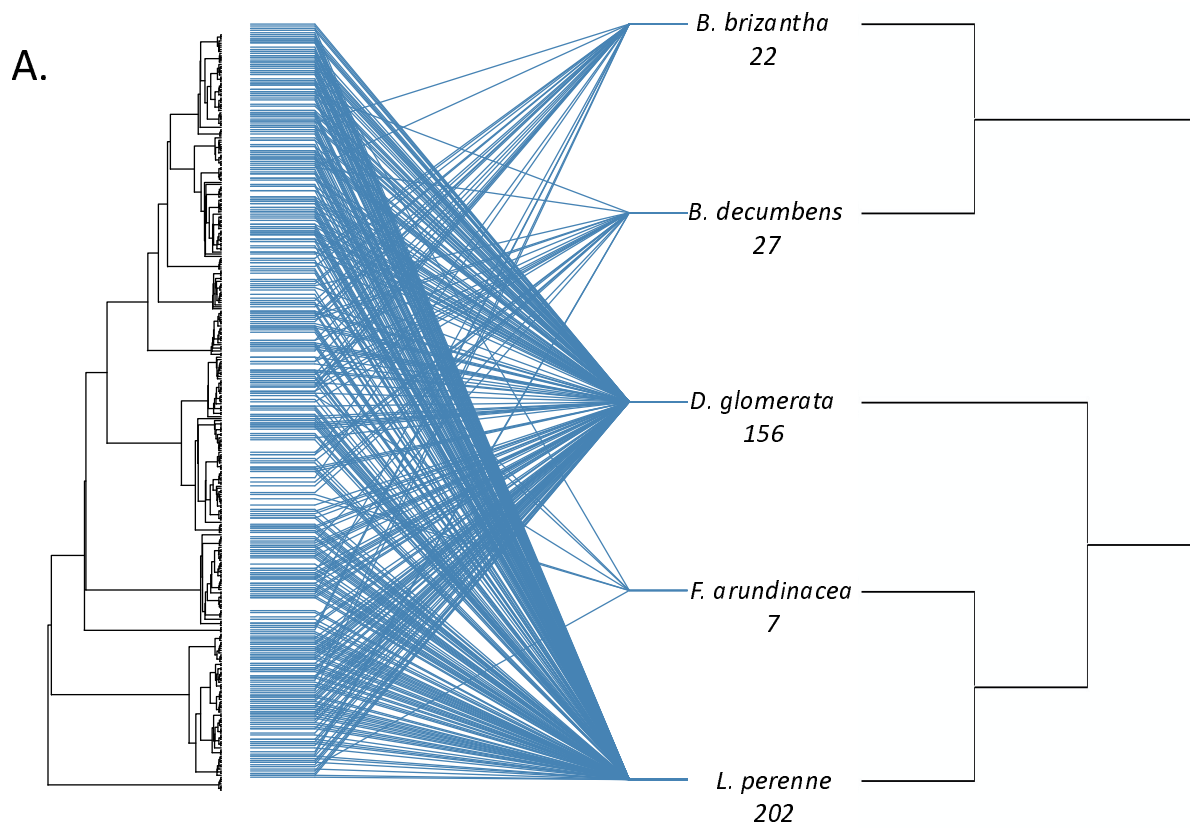


Figure 5. Cophylogenetic relationship analysis was conducted for (A) control plants (n=27) and (B) drought stressed plants (n=30). Blue lines in this tanglegram represent significant associations between phyllosphere bacteria on the left and their plant hosts on the right measured using ParaFitGlobal, which were determined if either of the ParaFit F statistics were below 0.05. Numbers under host species identity indicate the number of significant associations that a host species has with the bacterial phylogenetic tree. The bacterial phylogenetic tree was constructed in QIIME2 using FastTree which infers approximately-maximum-likelihood phylogenetic trees. The maximum-likelihood tree for the grass host phylogeny was constructed in MEGA. Only five grass species were included because host sequence information was not available for the *Brachiaria* hybrid.

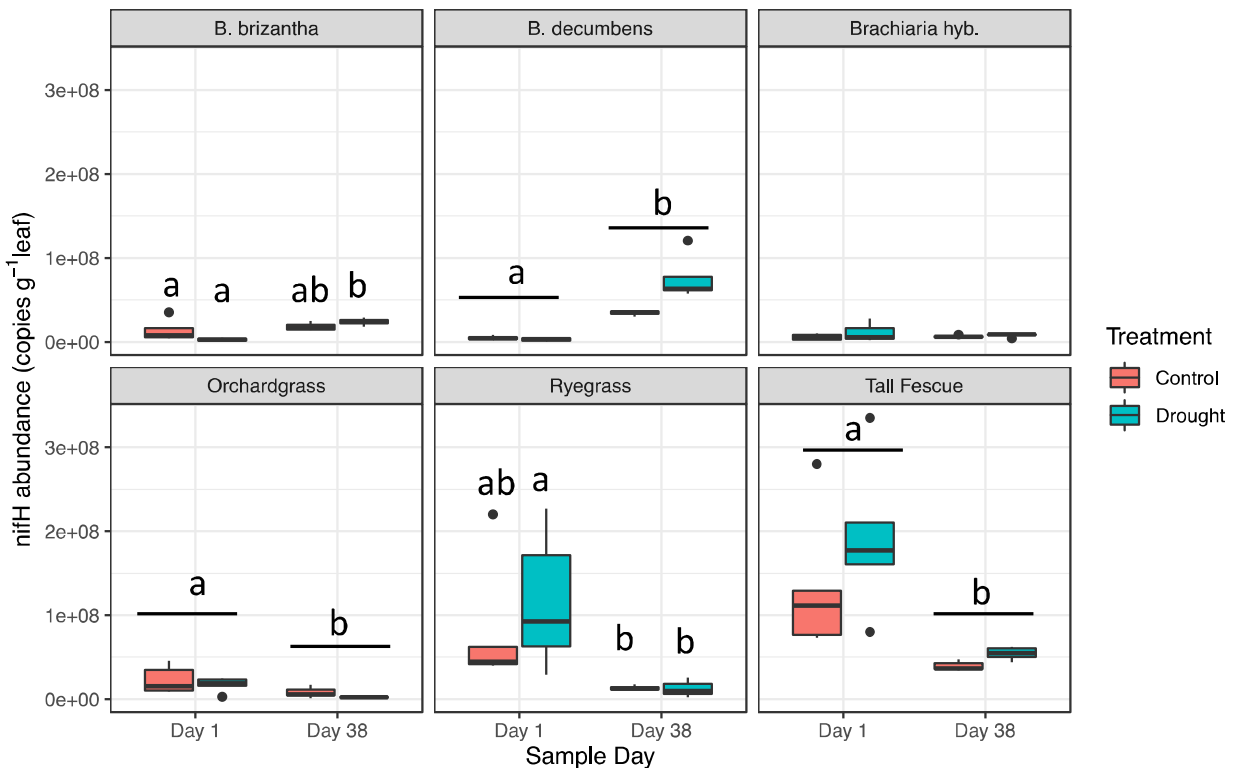


Figure 6. Abundance of the *nifH* gene was significantly different between host species and changed over time. However, it was not significantly impacted by drought stress. *nifH* abundance was measured using qPCR and standardized to number of copies per gram of leaf material.

Table 1. Phyllosphere community structure on native temperate and non-native tropical grasses change over time (day) and are impacted by host species and drought treatment. Impact of each variable on community structure was determined using a PERMANOVA on weighted UniFrac distance measures.

Variable	Weighted UniFrac Distance	
	R ²	P-value
Treatment	0.019	0.001***
Day	0.142	0.001***
Species	0.191	0.001***
Treatment*Day	0.017	0.008**
Treatment*Species	0.029	0.001***
Day*Species	0.101	0.001***
Treatment*Day*Species	0.043	0.0572

Table 2. The effect of host species on phyllosphere community composition on non-stressed hosts increased over time. The impact of host species on community structure was measured for communities from the well-watered control host plants at (A) the beginning (day 1) and (B) the end (day 38) of the drought period. Influence of host species was determined for all hosts together and separately for the native temperate grasses and non-native tropical grasses using a PERMANOVA of weighted UniFrac distances.

A.

DAY 1	All host plants		Temperate		Tropical	
	R ²	P-value	R ²	P-value	R ²	P-value
Species	0.38	0.01	0.34	0.039	0.31	0.094
Residuals	0.62		0.65		0.69	

B.

DAY 38	All host plants		Temperate		Tropical	
	R ²	P-value	R ²	P-value	R ²	P-value
Species	0.57	0.001	0.72	0.001	0.39	0.024
Residuals	0.43		0.28		0.61	