

1 **Arid Ecosystem Vegetation Canopy-Gap Dichotomy: Influence on Soil Microbial**  
2 **Composition and Nutrient Cycling Functional Potential**

3

4

5 Priyanka Kushwaha,<sup>a#</sup> Julia W. Neilson,<sup>a#</sup> Albert Barberán,<sup>a</sup> Yongjian Chen,<sup>a</sup> Catherine G.  
6 Fontana,<sup>a</sup> Bradley J. Butterfield,<sup>b</sup> Raina M. Maier<sup>a</sup>

7 <sup>a</sup>Department of Environmental Science, University of Arizona, Tucson, AZ

8 <sup>b</sup>Center for Ecosystem Science and Society (ECOSS) and Department of Biological Sciences,  
9 Northern Arizona University, Flagstaff, AZ

10

11 Running Head: Arid Soil Microbiome and Nutrient Cycling Potential

12

13 #Address correspondence to Priyanka Kushwaha, [pkushwaha@arizona.edu](mailto:pkushwaha@arizona.edu) and Julia W. Neilson,  
14 [jneilson@arizona.edu](mailto:jneilson@arizona.edu)

15

16 **Keywords:** deserts, microsite, soil microbiome, nutrient mineralization, ureC, urease enzymatic  
17 activity, aridity, functional traits

18

19

20

21

22

23 **ABSTRACT**

24 Increasing temperatures and drought in desert ecosystems are predicted to cause decreased  
25 vegetation density combined with barren ground expansion. It remains unclear how nutrient  
26 availability, microbial diversity, and the associated functional capacity vary between vegetated-  
27 canopy and gap soils. The specific aim of this study was to characterize canopy vs gap microsite  
28 effect on soil microbial diversity, the capacity of gap soils to serve as a canopy soil microbial  
29 reservoir, nitrogen (N)-mineralization genetic potential (*ureC* gene abundance) and urease enzyme  
30 activity, and microbial-nutrient pool associations in four arid-hyperarid geolocations of the  
31 western Sonoran Desert, Arizona (USA). Microsite combined with geolocation explained 57% and  
32 45.8% of the observed variation in bacterial/archaeal and fungal community composition,  
33 respectively. A core microbiome of amplicon sequence variants was shared between the canopy  
34 and gap soil communities; however, canopy soils included abundant taxa that were not present in  
35 associated gap communities, thereby suggesting that these taxa cannot be sourced from the  
36 associated gap soils. Linear mixed-effects models showed that canopy-soils have significantly  
37 higher microbial richness, nutrient content, and organic N-mineralization genetic and functional  
38 capacity. Furthermore, *ureC* gene abundance was detected in all samples suggesting that *ureC* is  
39 a relevant indicator of N-mineralization in deserts. Additionally, novel phylogenetic associations  
40 were observed for *ureC* with the majority belonging to *Actinobacteria* and uncharacterized  
41 bacteria. Thus, key N-mineralization functional capacity is associated with a dominant desert  
42 phylum. Overall, these results suggest that lower microbial diversity and functional capacity in  
43 gap soils may impact ecosystem sustainability as aridity drives open-space expansion in deserts.

44 **IMPORTANCE**

45 Increasing aridity will drive a shift in desert vegetation and interspace gap (microsite) structure  
46 toward gap expansion. To evaluate the impact of gap expansion, we assess microsite effects on  
47 soil nutrients, microbiome community composition and functional capacity, and the potential of  
48 gap soils to serve as microbial reservoirs for plant-root associated microbiomes in an arid  
49 ecosystem. Results indicate that gap soils have significantly lower bioavailable nutrients, microbial  
50 richness, and N-mineralization functional capacity. Further, abundance of the bacterial urease gene  
51 (*ureC*) correlates strongly with N-availability and its major phylogenetic association is with  
52 *Actinobacteria*, the dominant phylum found in deserts. This finding is relevant because it identifies  
53 an important N-mineralization capacity indicator in the arid soil microbiome. Such indicators are  
54 needed to understand the relationships between interplant gap expansion and microbial diversity  
55 and functional potential associated with plant sustainability. This will be a critical step in recovery  
56 of land degraded by aridity stress.

57

58 **1. INTRODUCTION**

59 Desert ecosystems are characterized by plant spatial heterogeneity, where vegetation cover is  
60 present in patches separated by gaps (1). Vegetation patches capture dust, water, and nutrients  
61 from wind and water erosion, forming fertility or resource islands in soils (2, 3). According to  
62 climate model predictions, a rise in global temperatures with associated extreme weather events  
63 and drought conditions will lead to an increase in aridity in these regions (4). This in turn will  
64 affect essential ecosystem services (5) through transformation of vegetation patterns as well as gap  
65 expansion in drylands (6), and might result in irreversible shifts from healthy to degraded soils (7).

66 Presently, there is limited knowledge of how loss of vegetative cover accompanied by expanded  
67 gaps will affect arid ecosystem services.

68 Previous work has shown that desert soil microbial communities are taxonomically distinct  
69 from other biomes and have lower functional diversity with respect to nutrient cycling (8). In  
70 addition, studies conducted in drylands have demonstrated that an increase in aridity results in a  
71 decrease in microbial abundance and diversity (9, 10). It is also known that microbial abundance  
72 and diversity in deserts are controlled by microsites (vegetation and gap microsites includesoils  
73 under plant canopies versus interspace gaps devoid of any vegetation) (11, 12), size of the fertility  
74 island (13, 14), and precipitation (15). However, the link between microbial diversity and the  
75 associated genetic potential and functional capacity remains largely unclear in deserts. Predicted  
76 global climatic changes that shift desert microsite structure may in turn influence the microsite-  
77 specific microbial community composition and its functional capacity. Therefore, it is critical to  
78 understand microbial functions not only under the vegetation canopy soils but also in the gaps.

79 A critical nutrient cycling service in desert ecosystems is soil organic matter mineralization, a  
80 process carried out by heterotrophic bacteria and fungi. Microbial mineralization of organic matter  
81 is a primary source of important inorganic nutrients including nitrogen (N) (16). In fact, N-  
82 availability is the main limiting factor for net primary productivity and ecosystem function in  
83 desert ecosystems. However, knowledge concerning the consortia of microorganisms that drive  
84 mineralization of organic N pools to inorganic forms in deserts is limited. N-mineralization is an  
85 important function impacting plant available nitrogen supplies. N-mineralization begins with  
86 proteolysis, the breakdown of proteins into peptides, amines, and amides (17). These proteolysis  
87 products are converted into urea and urea is further hydrolyzed into ammonium by the enzyme  
88 urease (18). Whereas proteases are ubiquitous and produced by all forms of life (19), the processes

89 of ureolysis are driven by specialized microbial guilds (18). Thus, the genetic potential and  
90 functional capacity of these ureolytic microorganisms should be evaluated to assess N-availability  
91 between canopy and gap soil microsites in deserts.

92 The goal of this study is to evaluate the effects of soil microsites (canopy vs gap) in different  
93 geolocations on the coupled association of microbial community composition, functional nutrient  
94 cycling capacity, and nutrient pools in the Sonoran Desert of the southwestern US. To do so, we  
95 compare microbial community composition and mineralization potential of gap and vegetated  
96 fertility islands at four arid to hyperarid locations (geolocations) in the *Larrea tridentata-Ambrosia*  
97 *dumosa* (creosote bush-white bursage) ecosystem of the western Sonoran Desert, Arizona (Fig. 1).  
98 The creosote bush-white bursage is the dominant vegetation ecosystem of the western Sonoran  
99 Desert and vegetation density is controlled by aridity (20). The study was designed to evaluate the  
100 impact of desert microsite and geolocation on: 1) microbial diversity and community composition,  
101 2) capacity of the gap soils to serve as a reservoir for the plant root-zone microbiome, and 3)  
102 associations between N-mineralization gene abundance (bacterial *ureC*, ureolysis), N enzyme  
103 activity (urease), and soil nutrient pools in the four Sonoran Desert geolocations. Abundance of  
104 the *ureC* gene was targeted to compare microbial populations critical to the cycling of organic N  
105 and to address whether urease functional capacities shift to alternate taxa in desert ecosystems.

## 106 **2. RESULTS**

### 107 **2.1 Soil Physicochemical Properties**

108 Physicochemical analysis was performed on soil samples collected from the vegetation canopy  
109 and gap soil microsites. Vegetation canopy was defined as the soil samples collected under the  
110 canopy of the creosote bush-white bursage vegetation, and gap was defined as the interplant barren  
111 spaces devoid of vegetation or biocrust (Fig. 1). Physicochemical analysis indicated that nutrient

112 values, particularly dissolved organic carbon (DOC), total organic carbon (TOC), total nitrogen  
113 (TN), dissolved nitrogen (DN), ammonium (NH<sub>4</sub>-N), nitrate (NO<sub>3</sub>-N), and bioavailable  
114 phosphorus (BAP) were significantly higher in canopy soils compared to gap soils (Table 1; Table  
115 S1). The soil chemical parameters that most strongly and significantly associated with canopy soils  
116 as opposed to gap soils included DN, NH<sub>4</sub>-N, TN, BAP, and DOC (Linear mixed-effect models;  
117 marginal R<sup>2</sup>>0.30; Table 1). Specifically, the content of DN was on average 1.5-to 4.6 -fold higher  
118 in canopy soils than in gap soils. Total carbon (TC) and DOC also showed significant geolocation  
119 effects, thereby indicating that certain nutrient pools varied with different geographic sites.  
120 Additionally, the major fraction of TN and DN was primarily organic as inorganic N values (NH<sub>4</sub>-  
121 N and NO<sub>3</sub>-N) were relatively low (Table S1).

## 122 **2.2 Microbial Community Composition**

123 The bacterial/archaeal communities across the four arid to hyperarid geolocations of the western  
124 Sonoran Desert were dominated by: *Actinobacteria* (40.9%), *Proteobacteria* (24.2%), *Firmicutes*  
125 (7.4%), *Gemmatimonadetes* (6%), *Chloroflexi* (5.2%), *Acidobacteria* (4.9%), *Planctomycetes*  
126 (2.1%), *Bacteroidetes* (1.5%), and *Verrucomicrobia* (0.9%) (Fig. S1). A comparison of this  
127 community composition with other desert and non-desert biomes revealed a profile distinct from  
128 non-desert biomes (Fig. S1; tropical/temperate forests and prairie grasslands). Specifically, the  
129 relative abundance of *Actinobacteria* (15.3-45.3%) and *Gemmatimonadetes* (1.3-6.7%) was  
130 consistently greater in desert biomes (Fig. S1), whereas *Acidobacteria* (17.3-28.1%), and  
131 *Verrucomicrobia* (22.8-28.1%) were more abundant in forests and grasslands (Fig. S1; Table S2).

132 For the sites sampled for this study, compositional differences between microbial  
133 communities in canopy and gap soils were assessed using *DESeq2* (Fig. S2 & S3). For  
134 bacteria/archaea, the relative abundances of *Actinobacteria*, *Firmicutes*, *Chloroflexi*, and

135 *Armatimonadetes* were significantly higher in gap soils ( $p < 0.05$ ), whereas *Proteobacteria*,  
136 *Gemmatimonadetes*, *Entotheonellaeota*, *Deinococcus-Thermus*, *Dependentiae*, and  
137 *Nanoarchaeota* had higher relative abundances in canopy soils (Fig. S2A). At the class level,  
138 *Alphaproteobacteria* and *Deltaproteobacteria* had greater relative abundances in the canopy soils,  
139 whereas *Bacilli*, *Rubrobacteria*, and *Thermoleophilia* classes had higher relative abundances in  
140 gap soils (Fig. S2B). There were 58 and 31 genera that were significantly more abundant in canopy  
141 and gap soils, respectively (Fig. S2C). Of these, genera belonging to *Actinobacteria* were split in  
142 relative abundance; eleven and nine genera had greater relative abundance in canopy and gap soils,  
143 respectively (Fig. S2C). Twenty *Alphaproteobacteria* and eight *Gammaproteobacteria* genera  
144 were more abundant in canopy soils and seven *Bacilli* genera were more abundant in gap soils  
145 (Fig. S2C).

146         Likewise, fungal community composition was significantly different between canopy and  
147 gap soils (Fig. S3). Ascomycota had greater relative abundance in canopy soils ( $p < 0.05$ ), whereas  
148 the relative abundances of Basidiomycota and Glomeromycota were significantly greater in gap  
149 soils (Fig. S3A). At the class level, Leotiomyces was the most abundant class in the canopy soils,  
150 whereas Geminibasidiomyces, Glomeromyces, and Pezizomyces were more abundant in the  
151 gap soils (Fig. S3B). In all, 37 fungal genera were significantly different between canopy and gap  
152 soil microsites, and of these, 32 genera had greater relative abundance in canopy soils and only  
153 five were more abundant in gap soils (Fig. S3C). The majority of the genera that were abundant in  
154 canopy soils belonged to Ascomycota, whereas the five genera with greater relative abundance in  
155 gap soils belonged to Ascomycota, Basidiomycota, Calcarisporiellomycota, and Glomeromycota  
156 (Fig. S3C).

### 157 **2.3 Microbial Community Diversity**

158 Linear mixed-effects models showed that the effect of microsite was greater in canopy soils than  
159 in gap soils and this effect was stronger for fungi (marginal  $R^2=0.58$ ) than for bacteria and archaea  
160 (marginal  $R^2=0.20$ ; Table 1). For bacteria/archaea, there was no difference between canopy and  
161 gap soil richness at sites 1 and 2; however, richness was greater in canopy soils than gap soils for  
162 the more arid sites 3 and 4 (Fig. 2A). In contrast, fungal richness was consistently greater under  
163 plant canopies across all the four geolocations (Fig. 2C). Non-metric multidimensional scaling  
164 (NMDS) revealed that soil microbial community composition was different between canopy and  
165 gap soils and among geolocations (Fig. 2B & D). In addition, geolocation 4 clustered separately  
166 from the other sites. Permutational multivariate analysis of variance (PERMANOVA) showed that  
167 geolocation explained 20% of the variation in bacterial/archaeal community composition  
168 ( $p=0.001$ ); and microsite within each location explained an additional 37% variation ( $p=0.001$ ;  
169 Fig. 2B). For fungal communities, 16.3% of the variation in community composition ( $p=0.001$ )  
170 was explained by geolocation, and microsite within geolocation explained an additional 29.5%  
171 variation ( $p=0.001$ ; Fig. 2D). Additionally, multivariate dispersion analysis revealed that the  
172 bacterial/archaeal ( $p=0.28$ ) and fungal ( $p=0.19$ ) community dispersions were not significantly  
173 different between microsites (Fig. S4), suggesting that the PERMANOVA results were not biased  
174 by the variation within microsites.

### 175 **2.4 Core Microbiome and Microsite Specific ASVs**

176 The number of unique and shared amplicon sequence variants (ASVs) in canopy and gap soil  
177 microbial communities was examined to investigate whether the gap microbial community has the  
178 capacity to serve as a reservoir for the plant root-zone microbiome community recruitment. Unique  
179 ASVs are identified as those present in either the canopy or the gap soils, but absent from the other,

180 whereas shared ASVs refer to those present in both microsites. Here we define the core  
181 microbiome of this system as the shared ASVs that are present in both canopy and gap soils. In  
182 this analysis, ASV number represents richness and ASV relative abundance (at the class level)  
183 represents community composition (Fig. 3). A total of 16,484 ASVs were identified for  
184 bacteria/archaea across all the sites. The richness of unique bacterial/archaeal ASVs was higher in  
185 canopy soils (7,573) than in gap soils (5,254) and the core microbiome was comprised of 3,657  
186 ASVs (Fig. 3A). At the site level, unique canopy soil ASVs represented 34-51% of the richness  
187 across the four geolocations (Fig. 3B) and 18-49% of the canopy community composition (Fig.  
188 3C; Table S3). In contrast, the core microbiome represented 51-82% (Fig. 3C; Table S4) and 71-  
189 81% (Table S4) of the community composition in the canopy and gap soil communities,  
190 respectively.

191         There was a total of 2,234 ASVs identified for fungi across all sites. Similar to the  
192 bacteria/archaea, fungal richness was also higher in canopy soils (1,295 ASVs) than gap (500  
193 ASVs) soils (Fig. 3D). However, the difference was much more pronounced for fungi; the ratio of  
194 unique canopy soil to gap soil fungal ASVs was 1.8 times greater than for bacterial/archaeal ASVs.  
195 Further, the core microbiome was comprised of just 439 ASVs (Fig. 3D). At the site level, unique  
196 fungal ASVs in canopy soils represented 48-73% of the richness while the core microbiome  
197 represented only 10-21% of the richness (Fig. 3E). For community composition, unique fungal  
198 canopy soils ASVs comprised 26-70% of the community (Fig. 3F; Table S5), in contrast to the  
199 core microbiome which represented 30-75% of the relative abundance of fungal taxa in canopy  
200 soils (Fig. 3F; Table S6).

## 201 **2.5 Fungal Guild Predictions**

202 Fungal functional guilds were determined from fungal ASVs using the FUNGuild tool. Linear  
203 mixed-effects models showed that the functional guild variability was influenced more  
204 significantly by microsite than geolocation (Table 1). Arbuscular mycorrhizal fungi represented  
205 the only guild that associated more significantly with gap soils (marginal  $R^2=0.38$ ; Fig. 4A). In  
206 contrast, plant pathogens (marginal  $R^2=0.68$ ) and wood saprotrophs (marginal  $R^2=0.57$ ) were more  
207 abundant in canopy soils (Fig. 4B & 4C; Table 1). In addition, epiphytes and endophytes also had  
208 higher abundance in canopy soils, which are functionally associated with plants (Table 1).

## 209 **2.6 qPCR Bacterial Gene Abundance**

210 The 16S rRNA and *ureC* bacterial genes were used to quantify bacteria and evaluate specific N-  
211 mineralization genetic potential of the bacterial community, respectively (Fig. 5). Bacterial  
212 abundance ranged from 7.83 to 8.65 log 16S rRNA gene copies  $g^{-1}$  soil (Fig. 5A). The 16S rRNA  
213 gene abundance indicated that bacteria were significantly more abundant in canopy soils than gap  
214 soils; however, the variability in bacterial abundance was explained more by geolocation than  
215 microsite (Table 1, compare the marginal and conditional  $R^2$  values). In contrast, the bacterial  
216 *ureC* gene abundance showed a strong and significant microsite effect and a weaker geolocation  
217 effect (Table 1). The *ureC* gene was detected in all but one sample and the abundance ranged from  
218 6.5-7.5 log gene copies  $g^{-1}$  soil (Fig. 5B). Similar to the *ureC* gene abundance pattern, urease  
219 enzyme activity was detected in all the samples ( $n=24$ ; Table S1). Both urease enzyme activity  
220 (marginal  $R^2=0.31$ ) and *ureC* gene abundance (marginal  $R^2=0.38$ ) were greater in canopy soils  
221 (Table 1). A limited clone library was prepared from Sonoran Desert soil samples and a  
222 phylogenetic analysis of associated *ureC* sequences revealed that 51% of the clones did not have

223 a phylogenetic affiliation in the database, 42% belonged to *Actinobacteria*, and 4% of the clones  
224 were associated with *Proteobacteria* (Table S7).

## 225 **2.7 Correlations between Soil Nutrients and Gene Abundance**

226 The organic matter pool is an important source of limiting nutrients such as nitrogen. Correlations  
227 were evaluated between TOC, different N forms and N-mineralization functional potential to  
228 characterize the associations between N-pools and microbial N-mineralization capacity.  
229 Correlations between TOC and TN ( $r=0.72$ ,  $p<0.001$ ) and TOC and DN ( $r=0.59$ ,  $p<0.01$ ) were  
230 stronger and more significant than between TOC and ammonium or nitrate (Table S8) indicating  
231 that the major N pool in the deserts is organic. Further analysis of N forms and gene abundance  
232 showed significant positive correlations between 16S rRNA gene abundance and both TN and DN  
233 (Table S9). In addition, bacterial *ureC* gene abundance strongly correlated with all N forms, but  
234 the strongest correlations were observed between *ureC* and TN and DN ( $r>0.70$ ; Table S9).

## 235 **3. DISCUSSION**

236 Climate change-induced acceleration of dryland expansion is predicted to stress arid ecosystems  
237 and transform vegetation patterns leading to expansion of interplant gaps (6, 7). Thus, there is a  
238 need to understand the significance of the canopy-gap microsite dichotomy to the soil nutrient  
239 pools and microbial nutrient cycling capacity of desert ecosystems to predict potential impacts of  
240 factors such as gap expansion. This study describes the effect of microsite (canopy versus gap  
241 soils) on microbial community composition, functional potential, and microbial associations with  
242 soil nutrient pools across four geolocations in the western Sonoran Desert, AZ. Specifically, we  
243 demonstrate that (1) there is a stronger microsite than geolocation effect on the microbial  
244 community composition; 2) there is significantly greater phylogenetic, genetic, and functional N-  
245 mineralization capacity associated with canopy soils relative to the gap soils; and 3) gap soil

246 microbial community lacks sufficient microbial diversity to serve as a complete reservoir for the  
247 plant root-zone microbiome.

### 248 **3.1 Microsite Differences in Microbial Diversity**

249 The Sonoran Desert samples showed that canopy soils have a higher microbial diversity than the  
250 gap soils. These results are in contrast to a study of desert palms conducted in semi-arid to arid  
251 regions of Tunisian Saharan Desert that found bacterial diversity to be greater in bulk soils  
252 compared to rhizosphere in five out of the seven sampling sites (21). Interestingly, a study in the  
253 Namib Desert demonstrated different patterns for bacterial and fungal diversity in bulk and  
254 speargrass rhizosphere soils. Bulk soils had lower fungal diversity than speargrass rhizosphere  
255 soils, whereas bacterial diversity was comparable in bulk and rhizosphere soils (22). It is important  
256 to note that the aridity conditions of our Sonoran Desert sites are more similar to the Namib Desert  
257 (arid-hyperarid) whereas the Tunisian Saharan Desert sites range from semi-arid to arid. Taken  
258 together, our results and the results from these previous studies suggest that the ratio of microbial  
259 richness of bulk to rhizosphere soils decreases as aridity increases.

260 This study also revealed interesting microsite impacts on microbial community  
261 composition. For the bacteria/archaea, *Actinobacteria*, *Firmicutes*, and *Chloroflexi* had  
262 significantly higher relative abundance in gap soils as compared to under plant canopy soils. This  
263 is likely a result of gap areas being more prone to desiccation due to the absence of the protective  
264 (shading) vegetation canopy (23). Previous research has shown that the relative abundance of  
265 *Actinobacteria* increases with decreased water availability and that desiccation stimulates  
266 ribosome synthesis for a quick head start during favorable conditions in members of this phylum  
267 (24). For example, *Actinobacteria* classes of *Rubrobacteria* and *Thermoleophilia*, known to be  
268 radio-tolerant and thermophilic, respectively, were abundant in gap soils (25, 26). Similarly,

269 *Firmicutes* are known for a resistance response strategy that maintains a stable ribosome content  
270 in both drought and wet conditions (24) and *Chloroflexi* exhibit adaptation strategies including  
271 resistance to desiccation and ultraviolet radiation (27). Taken together, adaptation to desiccation  
272 and dormancy are common strategies that facilitate microbial survival under the oligotrophic arid  
273 conditions of gap soils (28). In contrast, examples of populations found predominantly in canopy  
274 soils are the Alphaproteobacteria genera *Rhizobiales* and *Sphingomonadales*. These microbes are  
275 known for their ability to fix nitrogen and carry out phototrophy in deserts, respectively (29).

276 A strong significant microsite impact was also observed for fungal communities.  
277 Functionally, desiccation stress of gaps does not affect fungi as they are considered more resistant  
278 than bacteria and have hyphae to aid in accessing nutrients and water (30, 31). The higher fungal  
279 richness in canopy soils is likely driven by plant litter inputs in canopy soils (32). FUNGuild  
280 correctly predicted that plant pathogen, epiphyte, endophyte, and saprotrophs would be more  
281 associated with canopy soils. Ascomycetes were more abundant under plant canopies and these  
282 fungi play an important role in carbon and nitrogen cycling as they break down complex molecules,  
283 such as cellulose and lignin (33). Of the *Ascomycota*, the fungal classes enriched in canopy soils  
284 were Dothideomycetes, Eurotiomycetes, and Sordariomycetes. The characterized members of  
285 Sordariomycetes are involved in decomposition and nutrient cycling and act as plant and animal  
286 pathogens, endophytes, and saprobes (34, 35). Further, members of the order Sordariales  
287 (Sordariomycetes) and Agaricales (Agaricomycetes) are potent lignin degraders (36, 37).

288 Based on FUNGuild results, arbuscular mycorrhizae are more abundant in gap soils than  
289 under canopy soils. This is surprising, as arbuscular mycorrhizae are typically found in association  
290 with plant roots. It is noted that arbuscular mycorrhizae belonging to *Glomeromycetes* were also  
291 found to have higher abundance in gap rather canopy soils. Some species of this taxa are found

292 associated both with plants and free-living (38). Alternatively, an explanation for this observed  
293 pattern is the fungal loop hypothesis where fungi act as biological networks for transforming and  
294 translocating nutrient resources across barren spaces between plants (39).

295 In summary, microsite was a strong driver of microbial community composition for both  
296 bacteria/archaea and fungi. Possible explanations for this difference include: 1) the plant fertility  
297 island effect under oligotrophic conditions provides enhanced nutrient availability that supports a  
298 plant rhizosphere microbial community that is quite different from gaps (40–42), and 2) the plant-  
299 protective effect provides a safe haven for canopy soil microbial communities by buffering  
300 extreme temperatures, providing UV protection, and preventing desiccation by maintaining soil  
301 moisture (43).

### 302 **3.2 Gap Soils are not a Complete Reservoir for the Plant Canopy Soil Microbiome**

303 Studies of global drylands have demonstrated that increased aridity reduces microbial diversity (9,  
304 10). Thus, it is logical to hypothesize that without the protection of plants, gap soil microbial  
305 diversity may be more susceptible to aridity stress. Berdugo and colleagues described the response  
306 of drylands to increasing aridity as a three-step sequential process, initiated by a “vegetation  
307 decline” followed by a phase of “soil disruption”, and finally resulting in a “systemic breakdown”  
308 along with a reduction in relative abundance of saprotrophic and ectomycorrhizal fungi (5). This  
309 scenario raises two related questions: 1) does gap microbial diversity function as an ecosystem  
310 reservoir from which plants recruit microbes for plant-rhizosphere soils, ? (44) and 2) if the gap  
311 does not provide a plant microbiome reservoir, will aridity-induced gap expansion leave vegetation  
312 patches as the only reservoirs of critical microbial capacity?.

313 Our results demonstrate that the core microbiome of the plant root-zone was comprised of  
314 numerous taxa that were sourced from surrounding bulk (gap) soils; however, there was a

315 recognizable number of unique plant canopy soil microbial taxa that were not shared with the  
316 associated gap microbial communities. Importantly, the relative abundance of unique taxa in  
317 canopy soils increased with aridity; sites 3 & 4 had higher relative abundance of unique canopy  
318 soil microbial taxa than sites 1 & 2. These patterns suggest that gap soils do not provide a complete  
319 microbial reservoir for the plant root-zone microbiome and the source of the unique taxa in the  
320 canopy soil microbiome is unknown. For this reason, we suggest a potential loss of this plant-  
321 associated haven of microbial diversity as a result of decreasing plant density or death (45) in the  
322 scenario of increasing aridity (4). In this case, the ecosystem would lose a reservoir of microbial  
323 diversity that may be critical to plant survival in arid ecosystems (45). Therefore, the canopy-gap  
324 dichotomy documented in our study suggests that plant death could lead to ecosystem loss of  
325 critical microbial genetic potential, possibly causing the “soil disruption” and “systemic  
326 breakdown” described by Bergudo et al. (5).

### 327 **3.3 Organic Nitrogen is the Major Nitrogen Pool in the Sonoran Desert**

328 The carbon, nitrogen, and phosphorus levels found in this study are similar to that observed in  
329 many desert ecosystems that exhibit fertility island effects (3, 32, 46). TOC was 3.1 to 13.7 times  
330 higher in canopy soils than in gap soils across the geolocations studied. In desert ecosystems,  
331 organic carbon is an important nutrient source and a good indicator of the plant-associated fertility.  
332 Organic carbon supports heterotrophic microbes that carry out depolymerization and  
333 mineralization processes needed to increase N availability (47, 48). Multi-model analyses  
334 performed by Delgado-Baquerizo et al. (48) to characterize N availability found that aridity is the  
335 primary driver for N availability in gaps, but organic carbon is as significant as aridity for N  
336 availability under plant canopies. In another study by Delgado-Baquerizo et al., their model  
337 explained that increasing aridity will decrease N and C concentrations and increase phosphorus

338 (P) concentrations, hence decoupling C, N, and P cycles (49). Such reduced availability of C and  
339 N is predicted to disrupt the stoichiometric balance between C, N, and P, resulting in reduced plant  
340 productivity and microbial diversity and activity. Results from our study showed that inorganic N  
341 content in desert soils is relatively low and TOC strongly correlates with TN/DN. Therefore, we  
342 contend that organic N is the major N pool in the Sonoran Desert soils. Given that organic N is  
343 relevant source of N in this ecosystem, predicted changes in C and N availability due to climatic  
344 changes could disrupt the stoichiometric nutrient balance and subsequently lead to loss of  
345 ecosystem services.

#### 346 **3.4 Nitrogen Availability is Indicated by *ureC* in Sonoran Desert Soils**

347 The bacterial *ureC* gene abundance was significantly greater in canopy soils than gap soils;  
348 however, little is known about *ureC* gene abundance in natural ecosystems. Most studies on *ureC*  
349 gene abundance have been conducted in agricultural settings. The *ureC* abundance values detected  
350 in two unfertilized control agricultural soils, an organic farm soil, a Black soil, and vegetated  
351 sandstone and siltstone surface mine rock fragments were  $4.5\text{-}9.5 \times 10^7$ ,  $1\text{-}1.8 \times 10^8$ ,  $4.8\text{-}8.4 \times 10^7$ ,  
352 and  $0.68\text{-}3.4 \times 10^5$  gene copies  $\text{g}^{-1}$  soil, respectively (50–54). Given that the *ureC* gene abundance  
353 in the Sonoran Desert ecosystem is just slightly lower ( $0.35\text{-}2.87 \times 10^7$  gene copies  $\text{g}^{-1}$  soil) than  
354 unamended agricultural and Black soils, it suggests that canopy and gap soils have substantial N-  
355 mineralization genetic potential among ureolytic bacteria. Additionally, in a recently published  
356 organic-inorganic co-composting study, *ureC* gene was validated as a marker for ammonification  
357 (i.e. conversion of urea to ammonium) (55). Thus, our study supports the use of *ureC* as a marker  
358 for N-ammonification/mineralization in the desert ecosystems. Urease enzyme activity  
359 measurements, like *ureC* gene abundance, were significantly more abundant in canopy soils  
360 relative to gap locations indicating a strong link between genetic potential and functional capacity.

361 Additionally, *ureC* gene abundance correlated strongly with TN and DN. These correlations  
362 support the contention that mineralization of organic matter is an abundant and relevant source of  
363 N in this arid ecosystem.

### 364 **3.5 Phylogenetic Associations of *ureC* in the Sonoran Desert Soils**

365 Like *ureC* gene abundance, the diversity of ureolytic bacteria have been primarily studied in  
366 agricultural soils. *Proteobacteria* are the most abundant ureolytic bacteria in urea-amended or non-  
367 amended agricultural soils followed by *Actinobacteria*, *Verrucomicrobia*, and *Nitrospirae* (50,  
368 53). Other low-abundance phyla containing *ureC* taxa include *Acidobacteria*, *Bacteroidetes*,  
369 *Cloroflexi*, *Firmicutes*, and *Planctomycetes* (50, 53). In contrast, the majority of the identifiable  
370 *ureC* sequences in our limited clone library phylogenetically associated with *Actinobacteria* (42%;  
371 Table S7), whereas only 4% of the clones were associated with *Proteobacteria*. As discussed  
372 previously, *Actinobacteria* are more abundant in desert soils than other biomes. Therefore, this  
373 novel association of critical ureolysis functional capacity with the dominant desert phylum  
374 suggests that the limited and distinct microbial diversity observed in desert soils may not imply a  
375 loss of functional capacity, but rather an important shift in the association of that functional  
376 capacity to phylogenetic groups dominant in desert ecosystems. Moreover, it is noteworthy that  
377 the *ureC* primers used in this study were designed using conserved regions within *ureC* gene from  
378 sequences of *Proteobacteria* (56). Clearly, these primers are compatible with *Actinobacteria*,  
379 indicating that the *ureC* gene sequence is conserved across taxa. Further, 51% of the *ureC* gene  
380 clones from this study represented taxa that did not have a phylogenetic affiliation in the database  
381 (Table S7). This suggests that the *ureC* phylogenetic distribution in desert soils includes many  
382 unidentified bacteria. These results confirm the limitations of current databases and contend that

383 important functional nutrient cycling capacities in desert ecosystems are associated with novel  
384 organisms that are currently poorly characterized.

385 In summary, canopy soils of the Sonoran Desert were found to have significantly higher  
386 nutrient content, greater bacterial and fungal diversity, as well as greater N-mineralization genetic  
387 potential and urease functional capacity when compared to the associated gap soils. The *ureC* gene  
388 marker for N-mineralization was not only identified as a significant indicator of functional capacity  
389 in this desert ecosystem, but the shift in association of that functional capacity to phylogenetic  
390 groups dominant in desert ecosystems demonstrated an important pattern for desert ecosystem  
391 services. Additionally, we showed that Sonoran Desert gap soil microbial community lacks  
392 sufficient microbial diversity to serve as a complete reservoir for the plant root-zone microbiome.  
393 Therefore, it is critical that future research further characterizes the low microbial diversity and  
394 functional capacity of arid-ecosystem gap soils and determines whether the diversity of these soils  
395 is necessary as a reservoir for plant recruitment to the rhizosphere community. We propose that  
396 future studies investigate the microbial genetic capacity of gap soils along a wide aridity gradient,  
397 continue to identify novel associations between phylogenetic and functional diversity, and  
398 extensively evaluate the significance of the gap soil microbiome to the sustainability of plant  
399 vegetation density in desert ecosystems.

#### 400 **4. MATERIAL AND METHODS**

##### 401 **4.1 Site Description and Sample Collection**

402 This study focused on arid ( $0.05 \leq \text{Aridity Index} < 0.2$ ) and hyper arid ( $\text{Aridity Index} < 0.05$ )  
403 regions of the Western Sonoran Desert, AZ. Samples were collected from four distinct geographic  
404 locations along a 77 km north-south transect (Fig. 1) in October 2017. Site 1 was located in the  
405 Kofa National Wildlife Refuge; site 2 was on route 60 in Pioneer, AZ; site 3 on route 95 north of

406 Quartzsite, AZ; and site 4 south of Parker, AZ (Table S1). The mean annual temperature (MAT)  
407 and mean annual precipitation (MAP) of the geolocations were compiled from the WorldClim-  
408 Global database (57) (Table S1). The aridity index (AI) was compiled from the CGIAR-CSI's  
409 Global Potential Evapotranspiration and Global Aridity Index (Table S1;  
410 [https://cgiarcsi.community/2019/01/24/global-aridity-index-and-potential-evapotranspiration](https://cgiarcsi.community/2019/01/24/global-aridity-index-and-potential-evapotranspiration-climate-database-v2/)  
411 [-climate-database-v2/](https://cgiarcsi.community/2019/01/24/global-aridity-index-and-potential-evapotranspiration-climate-database-v2/)).

412 At each geolocation, a 30 m x 20 m sampling area was delineated, and three 30 m transects  
413 were laid out in a N/S direction within the sampling area beginning with transect 1 at the NW  
414 corner (Fig. 1D). Transects 2 and 3 were started 10 m east of the first transect. Vegetation metrics  
415 including number of vegetation patch and gaps, inter-distance between each vegetation patch, and  
416 number of grasses/shrubs/trees in each patch along the three transects were recorded (Table S1).  
417 The shrub species *Larrea tridentata* (creosote bush) and *Ambrosia dumosa* (white bursage)  
418 comprised the dominant vegetation (~70%) and the remaining plants were mixed species. For soil  
419 and microbial analyses, soil samples were collected from vegetation canopy and gap soil  
420 microsites, where 1) canopy is defined as under the canopy soil of the creosote bush-white bursage  
421 vegetation and 2) gap is defined as the interplant barren spaces devoid of any vegetation or  
422 biocrusts. Six grab samples were collected along each transect, three under the shrub canopy  
423 (canopy) and three in interplant barren spaces (gap; Fig. S5). At each transect, soil was removed  
424 from a 20 cm diameter soil pit excavated to a depth of 20 cm and combined with soils from the  
425 other two canopy or gap soil-pits along the same transect. The three composite soil samples were  
426 homogenized, sieved (2 mm pore size), and subsampled for soil chemical analyses in the field.  
427 This was done separately for under shrub canopy and gap soils along each transect. Soils were  
428 sampled to a depth of 20 cm to include soils relevant to root development and to minimize the

429 impact of UV radiation on surface soil microbial communities (58). Composite samples for each  
430 microsite were collected to capture the heterogeneity of the microsites along the transect. For  
431 microbial analysis, samples were collected from each sampling pit sidewall to a depth of 20 cm  
432 using sterile instruments. Microbial samples from three pits along the transect were homogenized  
433 and combined. Thus, a single microbial composite sample was generated for canopy and gap soil  
434 microsites from each transect. These samples were transported on ice to the lab and frozen at -  
435 80°C until ready for DNA extractions. Selected soil chemistry measurements (TOC, DOC, DN,  
436 and BAP) and 16S rRNA gene amplicon analyses have been reported by Chen et al. (2020) (59).  
437 In our present study, a detailed analyses of soil physicochemical and enzymatic properties,  
438 microbial diversity and composition, and gene abundance have been described.

#### 439 **4.2 Soil Physicochemical and Enzymatic Analyses**

440 Soil pH and electric conductivity (EC) were determined from a 1:2 soil-to-distilled water (dH<sub>2</sub>O)  
441 slurry after 30 min of shaking (60). Dry soils were milled (SPEX 8000D Mixer/Mill, Metuchen,  
442 NJ) prior to analysis for TC, TN, DOC, dissolved inorganic carbon (DIC), and DN. The TC and  
443 TN were analyzed using ECS 4010 C and N analyzer (Costech Analytical Technologies, Valencia,  
444 CA). The DOC, DIC, and DN (1:15 soil: dH<sub>2</sub>O; shaken at 200 rpm for 24 h at room temperature)  
445 were quantified by TOC-L analyzer (Shimadzu Scientific Instruments, Columbia, MD). For  
446 measurement of TOC, soil samples were digested with HCl to remove dissolved carbonates  
447 followed by washing with water and drying at 70°C. The dried samples were then combusted in  
448 the presence of CuO in vacuum and the released carbon dioxide was cryogenically distilled from  
449 the mixture of combustion gases, manometrically quantified, and converted to carbon mass  
450 (University of Arizona Accelerator Mass Spectrometry Lab). The NO<sub>3</sub>-N (1:5 w/v in KCl solution)  
451 was measured using the Cd-reduction method (61). The NH<sub>4</sub>-N (1:2 w/v in KCl solution) was

452 quantified using the HACH Ammonia Salicylate Reagent and Ammonia Cyanurate Reagent  
453 (HACH, Loveland, CO; Sinsabaugh Lab protocol 2005). BAP was measured using the Olsen  
454 method (62) and soil texture was determined using the hydrometer method (63). Urease enzyme  
455 activity was measured using the Kandeler and Gerber (64) method with modifications as per  
456 Fioretto et al. (65).

### 457 **4.3 Molecular Analyses of Microbial Communities**

458 DNA extraction from 0.5 g soil samples was carried out using Fast DNA SPIN for Soil Kit™ (MP  
459 Biomedicals, Solon OH, USA) with modifications to enhance DNA recovery from oligotrophic  
460 arid soil samples. All consumables and reagents lacking biomolecules were sterilized with UV  
461 light for 30 minutes. Further, modifications include: (1) 0.8 mL of binding matrix was used to bind  
462 DNA, and after mixing and settling the binding matrix, all supernatant was removed and discarded  
463 without filtering through the spin filter; (2) binding matrix was washed 1-2x times with 0.5 mL of  
464 6 M Guanidine thiocyanate (Sigma-Aldrich, St. Louis, MO) to remove organic matter; (3) the wash  
465 with SEWS-M reagent was repeated twice to remove co-extracted salts; (4) spin filters were dried  
466 at 37°C for 10 min under a laminar flow hood prior to DNA elution; (5) a two-fold DNA elution  
467 was performed using two sequential aliquots of 50 µL of UV-sterilized DEPC treated, nuclease-  
468 free water (Growcells, Irvine, CA) preheated to 60°C in which each 50 µL aliquot was followed  
469 by incubation at 60°C for 10 min and centrifugation for 1 min. The extracted DNA was quantified  
470 using the Qubit® dsDNA High Sensitivity Assay Kit (Life Technologies, NY, USA). All the  
471 samples had quantifiable DNA ranging from 3.17 to 41.5 ng µL<sup>-1</sup>. Paired-end amplicon sequencing  
472 of DNA extracts was performed using 16S rRNA gene primers 515F/806R (bacteria/ archaea) and  
473 ITS primers ITS1f-ITS2 (fungi) as described in Walters et al. (66). The purified PCR products  
474 from all samples were pooled together in equimolar concentrations and sequenced on a 2 × 150

475 bp Illumina MiSeq platform. All sequencing runs were conducted at the Microbiome Core, Steele  
476 Children's Research Center, University of Arizona.

#### 477 **4.4 Sequence Processing**

478 Raw reads were demultiplexed using idemp tool (<https://github.com/yhwu/idemp>). Further  
479 bioinformatics analyses were carried out using DADA2 pipeline (67). The demultiplexed reads  
480 were trimmed to retain 140 bases of the forward and reverse reads. The paired-end reads were  
481 joined using the default overlap of at least 12 bases and then grouped into ASVs. The ASVs were  
482 subjected to chimera removal. Post quality filtering, a total of 4,193,573 sequence reads remained  
483 for 16S rRNA gene and 2,901,758 for ITS, with an average of  $149,770 \pm 33,143$  and  $120,907 \pm$   
484  $37,410$  sequence reads per sample for 16S rRNA gene and ITS region, respectively. Taxonomy  
485 assignments were determined with the RDP Classifier (68) using SILVA (69) and UNITE ITS (70)  
486 databases for bacterial/archaeal and fungal communities, respectively. Contaminants were  
487 removed from the ASV tables through comparisons of samples and the blanks, leaving 16,4844  
488 and 2,234 ASVs for bacterial/archaeal and fungal communities, respectively. The taxonomy tables  
489 obtained were then normalized using a cumulative-sum scaling approach (71). The taxonomic  
490 distribution of bacterial/archaeal communities were compared to other desert and non-desert  
491 biomes using published datasets (8, 10). Additionally, the FUNGuild tool was utilized to predict  
492 fungal guilds from ITS taxonomic assignments (72).

#### 493 **4.5 Bacterial Gene Abundance Using qPCR**

494 Bacterial 16S rRNA and *ureC* genes were quantified to assess the bacterial copy number and N-  
495 mineralization capacity, respectively. The V3-V4 variable region of the 16S rRNA gene was  
496 amplified using primers 338F/518R (73). The bacterial *ureC* gene, encoding a subunit of the urease  
497 enzyme, was amplified using the primers ureC1F/ureC2R (56). Conserved regions within *ureC*

498 gene from *Ralstonia eutropha* and other ureolytic proteobacteria including species from the genera  
499 *Nitrosospira* and *Nitrosococcus* were used to design degenerate primers for the *ureC* gene (56).

500 qPCR standard for 16S rRNA gene quantification was described previously (74). For the  
501 bacterial *ureC* gene, a limited clone library was constructed from the collected soil samples as  
502 described in (75) and the resulting *ureC* sequences were matched to NCBI protein database using  
503 BLASTX. The clone for the standard was selected based on 100% sequence match to the query in  
504 the NCBI database (Table S7). Plasmids were quantified using Qubit® dsDNA High Sensitivity  
505 Assay Kit. The qPCR reactions were performed in triplicate using a 96-well plate on a CFX96  
506 Real-Time Detection System (Bio-Rad Laboratories, Hercules, CA). In a 10 µL reaction, 5 µL  
507 SsoFast™ EvaGreen® Supermix (Bio-Rad Laboratories, Hercules, CA), either 0.2 µM (*ureC*) or  
508 0.3 µM (16S rRNA gene) primers, and 1 µL of sample DNA extract or plasmid standards were  
509 used.

510 All qPCR standards and samples were run in triplicate. Amplification conditions for 16S  
511 rRNA gene were as described previously (74) with an amplification efficiency of 97-100%,  $R^2$  of  
512 0.99, and a calibration curve with  $10^3$  to  $10^8$  copies of 16S rRNA gene. The amplification  
513 conditions for *ureC* gene were 98°C for 3 min, 50 cycles of 98°C for 20s, 60°C for 25s with an  
514 amplification efficiency greater than 91%, an  $R^2$  of 0.98-0.99 and a calibration curve of  $10^2$  to  $10^6$   
515 gene copies. Melt curves were generated at the end of each run with a temperature range of 65 to  
516 95°C in 0.5°C increments. Amplicon specificity was confirmed by melt curve analysis and gel  
517 electrophoresis. Inter-plate calibration was performed to normalize for inter-plate variations in  
518 amplification efficiencies among plates being assayed for the same gene (CFX Manager™  
519 software).

## 520 **4.6 Statistical Analyses**

521 Statistical analyses were implemented in R (R Core Team, 2019). Significant differences of soil  
522 physicochemical properties, microbial richness, and gene abundance between canopy and gap soils  
523 were determined using linear mixed-effects models using the *lme4* and *lmerTest* packages.  
524 Microsite and geolocation were used as fixed and random effects, respectively. The Satterthwaite's  
525 (76) approximations were used to estimate the statistical significance of the fixed effect. R<sup>2</sup>-value  
526 of the fixed effect was assessed as described in Nakagawa & Schielzeth (77). Community  
527 dissimilarity was calculated using the Bray-Curtis distance and NMDS ordination was used for  
528 visualization. Microbial community differences among geolocations and microsites were  
529 examined using PERMANOVA (78). Further, community dispersion within microsite was  
530 (multivariate dispersion) was calculated. FUNGuild differences were also evaluated using negative  
531 binomial generalized linear mixed-effects models. Additional statistical tests are described in  
532 Supplemental methods. The R scripts used for the statistical analyses are available on GitHub:  
533 [https://github.com/kush-priyanka/Arid\\_Soil\\_Microbiome](https://github.com/kush-priyanka/Arid_Soil_Microbiome).

## 534 **4.7 Data Availability**

535 Raw sequence data for 16S rRNA gene and ITS were submitted to NCBI's Sequence Read Archive  
536 under the accession number PRJNA610058.

## 537 **5. ACKNOWLEDGEMENTS**

538 The research was supported by the National Institute of Environmental Health Sciences grant  
539 P42ES004940 and University of Arizona's Research, Discovery, and Innovation Accelerate for  
540 Success Grant 2018-2019. We thank Mary Kay Amistadi and Rachel Neville of the Arizona  
541 Laboratory for Emerging Contaminants for performing DOC, DIC, DN analyses. We are also  
542 thankful to Dr. Gregory Hodgins at the Accelerator Mass Spectrometry lab for TOC

543 measurements. Many thanks to Dr. Daniel Laubitz at Microbiome Core, University of Arizona  
544 Steele Children’s Research Center for providing quality control and technical support for amplicon  
545 sequencing. In addition, we acknowledge Karen Serrano, Lia Ossanna, Emma Ison, and Stephanie  
546 Honeker for performing pH and EC; DNA extractions, Bioavailable Phosphorus; and Soil Texture  
547 analyses, respectively.

548

549

550

551

552

553

554

555

556

557

558

559

560

561

562

563

564

565 **6. REFERENCES**

- 566 1. Aguiar MR, Sala OE. 1999. Patch structure, dynamics and implications for the functioning  
567 of arid ecosystems. *Trends Ecol Evol* 14:273–277.
- 568 2. Schlesinger WH, Raikks JA, Hartley AE, Cross AF. 1996. On the spatial pattern of soil  
569 nutrients in desert ecosystems. *Ecology* 77:364–374.
- 570 3. de Graaff M-A, Throop HL, Verburg PSJ, Iii JAA, Campos X. 2014. A synthesis of  
571 climate and vegetation cover effects on biogeochemical cycling in shrub-dominated  
572 drylands. *Ecosystems* 17:931–945.
- 573 4. Seager R, Ting M, Held I, Kushnir Y, Lu J, Vecchi G, Huang HP, Harnik N, Leetmaa A,  
574 Lau NC, Li C, Velez J, Naik N. 2007. Model projections of an imminent transition to a  
575 more arid climate in southwestern North America. *Science* (80- ) 316:1181–1184.
- 576 5. Berdugo M, Delgado-Baquerizo M, Soliveres S, Hernández-Clemente R, Zhao Y, Gaitán  
577 JJ, Gross N, Saiz H, Maire V, Lehman A, Rillig MC, Solé R V., Maestre FT. 2020. Global  
578 ecosystem thresholds driven by aridity. *Science* (80- ) 367:787–790.
- 579 6. Sun GQ, Wang CH, Chang LL, Wu YP, Li L, Jin Z. 2018. Effects of feedback regulation  
580 on vegetation patterns in semi-arid environments. *Appl Math Model* 61:200–215.
- 581 7. Kéfi S, Rietkerk M, Alados CL, Pueyo Y, Papanastasis VP, ElAich A, De Ruiter PC.  
582 2007. Spatial vegetation patterns and imminent desertification in Mediterranean arid  
583 ecosystems. *Nature* 449:213–217.
- 584 8. Fierer N, Leff JW, Adams BJ, Nielsen UN, Bates ST, Lauber CL, Owens S, Gilbert JA,  
585 Wall DH, Caporaso JG. 2012. Cross-biome metagenomic analyses of soil microbial  
586 communities and their functional attributes. *PNAS* 109:21390–21395.
- 587 9. Maestre FT, Delgado-Baquerizo M, Jeffries TC, Eldridge DJ, Ochoa V, Gozalo B, Quero

- 588 JL, García-Gómez M, Gallardo A, Ulrich W, Bowker MA, Arredondo T, Barraza-Zepeda  
589 C, Bran D, Florentino A, Gaitán J, Gutiérrez JR, Huber-Sannwald E, Jankju M, Mau RL,  
590 Miriti M, Naseri K, Ospina A, Stavi I, Wang D, Woods NN, Yuan X, Zaady E, Singh BK.  
591 2015. Increasing aridity reduces soil microbial diversity and abundance in global drylands.  
592 PNAS 112:15684–15689.
- 593 10. Neilson JW, Califf K, Cardona C, Copeland A, van Treuren W, Josephson KL, Knight R,  
594 Gilbert JA, Quade J, Caporaso JG, Maier RM. 2017. Significant impacts of increasing  
595 aridity on the arid soil microbiome. *mSystems* 2:e00195-16.
- 596 11. Bachar A, Ines M, Soares M, Gillor O. 2012. The effect of resource islands on abundance  
597 and diversity of bacteria in arid soils. *Microb Ecol* 63:694–700.
- 598 12. Ben-David EA, Zaady E, Sher Y, Nejdat A. 2011. Assessment of the spatial distribution  
599 of soil microbial communities in patchy arid and semi-arid landscapes of the Negev  
600 Desert using combined PLFA and DGGE analyses. *FEMS Microbiol Ecol* 76:492–503.
- 601 13. Hortal S, Bastida F, Armas C, Lozano YM, Moreno JL, García C, Pugnaire FI. 2013. Soil  
602 microbial community under a nurse-plant species changes in composition, biomass and  
603 activity as the nurse grows. *Soil Biol Biochem* 64:139–146.
- 604 14. Garcia DE, Lopez BR, de-Bashan LE, Hirsch AM, Maymon M, Bashan Y. 2018.  
605 Functional metabolic diversity of the bacterial community in undisturbed resource island  
606 soils in the southern Sonoran Desert. *L Degrad Dev* 29:1467–1477.
- 607 15. Nielsen UN, Ball BA. 2015. Impacts of altered precipitation regimes on soil communities  
608 and biogeochemistry in arid and semi-arid ecosystems. *Glob Chang Biol* 21:1407–1421.
- 609 16. De Neve S. 2017. Organic matter mineralization as a source of nitrogen, p. 65–83. *In*  
610 *Advances in research on fertilization management of vegetable crops*. Springer.

- 611 17. Jan MT, Roberts P, Tonheim SK, Jones DL. 2009. Protein breakdown represents a major  
612 bottleneck in nitrogen cycling in grassland soils. *Soil Biol Biochem* 41:2272–2282.
- 613 18. Gresham TLT, Sheridan PP, Watwood ME, Fujita Y, Colwell FS. 2007. Validation of  
614 ureC-based primers for groundwater detection of urea-hydrolyzing bacteria. *Geomicrobiol*  
615 *J* 24:353–364.
- 616 19. Baraniya D, Puglisi E, Ceccherini MT, Pietramellara G, Giagnoni L, Arenella M,  
617 Nannipieri P, Renella G. 2016. Protease encoding microbial communities and protease  
618 activity of the rhizosphere and bulk soils of two maize lines with different N uptake  
619 efficiency. *Soil Biol Biochem* 96:176–179.
- 620 20. Butterfield BJ, Betancourt JL, Turner RM, Briggs JM. 2010. Facilitation drives 65 years  
621 of vegetation change in the Sonoran Desert. *Ecology* 91:1132–1139.
- 622 21. Mosqueira MJ, Marasco R, Fusi M, Michoud G, Merlino G, Cherif A, Daffonchio D.  
623 2019. Consistent bacterial selection by date palm root system across heterogeneous desert  
624 oasis agroecosystems. *Sci Rep* 9:1–12.
- 625 22. Marasco R, Mosqueira MJ, Fusi M, Ramond J-B, Merlino G, Booth JM, Maggs-Kölling  
626 G, Cowan DA, Daffonchio D. 2018. Rhizosheath microbial community assembly of  
627 sympatric desert speargrasses is independent of the plant host. *Microbiome* 6:215.
- 628 23. Butterfield BJ, Bradford JB, Armas C, Prieto I, Pugnaire FI. 2016. Does the stress-  
629 gradient hypothesis hold water? Disentangling spatial and temporal variation in plant  
630 effects on soil moisture in dryland systems. *Funct Ecol* 30:10–19.
- 631 24. Barnard RL, Osborne CA, Firestone MK. 2013. Responses of soil bacterial and fungal  
632 communities to extreme desiccation and rewetting. *ISME J* 7:2229–2241.
- 633 25. Chanal A, Chapon V, Benzerara K, Barakat M, Christen R, Achouak W, Barras F, Heulin

- 634 T. 2006. The desert of Tataouine: An extreme environment that hosts a wide diversity of  
635 microorganisms and radiotolerant bacteria. *Environ Microbiol* 8:514–525.
- 636 26. Hu D, Zang Y, Mao Y, Gao B. 2019. Identification of molecular markers that are specific  
637 to the class thermoleophilia. *Front Microbiol* 10:1–13.
- 638 27. Battistuzzi FU, Hedges SB. 2009. A major clade of prokaryotes with ancient adaptations  
639 to life on land. *Mol Biol Evol* 26:335–343.
- 640 28. Leung PM, Bay SK, Meier D V., Chiri E, Cowan DA, Gillor O, Woebken D, Greening C.  
641 2020. Energetic basis of microbial growth and persistence in desert ecosystems. *mSystems*  
642 5:e00495-19.
- 643 29. Van Goethem MW, Makhalanyane TP, Cowan DA, Valverde A. 2017. Cyanobacteria and  
644 Alphaproteobacteria may facilitate cooperative interactions in niche communities. *Front*  
645 *Microbiol* 8:1–11.
- 646 30. Gordon H, Haygarth PM, Bardgett RD. 2008. Drying and rewetting effects on soil  
647 microbial community composition and nutrient leaching. *Soil Biol Biochem* 40:302–311.
- 648 31. de Vries FT, Liiri ME, Bjørnlund L, Bowker MA, Christensen S, Setälä HM, Bardgett  
649 RD. 2012. Land use alters the resistance and resilience of soil food webs to drought. *Nat*  
650 *Clim Chang* 2:276–280.
- 651 32. Ochoa-Hueso R, Eldridge DJ, Delgado-Baquerizo M, Soliveres S, Bowker MA, Gross N,  
652 Le Bagousse-Pinguet Y, Quero JL, García-Gómez M, Valencia E, Arredondo T,  
653 Beinticincio L, Bran D, Cea A, Coaguila D, Dougill AJ, Espinosa CI, Gaitán J, Guuroh  
654 RT, Guzman E, Gutiérrez JR, Hernández RM, Huber-Sannwald E, Jeffries T, Linstädter  
655 A, Mau RL, Moneris J, Prina A, Pucheta E, Stavi I, Thomas AD, Zaady E, Singh BK,  
656 Maestre FT. 2018. Soil fungal abundance and plant functional traits drive fertile island

- 657 formation in global drylands. *J Ecol* 106:242–253.
- 658 33. Zeng Q, Liu Y, Xiao L, An S. 2020. Climate and soil properties regulate soil fungal  
659 communities on the Loess Plateau. *Pedobiologia (Jena)* 81–82:150668.
- 660 34. Suleiman MK, Dixon K, Commander L, Nevill P, Quoreshi AM, Bhat NR, Manuvel AJ,  
661 Sivadasan MT. 2019. Assessment of the diversity of fungal community composition  
662 associated with *Vachellia pachyceras* and its rhizosphere soil from Kuwait Desert. *Front*  
663 *Microbiol* 10:63.
- 664 35. Qin H, Wang H, Strong PJ, Li Y, Xu Q, Wu Q. 2014. Rapid soil fungal community  
665 response to intensive management in a bamboo forest developed from rice paddies. *Soil*  
666 *Biol Biochem* 68:177–184.
- 667 36. Pöggeler S. 2011. Evolution of multicopper oxidase genes in Coprophilous and Non-  
668 Coprophilous members of the order Sordariales. *Curr Genomics* 12:95–103.
- 669 37. Entwistle EM, Zak DR, Edwards IP. 2013. Long-term experimental nitrogen deposition  
670 alters the composition of the active fungal community in the forest floor. *Soil Sci Soc Am*  
671 *J* 77:1648–1658.
- 672 38. Hempel S, Renker C, Buscot F. 2007. Differences in the species composition of  
673 arbuscular mycorrhizal fungi in spore, root and soil communities in a grassland  
674 ecosystem. *Environ Microbiol* 9:1930–1938.
- 675 39. Collins SL, Sinsabaugh RL, Crenshaw C, Green L, Porras-Alfaro A, Stursova M, Zeglin  
676 LH. 2008. Pulse dynamics and microbial processes in aridland ecosystems. *J Ecol* 96:413–  
677 420.
- 678 40. Pointing SB, Belnap J. 2012. Microbial colonization and controls in dryland systems. *Nat*  
679 *Rev Microbiol* 10:551–562.

- 680 41. Mueller RC, Belnap J, Kuske CR. 2015. Soil bacterial and fungal community responses to  
681 nitrogen addition across soil depth and microhabitat in an arid shrubland. *Front Microbiol*  
682 6:891.
- 683 42. Steven B, Gallegos-Graves LV, Yeager C, Belnap J, Kuske CR. 2014. Common and  
684 distinguishing features of the bacterial and fungal communities in biological soil crusts  
685 and shrub root zone soils 69:302–312.
- 686 43. Cortina J, Maestre FT. 2005. Plant effects on soils in drylands: implications for  
687 community dynamics and ecosystem restoration. Kluwer Academic Publishers, Dordrecht,  
688 the Netherlands.
- 689 44. Yeoh YK, Dennis PG, Paungfoo-Lonhienne C, Weber L, Brackin R, Ragan MA, Schmidt  
690 S, Hugenholtz P. 2017. Evolutionary conservation of a core root microbiome across plant  
691 phyla along a tropical soil chronosequence. *Nat Commun* 8:1–9.
- 692 45. Honeker LK, Gullo CF, Neilson JW, Chorover J, Maier RM. 2019. Effect of re-  
693 acidification on buffalo grass rhizosphere and bulk microbial communities during  
694 phytostabilization of metalliferous mine tailings. *Front Microbiol* 10:1209.
- 695 46. Butterfield BJ, Briggs JM. 2009. Patch dynamics of soil biotic feedbacks in the Sonoran  
696 Desert. *J Arid Env* 73:96–102.
- 697 47. Cookson WR, Müller C, O’Brien PA, Murphy D V., Grierson PF. 2006. Nitrogen  
698 dynamics in an Australian semiarid grassland soil. *Ecology* 87:2047–2057.
- 699 48. Delgado-Baquerizo M, Maestre FT, Gallardo A, Quero JL, Ochoa V. 2013. Aridity  
700 modulates N availability in arid and semiarid mediterranean grasslands. *PLoS One*  
701 8:59807.
- 702 49. Delgado-Baquerizo M, Maestre FT, Gallardo A, Bowker MA, Wallenstein MD, Luis

703 Quero J, Ochoa V, Gozalo B, García-Gómez M, Soliveres S, García-Palacios P, Berdugo  
704 M, Valencia E, Escolar C, Arredondo T, Barraza-Zepeda C, Bran D, Carreira JA, Chaieb  
705 M, Conceição AA, Derak M, Eldridge DJ, Escudero A, Espinosa CI, Gaitán J, Gatica MG,  
706 Gómez-González S, Guzman E, Gutiérrez JR, Florentino A, Hepper E, Hernández RM,  
707 Huber-Sannwald E, Jankju M, Liu J, Mau RL, Miriti M, Moneris J, Naseri K, Noumi Z,  
708 Polo V, Prina A, Pucheta E, Ramírez E, Ramírez-Collantes DA, Romão R, Tighe M,  
709 Torres D, Torres-Díaz C, Ungar ED, Val J, Wamiti W, Wang D, Zaady E. 2013.  
710 Decoupling of soil nutrient cycles as a function of aridity in global drylands. *Nature*  
711 502:672–676.

712 50. Sun R, Li W, Hu C, Liu B. 2019. Long-term urea fertilization alters the composition and  
713 increases the abundance of soil ureolytic bacterial communities in an upland soil. *FEMS*  
714 *Microbiol Ecol* 95:44.

715 51. Ouyang Y, Reeve JR, Norton JM. 2018. Soil enzyme activities and abundance of  
716 microbial functional genes involved in nitrogen transformations in an organic farming  
717 system. *Biol Fertil Soils* 54:437–450.

718 52. Ouyang Y, Norton JM. 2020. Short-term nitrogen fertilization affects microbial  
719 community composition and nitrogen mineralization functions in an agricultural soil. *Appl*  
720 *Environ Microbiol* 86.

721 53. Wang L, Luo X, Liao H, Chen W, Wei D, Cai P, Huang Q. 2018. Ureolytic microbial  
722 community is modulated by fertilization regimes and particle-size fractions in a Black soil  
723 of Northeastern China. *Soil Biol Biochem* 116:171–178.

724 54. Yarwood S, Wick A, Williams M, Daniels WL. 2015. Parent material and vegetation  
725 influence soil microbial community structure following 30-Years of rock weathering and

- 726 pedogenesis. *Microb Ecol* 69:383–394.
- 727 55. Yu H, Xie B, Khan R, Yan H, Shen G. 2020. The changes in functional marker genes  
728 associated with nitrogen biological transformation during organic-inorganic co-  
729 composting. *Bioresour Technol* 295.
- 730 56. Koper TE, El-Sheikh AF, Norton JM, Klotz MG. 2004. Urease-encoding genes in  
731 ammonia-oxidizing bacteria. *Appl Env Microbiol* 70:2342–2348.
- 732 57. Fick SE, Hijmans RJ. 2017. WorldClim 2: new 1-km spatial resolution climate surfaces  
733 for global land areas. *Int J Clim* 37:4302–4315.
- 734 58. Maier RM, Drees KP, Neilson JW, Henderson DA, Quade J, Betancourt JL, Navarro-  
735 Gonzalez R, Rainey FA, McKay CP. 2004. Microbial life in the Atacama Desert. *Science*  
736 (80- ) 306:1289–1291.
- 737 59. Chen Y, Neilson JW, Kushwaha P, Maier RM, Barberán A. 2020. Life-history strategies  
738 of soil microbial communities in an arid ecosystem. *ISME J* 1–9.
- 739 60. Honeker LK, Neilson JW, Root RA, Gil-Loaiza J, Chorover J, Maier RM. 2017. Bacterial  
740 rhizoplane colonization patterns of *Buchloe dactyloides* growing in metalliferous mine  
741 tailings reflect plant status and biogeochemical conditions HHS Public Access. *Microb*  
742 *Ecol* 74:853–867.
- 743 61. Dorich RA, Nelson DW. 1984. Evaluation of manual cadmium reduction methods for  
744 determination of nitrate in potassium chloride extracts of soils. *Soil Sci Soc Am J* 48:72–  
745 75.
- 746 62. Olsen S, Cole C, Watanabe E, Dean L. 1954. Estimation of available phosphorus in soils  
747 by extraction with sodium bicarbonate U.S. Gov. Print Office. Washington, DC.
- 748 63. Gee GW, Bauder JW. 1979. Particle size analysis by hydrometer: A simplified method for

- 749 routine textural analysis and a sensitivity test of measurement parameters. *Soil Sci Soc*  
750 *Am J* 43:1004–1007.
- 751 64. Kandeler E, Gerber H. 1988. Short-term assay of soil urease activity using colorimetric  
752 determination of ammonium. *Biol Fert Soils* 6:68–72.
- 753 65. Fioretto A, Papa S, Pellegrino A, Ferrigno A. 2009. Microbial activities in soils of a  
754 Mediterranean ecosystem in different successional stages. *Soil Biol Biochem* 41:2061–  
755 2068.
- 756 66. Walters W, Hyde ER, Berg-Lyons D, Ackermann G, Humphrey G, Parada A, Gilbert JA,  
757 Jansson JK, Gregory Caporaso J, Fuhrman JA, Apprill A, Knight R, Walters CW. 2016.  
758 Improved bacterial 16S rRNA gene (V4 and V4-5) and fungal internal transcribed spacer  
759 marker gene primers for microbial community surveys. *mSystems* 1:e00009-15.
- 760 67. Callahan BJ, Mcmurdie PJ, Rosen MJ, Han AW, Johnson AJA, Holmes SP. 2016.  
761 DADA2: high-resolution sample inference from illumina amplicon data. *Nat Methods*  
762 13:581–583.
- 763 68. Wang Q, Garrity GM, Tiedje JM, Cole JR. 2007. Naive bayesian classifier for rapid  
764 assignment of rRNA sequences into the new bacterial taxonomy. *Appl Env Microbiol*  
765 73:5261–5267.
- 766 69. Quast C, Pruesse E, Yilmaz P, Gerken J, Schweer T, Yarza P, Rg Peplies J, Glö Ckner  
767 FO. 2013. The SILVA ribosomal RNA gene database project: improved data processing  
768 and web-based tools. *Nucleic Acids Res* 41:D590–D596.
- 769 70. Nilsson RH, Larsson K-H, Taylor AFS, Bengtsson-Palme J, Jeppesen TS, Schigel D,  
770 Kennedy P, Picard K, Gi Ockner 10 FO, Tedersoo L, Saar I, Abarenkov K. 2018. The  
771 UNITE database for molecular identification of fungi: handling dark taxa and parallel

- 772 taxonomic classifications. *Nucleic Acid Res* 47:D259–D264.
- 773 71. Paulson JN, Pop M, Bravo HC. 2013. metagenomeSeq: Statistical analysis for sparse  
774 high-throughput sequencing. *Bioconductor Packag* 1:191.
- 775 72. Nguyen NH, Song Z, Bates ST, Branco S, Tedersoo L, Menke J, Schilling JS, Kennedy  
776 PG. 2016. FUNGuild: An open annotation tool for parsing fungal community datasets by  
777 ecological guild. *Fungal Ecol* 20:241–248.
- 778 73. Einen J, Thorseth IH, Øvreås L, Øvreås Ø. 2008. Enumeration of Archaea and Bacteria in  
779 seafloor basalt using real-time quantitative PCR and fluorescence microscopy. *FEMS*  
780 *Microbiol Lett* 282:182–187.
- 781 74. Ortiz M, Legatzki A, Neilson JW, Fryslie B, Nelson WM, Wing RA, Soderlund CA, Pryor  
782 BM, Maier RM. 2013. Making a living while starving in the dark: metagenomic insights  
783 into the energy dynamics of a carbonate cave. *ISME J* 8:478–491.
- 784 75. Delgado-Baquerizo M, Maestre FT, Eldridge DJ, Singh BK. 2016. Microsite  
785 differentiation drives the abundance of soil ammonia oxidizing bacteria along aridity  
786 gradients. *Front Microbiol* 7:505.
- 787 76. Satterthwaite FE. 1941. Synthesis of variance. *Psychometrika* 6:309–316.
- 788 77. Nakagawa S, Schielzeth H. 2013. A general and simple method for obtaining R<sup>2</sup> from  
789 generalized linear mixed-effects models. *Methods Ecol Evol* 4:133–142.
- 790 78. Anderson MJ. 2001. A new method for non-parametric multivariate analysis of variance.  
791 *Austral Ecol* 26:32–46.

792

793 **TABLE 1: Abiotic and biotic differences between canopy and gap soil microsites.** Results of  
794 linear mixed-effects models are shown. Microsite and geolocation were used as fixed and random  
795 effects, respectively. A positive number in the canopy and gap differences column (difference in  
796 average canopy and gap values) represents a higher value in canopy soils and a negative number  
797 denotes higher value in gap soils. Significant *p-values* are represented as \*  $p < 0.05$ ; \*\*  $p < 0.01$ ;  
798 \*\*\*  $p < 0.001$ , and ns, not significant. The marginal  $R^2$  represents only the variance of the microsite.  
799 The conditional  $R^2$  takes both the microsite and geolocation into account. The variables are ranked  
800 based on their decreasing marginal  $R^2$  values.

Abiotic variables <sup>a</sup>	Canopy and Gap differences	Significance	Marginal $R^2$	Conditional $R^2$
DN ( $\mu\text{g g}^{-1}$ )	20.80	***	0.51	0.65
DIC ( $\mu\text{g g}^{-1}$ )	24	***	0.50	0.73
Ammonium ( $\mu\text{g g}^{-1}$ )	1.7	***	0.46	0.60
TN ( $\mu\text{g g}^{-1}$ )	154.3	***	0.43	0.74
BAP ( $\mu\text{g g}^{-1}$ )	3.5	***	0.42	0.43
DOC ( $\mu\text{g g}^{-1}$ )	94.2	***	0.31	0.84
Nitrate ( $\mu\text{g g}^{-1}$ )	10.8	**	0.30	0.41
EC ( $\mu\text{s cm}^{-1}$ )	153	**	0.26	0.58
TOC (%)	0.24	*	0.22	0.22
TC ( $\mu\text{g g}^{-1}$ )	1372.5	**	0.14	0.75
pH	-0.15	ns	0.07	0.49
Sand (%)	-1.4	ns	0.01	0.82
Silt (%)	1.6	ns	0.01	0.78
Clay (%)	-0.13	ns	0.00	0.18
Biotic variables	Canopy and Gap differences	Significance	Marginal $R^2$	Conditional $R^2$
ITS Richness	131.7	***	0.58	0.63
16S Richness	261.3	*	0.20	0.21
<i>ureC</i> gene abundance (copy # $\text{g}^{-1}$ soil)	0.23	***	0.38	0.49
Urease enzyme activity ( $\mu\text{g NH}_4\text{-N g}^{-1} \text{h}^{-1}$ )	5.4	**	0.31	0.49
16S rRNA gene abundance (copy # $\text{g}^{-1}$ soil)	0.18	**	0.18	0.56
Fungal Functional Guild				
Epiphyte	0.0001	***	0.89	0.98
Endophyte	0.02	***	0.71	0.71

<b>Plant Pathogen</b>	0.09	***	0.68	0.73
<b>Wood Saprotroph</b>	0.01	**	0.57	0.57
<b>Ericoid Mycorrhizal fungi</b>	0.002	*	0.44	0.44
<b>Fungal Parasite</b>	0.001	**	0.44	0.44
<b>Ectomycorrhizal fungi</b>	0.003	*	0.40	0.40
<b>Arbuscular Mycorrhizal fungi</b>	-0.02	*	0.38	0.40
<b>Soil Saprotroph</b>	0.001	-	0.05	0.05

<sup>a</sup>DN=dissolved nitrogen, DIC=dissolved inorganic carbon, TN=total nitrogen, BAP=bioavailable phosphorus, DOC=dissolved organic carbon, EC=electrical conductivity, TOC=total organic carbon, TC=total carbon

801

802 **FIG 1: Sampling design in this study.** A) map of the Sonoran, Mojave, and Chihuahuan Deserts  
803 in North America; B) map of the Sonoran Desert sampling area in the state of Arizona; C) aridity  
804 index map showing the soil sampling sites; and D) field sampling layout. Along each transect, a  
805 composite sample of canopy (C) and gap (G) soils was collected by combining three canopy and  
806 three gap samples.

807  
808 **FIG 2: Richness and non-metric dimensional scaling (NMDS) ordination plots for the**  
809 **microbial community.** A) observed number of bacterial/archaeal amplicon sequence variants  
810 (ASV) or phylotypes; B) ordination plot for bacterial/archaeal community composition; C)  
811 observed number of fungal ASVs (or phylotypes); and D) ordination plot for fungal community  
812 composition. Richness is depicted as the mean value and standard deviation.

813  
814 **FIG 3: Unique and shared ASVs across the canopy and gap soil microsites.** A) venn-diagram  
815 represents unique and shared bacterial/archaeal amplicon sequence variants (ASVs) in the  
816 microsites; B) number of unique and shared bacterial/archaeal ASVs in the microsites at site level;  
817 C) relative abundance of canopy soil bacterial/archaeal ASVs; D) venn-diagram represents unique  
818 and shared fungal ASVs in the microsites; E) number of unique and shared fungal ASVs in the  
819 microsites at site level; and F) relative abundance of canopy soil fungal ASVs. The relative  
820 abundance values of unique and shared bacterial/archaeal and fungal ASVs (at class level) across  
821 the microsites are represented in Tables S3-6.

822

823 **FIG 4: Predicted fungal guilds in Sonoran Desert soil samples:** A) arbuscular mycorrhizal  
824 fungi, B) plant pathogen, and C) wood saprotroph. Proportions are depicted as the mean value and  
825 standard deviation.

826

827 **FIG 5: Gene abundance in Sonoran Desert soil samples.** A) 16S rRNA gene and B) *ureC*. Copy numbers  
828 were calculated per gram of dry soil and are depicted as the mean value and standard deviation.

829

830

831