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4 **Fragmentation disrupts microbial effects on native plant community productivity**

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6 **Running title:** Fragmentation disrupts microbial effects

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47 **Abstract**

- 48 1. Anthropogenic habitat fragmentation – the breaking up of natural landscapes – is a
49 pervasive threat to biodiversity and ecosystem function worldwide. Fragmentation
50 results in a mosaic of remnant native habitat patches embedded in human-modified
51 habitat known as the “matrix”. By introducing novel environmental conditions in
52 matrix habitats and reducing connectivity of native habitats, fragmentation can
53 dramatically change how organisms experience their environment. The effects of
54 fragmentation can be especially important in urban landscapes, which are
55 expanding across the globe. Despite this surging threat and the importance of
56 microbiomes for ecosystem services, we know very little about how fragmentation
57 affects microbiomes and even less about their consequences for plant-microbe
58 interactions in urban landscapes.
- 59 2. By combining field surveys, microbiome sequencing, and experimental mesocosms,
60 we (1) investigated how microbial community diversity, composition, and
61 functional profiles differed between 15 native pine rockland fragments and the
62 adjacent urban matrix habitat, (2) identified habitat attributes that explained
63 significant variation in microbial diversity of native core habitat compared to urban
64 matrix, and (3) tested how changes in urbanized and low connectivity microbiomes
65 affected plant community productivity.
- 66 3. We found urban and native microbiomes differed substantively in diversity,
67 composition, and functional profiles, including symbiotic fungi decreasing 81% and
68 pathogens increasing 327% in the urban matrix compared to native
69 habitat. Further, fungal diversity rapidly declined as native habitats became

70 increasingly isolated, with ~50% of variation across the landscape explained by
71 habitat connectivity alone. Interestingly, microbiomes from native habitats
72 increased plant productivity by ~300% while urban matrix microbiomes had no
73 effect, suggesting that urbanization may decouple beneficial plant-microbe
74 interactions. In addition, microbial diversity within native habitats explained
75 significant variation in plant community productivity, with higher productivity
76 linked to more diverse microbiomes from more connected, larger fragments.

77 4. *Synthesis*. Taken together, our study not only documents significant changes in
78 microbial diversity, composition, and functions in the urban matrix, but also
79 supports that two aspects of habitat fragmentation — the introduction of a novel
80 urban matrix and reduced habitat connectivity — disrupt microbial effects on plant
81 community productivity, highlighting preservation of native microbiomes as critical
82 for productivity in remnant fragments.

83

84 **Introduction**

85 Land conversion and anthropogenic habitat fragmentation - the breaking up of natural
86 landscapes - are widespread threats to biodiversity (Fletcher et al. 2018; Haddad et al.
87 2015). An estimated 75% of Earth's terrestrial surface has been subjected to substantive
88 human impacts, such as habitat fragmentation and loss (Venter et al. 2016). Fragmentation
89 can have meaningful consequences for biodiversity and native communities through
90 changes to remaining habitat patches (e.g., degradation in habitat quality, decrease in
91 habitat area) and their connectivity (e.g., increased patch isolation) (Laurance et al. 2002;
92 Haddad et al. 2015). Anthropogenic fragmentation can also affect communities through

93 introduction of novel “matrix” habitat(s), which is non-natural habitat interspersed
94 between native fragments (Kupfer et al. 2006). Historically matrix habitats were viewed as
95 uniform and ecologically less important than remaining fragments. However,
96 contemporary habitat fragmentation research has found that novel matrix habitats are also
97 critical component to understanding fragmentation’s effects on organisms and
98 communities persisting in human-modified landscapes (Driscoll et al. 2013; Jules &
99 Shahani 2003; Öckinger et al. 2012; Ricketts 2001; Williams et al. 2006;).

100

101 The introduction of novel matrix habitat into landscapes can have consequences for native
102 communities through several pathways. The type and characteristics of matrix habitats can
103 affect organisms’ dispersal between patches. At one extreme, matrix habitats can form
104 impassable barriers to species, decreasing dispersal to and from fragments and isolating
105 populations on individual remnants (Kupfer et al. 2006; Ricketts, 2001). At the other
106 extreme, matrices can act as conduits between native patches, providing additional
107 resources and transportation that increase dispersal (Bridgman et al. 2012; Stasek et al.
108 2008). Moreover, the introduction of novel matrix habitats surrounding fragments can also
109 affect organisms by changing availability and/or quality of habitat in the landscape
110 (Driscoll et al. 2013). For instance, the matrix can be uninhabitable for certain species, such
111 that they can only utilize remaining native habitat. However, in some cases, the matrix can
112 provide a new, preferred habitat or mitigate negative effects of habitat loss by providing
113 suitable, but lower quality habitat or additional resources (Ewers and Didham 2006; Evans
114 et al. 2016). The matrix habitat may also affect native communities by altering biotic
115 interactions that organisms usually experience in their native habitats. Matrix habitat can

116 have different environmental conditions than the native habitats they supplant (e.g., urban
117 environments with higher temperatures and increased pollutants; Grimm et al. 2008;
118 Johnson & Munshi-South 2017). These environmental changes can filter which organisms
119 occur in the area thereby changing the pool of possible competitors, predators, prey, and
120 mutualists available to interact with (Barros et al. 2019; Grimm et al. 2008; Miles et al.
121 2019; Moreira et al. 2019; Parsons et al. 2019). These changes in interactions can be as
122 important for organismal and community responses to the introduction of matrix habitats
123 as its abiotic features (Grimm et al. 2008; Johnson & Munshi-South 2017; Zarnetske et al.
124 2017). Through all of these ecological mechanisms, the matrix habitat can shape how
125 individuals and communities interact within fragmented landscapes.

126

127 In recent years, it has become increasingly clear that matrix habitats are vital to
128 understanding how communities are impacted by fragmented landscapes (Ewers &
129 Didham 2006; Jules & Shahani 2003; Kupfer et al. 2006). As a result, investigating how
130 communities differ in composition and productivity between matrix habitats and the cores
131 of native fragments is garnering more attention in the recent habitat fragmentation
132 literature (Evans et al. 2016; Ewers & Didham 2006; Jules & Shahani 2003; Kupfer et al.
133 2006; Lasmer et al. 2021; Matthews et al. 2021; Schlapfer et al. 2018). Studies examining
134 consequences of matrix type for plant species diversity and richness have found their
135 effects can be on par or even stronger than effects of habitat isolation and patch area (e.g.,
136 Guirado et al. 2006; Metzger 2000; Ockinger et al. 2012; Williams et al. 2006). For example,
137 Ockinger et al. 2012 found variation in plant richness of grassland fragments was explained
138 solely by matrix type. Fragments surrounded by a natural forest had higher richness than

139 fragments surrounded by agricultural matrix. However, these studies mainly focus on
140 macro-organisms despite microorganisms' importance to many communities and
141 ecosystem processes (e.g., nutrient cycling) as well as plant and animal health and
142 functioning (Christian et al. 2015; Fierer 2017; Mony et al 2022; Wagg et al. 2019; Wardle
143 2004).

144

145 Microbiomes are also likely to be affected by fragmentation of natural ecosystems. For
146 instance, microbial dispersal across landscapes, which depends on dispersal limitation and
147 environmental filtering, can be impacted by fragmentation through both reducing habitat
148 connectivity (e.g., increasing distance between suitable habitat/hosts) as well as shifts in
149 abiotic/biotic conditions on remanent fragments (e.g., reducing habitat quality) (Martiny et
150 al. 2006; Mony et al. 2022). Recent work has demonstrated that at least some microbial
151 taxa experience dispersal limitation as well as provided support for changes in the
152 environmental that are often associated with fragmentation affecting microbial
153 communities (Boeraeve et al. 2018; Flores-Rentería et al. 2016; Grilli et al. 2014;
154 Kieseewetter & Afkhami 2021; Peay et al. 2010; 2012; Vannette et al. 2016; Xiao et al. 2018).
155 Further, Kieseewetter & Afkhami (2021) demonstrated microbial communities responded to
156 fragmentation-driven changes in habitat connectivity, which in turn affected native plant
157 performance/allocation (at the individual plant level). However, it remains unclear how
158 fragmentation-driven connectivity and matrix effects on the microbiome scale up to impact
159 plant community-level dynamics and plant productivity in human-modified landscapes.

160

161 Here, we aim to improve our understanding of how urbanization and consequent habitat
162 fragmentation shape microbiomes and how these changes in microbial communities scale
163 up to affect plant community composition and productivity. We do this by integrating field
164 surveys, microbiome sequencing, and experimental mesocosm studies in the highly-
165 fragmented pine rockland habitat of South Florida using microbiomes from native core
166 habitats and urban matrix habitats. Microbiomes from native core habitat (hereafter
167 referred to as “native microbiomes”) are microbial communities collected from near the
168 center of pine rockland fragments, while urban matrix microbiomes (hereafter “urban
169 microbiomes”) are microbiomes collected from adjacent urban environments in which the
170 native pine rockland remnant habitat are embedded. Specifically, we (1) assessed
171 diversity, composition, and functional profiles of microbiomes from 15 habitat pairs of
172 native fragment core habitats and adjacent urban matrix habitats, (2) identified habitat
173 attributes that explained variation in microbial diversity within native versus urban
174 habitats, and (3) experimentally tested how these 15 native and 15 urban microbiomes
175 affect plant community productivity and composition. Given the strong contrast in the
176 habitat characteristics of native and urban habitats (Nugent & Allison, 2022), we
177 hypothesized that urban microbiomes will significantly differ in community composition,
178 diversity, and functional guilds from the native microbiomes. We also predicted that
179 habitat connectivity will play an important role in explaining variation in native
180 microbiomes due to the increased isolation native habitats experience in a fragmented
181 landscape, while unique stresses and frequent disturbances of the urban matrix (Nugent &
182 Allison, 2022) may lead to more stochastic urban microbial communities. Finally, we
183 hypothesized that native plant community productivity in our experimental mesocosms

184 would be greater when microbes from either habitat type are present compared to sterile
185 treatments, but plant community productivity will benefit more from native habitat
186 microbiomes than their urban counterparts.

187

188 **Methods**

189

190 *Study system*

191 The pine rocklands is a critically imperiled ecosystem occurring exclusively in South
192 Florida and the Florida Keys with limited distribution on some Caribbean islands (FNAI
193 2010). This habitat is characterized by frequent fire, oolitic limestone outcroppings, open
194 pine canopy, patchy subcanopy of palms and shrubs, and a rich herbaceous layer (Snyder et
195 al. 1990). Urbanization of Miami-Dade County introduced a novel matrix habitat that highly
196 fragmented the pine rocklands along the Miami rock ridge, an ~80km oolitic limestone
197 ridge where pine rocklands historically existed (Possley et al. 2008; Snyder et al. 1990).
198 The remaining habitat (~2% of the historic range; Figure 1) occurs along this ridge in 1-
199 300 acre fragments.

200

201 *Habitat classifications, soil collections, and habitat attributes*

202 To assess how microbial communities differed between native habitat and urban, we
203 identified 15 pine rockland fragments (Figure 1) spanning the historic pine rockland range
204 and whose adjacent matrix habitats (i.e., those within 500m buffer zones around the
205 fragments' centroids) included high-intensity urban areas as classified by Kawula & Redner
206 (2018). To determine each native habitat fragment's core, we created a 10m buffer around

207 each fragment's centroid (ArcMap 10.5; ESRI, Redlands, CA, USA). For each of the 15 native-
208 urban fragment pairs, we randomly selected three points within the native core buffer zone
209 and three points within the urban matrix buffer zone. Soils were aseptically collected from
210 each point, homogenized within fragment-habitat type combination (totaling 30
211 microbiomes), and transported to the University of Miami (Coral Gables, FL) on ice. We also
212 collected data on ten habitat attributes at each site, including three common metrics of
213 habitat connectivity to quantify the degree of fragmentation (i.e., distance to nearest
214 neighboring patch, patch-based weighted sum, number of fragments in 10km buffer;
215 Calabrese & Fagan, 2004; Kindlmann & Burel, 2008), three patch characteristics (fragment
216 area, time since most recent fire, plant richness), two urban matrix attributes (percent
217 transportation, evenness of the surrounding matrix habitat types), and two soil properties
218 (pH, Total Kjeldahl Nitrogen). We also determined phosphorus availability for a subset of
219 site pairs. See SI Methods on "Habitat attributes" for more details on these metrics. Field
220 work was permitted by Miami-Dade County Parks and Recreation (permit #309).

221

222 *Microbial DNA extractions, library preparations, processing, and characterization*

223 We extracted microbial DNA from 0.3g of soil for all sites using the DNeasy 96 PowerSoil
224 Pro QIAcube HT kit (Qiagen; 47021; Revillini et al. 2022). We checked DNA concentrations
225 and performed initial quality control using endpoint PCR, gel electrophoresis, and Qubit 4
226 Fluorometer (Invitrogen; Q33327) measurements. We prepared 60 prokaryotic and fungal
227 libraries using a two-step dual indexing protocol with *16S 515F/806R* (forward/reverse)
228 and *ITS ITS1/ITS2* amplicons and magnetic bead cleanups (Gohl et al., 2016). Libraries
229 were subsequently sequenced at University of Miami's Genomics Core with the Illumina

230 MiSeq platform (v3, 300bp paired end). Lack of contamination was confirmed with
231 negative controls (Ultrapure water replaced soil during extractions). After sequencing,
232 reads were processed using *QIIME2* (v2020.1) and denoised and grouped into exact
233 sequence variants (i.e., ESVs; 100% sequence similarity) using *dada2*. All samples reached
234 saturation (i.e., refraction curve plateaued) by ~2000 reads. We normalized samples,
235 calculated Shannon diversity, species richness, and Pielou's evenness (*vegan*; Oksanen
236 2020), and classified ESVs into microbial "species" using microbial databases (McDonald et
237 al. 2012; Werner et al. 2012; Nilsson et al. 2019). Details are available in SI Methods: "DNA
238 extractions, library preparation, and processing" and "Microbial DNA characterization").

239

240 *Understory community experiment*

241 We grew eight pine rockland native, understory plant species (ranging in life histories and
242 habitat specialization; Table S2) in a factorial mesocosm experiment with microbiomes
243 from 15 fragment origins x 2 habitat types (urban/native) x 2 microbial treatment
244 (live/sterilized) x 3 replicates (total=180 mesocosms). Sterile 1.6L pots were inoculated
245 with either *live* (microbial active) or *sterilized* soils from each of the 30 fragment-habitat
246 type combinations (i.e., 15 pairs of urban and native habitat microbiomes). Each pot
247 contained 30% (by volume) sterilized Miami rock ridge limestone, 50% sterile pine
248 rocklands background soil, a 15% inoculum layer of microbially active or sterile soil from
249 the relevant fragment-habitat type combination, and a 5% sterile soil cap to avoid
250 microbial desiccation (David et al. 2020; Kieseewetter & Afkhami 2021; SI Methods:
251 "Experimental setup"). To minimize site-specific abiotic effects of soils, we used a
252 standardized sterile background soil collected from a pine rockland site that is roughly

253 central to all fragments used in our study (Porter Russell Pine Rockland). This was done to
254 ensure that the majority of soil in each mesocosm had identical abiotic properties, with
255 only biotic components of the soil being able to disperse from the much smaller amount of
256 treatment soil (inoculum) to colonize the rest of the pot. Sterilized limestone/soils were
257 autoclaved three times at 121°C (2 hours/cycle) to remove the soil biota, which includes
258 fungi and prokaryotes (the focus of this study) as well as other soil fauna. All seeds were
259 collected from Miami-Dade pine rocklands by the authors and collaborators from pine
260 rockland restoration groups (Table S2), surface sterilized, and planted into prepared
261 mesocosms (1 seed/species/pot). Germination and survival were recorded weekly. After 7
262 months (December 2019-July 2020), plants were harvested, dried, and weighed.
263 Aboveground biomass was collected separately for individual plants within each
264 mesocosm, while belowground biomass was collected for each mesocosm.

265

266 *Data analysis*

267 All statistical analyses were performed with R v3.6.1 (R Core Team, 2021). To determine if
268 microbial diversity differed between native and urban habitats, we performed mixed linear
269 models (*lme4*; Bates, 2019) with habitat type as the explanatory variable, fragment-urban
270 pair identity as a random effect, and either fungal or prokaryotic diversity as the response
271 variable. Upon finding significant diversity effects, we investigated which components of
272 Shannon diversity (species richness or Pielou's evenness) were responsive to habitat type
273 using mixed linear models with fungal/prokaryotic diversity components as new response
274 variables and the same explanatory and random variables (Table S3). We then performed
275 PERMANOVA to test if microbial community composition differed between habitat types

276 using *adonis* (*vegan*, Oksanen 2020), with a Bray-Curtis distance matrix of prokaryotic or
277 fungal taxa, and a random effect of fragment identity. To investigate which microbial
278 families are particularly important for distinguishing habitat type, we performed random
279 forest models with Boruta feature selection, accounting for multiple comparisons (*Boruta*,
280 Kursa & Rudnicki 2020). We used FUNGuild (Nguyen et al. 2016) to classify fungal taxa into
281 trophic modes – pathotrophs, saprotrophs, and symbiotrophs – and permutational ANOVA
282 (*aovp* in *lmPerm*; Wheeler & Torchiano 2016) to determine if the relative abundance of
283 trophic modes differed between habitat types (Table S4). After determining microbiomes
284 diverged significantly between habitat types, we evaluated what habitat attributes had
285 main effects that explained variation in microbial diversity in each habitat (Table S5). We
286 conducted model selection using *dredge* (MuMIn package, Bartón 2022) with diversity as
287 the response variable and habitat attributes as explanatory variables. The best regression
288 model was selected based on corrected Akaike information criterion (Akaike 1974). We
289 then calculated relative importance of each habitat attribute selected by the best model
290 using *calc.relimp* (*relaimpo*; Grömping, 2006).

291

292 To investigate plant community productivity, we used ANOVA with the mesocosms' total
293 biomass (overall productivity) as the response and microbial treatment, habitat type, and
294 their interaction as explanatory variables, and planned contrasts between live and sterile
295 inoculum treatments within each habitat type. After finding significant effects on overall
296 productivity, we conducted these analyses separately for shoot and root biomass (summed
297 across all species for each mesocosm) to determine whether changes in productivity
298 aboveground, belowground, or both were responsible for changes in overall productivity

299 (Table S6). For each habitat type, we also investigated how shifts in microbial diversity,
300 richness, and evenness may affect plant community properties with a linear regression
301 where the response variable was a plant community metric (e.g., mesocosm total biomass)
302 and the explanatory variable was a prokaryotic or fungal diversity metric and its
303 interaction with microbial treatment (Table S7). Since we could preserve plant species
304 identity for shoot biomass, we tested whether the effect of urban and native microbiomes
305 on shoot mass was similar across plant species within the community for the four species
306 with the highest germination using permutational ANOVAs (aovp function) and planned
307 contrasts between the live and sterile treatments within habitat type (Table S8). We
308 additionally summed the shoot biomass for the other four plant species with lower
309 germination and repeated the same analysis. Finally, to determine if plant community
310 composition was affected by microbial treatment, habitat type and their interaction, we
311 performed a PERMANOVA (*adonis*) with a response matrix of the presence/absence of each
312 plant species in each mesocosm (SI Methods: "Data analysis").

313

314 **Results**

315

316 *Native and urban habitats significantly differed in microbial diversity, composition, and*
317 *functional guilds.*

318 Prokaryotic diversity in native habitat fragments was ~12% greater than in the adjacent
319 urban habitat ($\chi^2(1,N=15)=3.94,P=0.047$; Figure 2a). This change in diversity was driven
320 primarily by a 47% increase in prokaryotic richness in native habitat ($\chi^2(1,N=15)=5.54$,
321 $P=0.019$; Figure 2b). In contrast, fungal communities from native habitats were ~18% less

322 diverse than those from urban habitats ($\chi^2(1,N=15)=9.13,P=0.003$; Figure 2a).
323 Interestingly, the difference in fungal diversity was largely driven by a 21% increase in
324 Pielou's evenness in the urban habitat ($\chi^2(1,N=15)=22.0, P<0.001$; Figure 2c) rather than
325 changes in richness ($\chi^2(1,N=15)=0.032, P=0.858$; Figure 2b).

326

327 Additionally, both prokaryotic and fungal community composition differed significantly
328 between native and urban habitats (prokaryotic: $F_{1,29}=1.11, P=0.020$; fungi: $F_{1,29}=2.54,$
329 $P=0.001$; Figure 3). Using random forest models, we identified seven prokaryotic families
330 and 14 fungal families that were important for distinguishing between habitat types (Table
331 1). Interestingly, all eight prokaryotic families had higher relative abundances in native
332 habitats than in urban, while the 15 fungal families' associations with urban versus native
333 habitat were more equally distributed (Table 1). Grouping these fungal families into
334 putative trophic modes (i.e., saprotrophs, pathotrophs, symbiotrophs; Table 1) provided
335 additional insight. We found that plant-associated guilds -- pathotrophs and symbiotrophs -
336 - had higher relative abundance in urban and native habitats, respectively. These results
337 were additionally supported by a FUNGuild analysis of the broader fungal community,
338 which found fungal pathogens were 327% more abundant in urban than native core
339 ($F_{1,28}=16.4, P<0.001$) and symbiotrophs were 420% more abundant in native habitat,
340 although the latter trend was non-significant ($F_{1,28}=1.96, P=0.172$).

341

342 *Habitat attributes, including connectivity, were better able to explain variation in native*
343 *microbial diversity than urban diversity.*

344 In native habitats, 81% of the variation in fungal diversity ($R^2=0.811$, $F_{3,10}=14.32$, $P=0.001$)
345 and ~40% of the variation in prokaryotic diversity ($R^2=0.432$, $F_{2,11}=4.19$, $P=0.044$) were
346 explained by habitat attributes. Specifically, 46% of landscape-wide variation in fungal
347 diversity was explained by habitat connectivity (patch-based weighted sum: $F_{1,10}=25.98$,
348 $P<0.001$; Figure 4a), and another 35% was explained by the fragment's plant richness
349 ($F_{1,10}=8.11$, $P=0.017$) and the adjacent urban habitat's percent transportation ($F_{1,10}=6.08$,
350 $P=0.033$; Figure 4a). Matrix evenness ($F_{1,11}=6.22$, $P=0.030$) and soil pH ($F_{1,11}=5.97$,
351 $P=0.033$) also each explained ~20% of the variation in prokaryotic diversity (Figure 4a).
352 In contrast, in urban microbiomes prokaryotic diversity was not explained by any of the ten
353 habitat attributes, while transportation, time-since-fire, and number of fragments in a
354 10km buffer together explained ~70% of the variation in fungal diversity ($R^2=0.701$,
355 $F_{2,11}=7.82$, $P=0.006$; Figure 4c), with the habitat connectivity only explaining 17% of the
356 total variation (number of fragments within a 10km buffer: $F_{1,11}=6.00$, $P=0.034$; Figure 4c).
357 We also found that in native habitats ~80% and 41% of the variation in fungal richness and
358 prokaryotic richness, respectively, were explained by habitat attributes (fungal: $F_{4,9}=8.96$,
359 $P=0.003$; bacterial: $F_{2,11}=3.82$, $P=0.055$; Figure 4b). Specifically, ~15% of the variation in
360 fungal richness was explained by fragment area ($F_{1,9}=13.72$, $P=0.005$), while the other 65%
361 was explained by matrix evenness (i.e. the evenness of the habitat types in the matrix
362 surrounding the fragment) ($R^2=0.088$, $F_{1,9}=5.42$, $P=0.045$), plant richness ($R^2=0.472$,
363 $F_{1,9}=31.96$, $P<0.001$), and soil pH ($R^2=0.094$, $F_{1,9}=11.28$, $P=0.008$). Prokaryotic richness was
364 explained equally by two attributes – % transportation ($R^2=0.181$, $F_{1,11}=3.94$, $P=0.073$) and
365 soil pH ($R^2=0.229$, $F_{1,11}=4.84$, $P=0.050$; Figure 4b). Similar to diversity, the habitat
366 attributes explained none of the variation in prokaryotic richness and a smaller amount of

367 the variation in fungal richness (60%) in the urban habitats ($F_{2,11}=8.21$, $P=0.007$; Figure
368 **4d**). Overall, we found that habitat attributes, including measures of habitat connectivity,
369 are explaining less, if any, variation in urban microbiomes compared to native
370 microbiomes.

371

372 *Microbiomes from highly-connected native habitats increase plant community productivity*
373 *unlike urban microbiomes.*

374 Native habitat microbiomes increased overall plant community productivity by ~290%
375 (contrast of native live vs. sterilized treatments: $F_{1,177}=17.5$, $P<0.001$; Figure **5a**), while
376 urban microbiomes had a weakly positive, nonsignificant effect on plant community
377 productivity (contrast of urban live vs. sterilized treatments: $F_{1,177}=3.25$, $P=0.073$; Figure
378 **5a**). When we investigated the components of overall productivity -- shoot and root
379 biomass -- to determine if microbiome effects on above- or below-ground productivity
380 were driving microbial effects on overall productivity, we found that both components
381 were impacted in similar ways to overall productivity. Aboveground productivity of
382 communities grown with native habitat microbiomes increased ~319% and belowground
383 productivity increased ~244% compared to the sterile treatment (shoot: $F_{1,177}=17.5$,
384 $P<0.001$; root: $F_{1,177}=12.0$, $P<0.001$; Figure **5b-c**). Urban microbiomes had weakly positive,
385 nonsignificant effects on communities' aboveground and belowground biomass (shoot:
386 $F_{1,177}=2.44$, $P=0.120$; root: $F_{1,177}=3.60$, $P=0.059$; Figure **5b-c**). These results indicate that
387 native microbiomes play key roles in plant productivity and urbanization of the
388 microbiome likely reduces beneficial effects of and plant reliance on the soil microbiome.
389 Despite a decreased plant reliance on the urban soil microbiomes, these communities had

390 an overall higher productivity than communities grown with native soil microbiomes.
391 While our experiment used a design aimed at minimizing the differences in abiotic
392 conditions (i.e. by using the same soil/limestone rock for 85% of the soil structure in all
393 mesocosms), the inoculum soils from urban habitats had ~58% higher nitrogen (3943.7
394 +/- 414.5 sem) and ~280% higher phosphorus (9.19 +/- 1.41 sem) compared to the
395 inoculum soils from native habitats (nitrogen: 2517.3 +/- 623.2 sem; phosphorus: 2.43 +/-
396 0.17 sem) (nitrogen: $F_{1,26}=3.63$, $P=0.068$; phosphorus: $F_{1,12}=22.72$, $P<0.001$), which likely
397 explains the overall higher plant community productivity in urban mesocosms. Further,
398 plant community richness was higher in urban than native habitats ($F_{1,157}=6.95$, $P=0.005$)
399 and increased when microbes were present ($F_{1,157}=10.2$, $P=0.001$); however, unlike
400 productivity, the strength of microbial effects was consistent between habitat types (i.e., no
401 significant interactive effect).

402

403 To determine whether aboveground biomass responses to native and urban microbiomes
404 were driven by particular plant community members, we evaluated individual responses of
405 the four highest germinating species (Fig S5). Three of four species - *Bidens alba*, *Passiflora*
406 *suberosa*, and *Psychotria nervosa* -- grew significantly better with native habitat
407 microbiomes, increasing productivity from double to ~25 times more compared to
408 sterilized treatment (Fig S5a, c, d; $P=0.020$, $P=0.008$, $P=0.019$). The fourth, *Callicarpa*
409 *americana*, also followed this trend (~19.5 times more biomass with native microbiomes;
410 Fig S5b; $P=0.111$). The consistent beneficial effects of native microbiome on plant biomass
411 across these dominant species indicates the community-level productivity response to
412 native microbiomes was not driven by an outlier species but rather may represent a

413 general importance of native habitat microbiomes to native plant productivity. Conversely,
414 urban microbiomes significantly increased productivity of only one species, *P. suberosa*
415 ($P < 0.001$; Fig S3c), with the other three dominant species unresponsive to urban
416 microbiomes. Additionally, when we evaluated the combined aboveground biomass of the
417 four lower-germinating species (*Chamaecrista fasciculata*, *Senna ligustrina*, *Sorghastrum*
418 *secundum*, *Vernonia blodgettii*), we found that shoot biomass was higher in microbially live
419 treatments ($P < 0.001$), regardless of habitat type, indicating that the difference between
420 urban and native microbiomes is less important for productivity of these low germinating
421 taxa.

422

423 To further understand how fragmentation-associated changes in microbiomes affect plant
424 communities, we evaluated the relationship between microbial community properties (i.e.
425 diversity, richness and evenness) and plant productivity and richness in our understory
426 mesocosm. For instance, native microbiome diversity explained significant variation in
427 plant community productivity and richness (Figure 4). Specifically, shifts towards greater
428 fungal diversity – which occurred at sites with higher habitat connectivity ($F_{1,10} = 25.98$,
429 $P < 0.001$) – were associated with increased shoot biomass in our mesocosm experiment
430 ($F_{1,86} = 4.43$, $P = 0.038$; Figure 4a). Similarly, high fungal richness – which occurred in larger
431 fragments ($F_{1,9} = 13.72$, $P = 0.005$) – was associated with increased total biomass ($F_{1,86} = 8.53$,
432 $P = 0.004$), shoot biomass ($F_{1,86} = 6.98$, $P = 0.010$), root biomass ($F_{1,86} = 6.92$, $P = 0.010$) and
433 richness ($F_{1,86} = 5.47$, $P = 0.022$) (Figure 4b). Importantly, these positive relationships
434 between microbial diversity/richness and plant productivity/richness were only found
435 when the microbiome was present (live soil inoculations). In sterile soil treatments

436 (microbes absent), soils from sites with higher fungal diversity and richness were
437 associated with decreased plant community productivity and richness. Together, these
438 findings indicate the positive effects on productivity are likely caused by changes in
439 microbial community properties rather than changes in soil properties. Plant communities
440 grown with urban microbiomes only responded to changes in prokaryotic richness – where
441 increases in richness were linked to increased shoot biomass ($F_{1,86}=6.83$, $P=0.011$).
442 However, changes in urban prokaryotic richness were not explained by any of our
443 fragmentation metrics, suggesting these shifts in prokaryotic richness affecting shoot
444 biomass were driven by non-fragmentation factors.

445

446 **Discussion**

447

448 Our study demonstrates how habitat fragmentation-associated changes in microbiomes of
449 both native and urban habitats can alter plant-microbe interactions and plant community
450 productivity. Specifically, urban microbiomes had lower prokaryotic diversity, increased
451 fungal pathogens, and weakened relationships between microbial properties and
452 commonly-important habitat attributes. The urban microbiomes, in turn, showed reduced
453 microbial benefits for plant community productivity and decoupling of the strong positive
454 relationships between microbial diversity and plant community productivity. In contrast,
455 native core microbiomes had greater prokaryotic diversity, fewer pathogens and possibly
456 more symbiotic microbes, and these microbiomes significantly benefited both community-
457 wide and species-specific plant productivity (~300% greater productivity). The dichotomy
458 between the native and urban microbiomes' makeup and effects on plant productivity

459 illustrates how introduction of novel, nonnative habitats into fragmented landscapes could
460 disrupt natural plant-microbiome interactions. Interestingly, the beneficial effects of native
461 habitat microbiomes may also be weakened by fragmentation through changes in habitat
462 connectivity, as compositional shift in the microbiomes from low connectivity sites were
463 associated with declines in plant productivity. Microbial diversity in the native habitat was
464 well explained by common habitat attributes (e.g., 80% of variation in fungal diversity),
465 and importantly, this included a significant decline in diversity in patches with lower
466 habitat connectivity. The positive effects of native microbiomes were also sensitive to these
467 fragmentation-associated changes in microbial diversity and richness, with microbiomes
468 from low connectivity sites providing reduced growth benefits to our understory plant
469 communities. Overall, these results support fragmentation affecting plant productivity
470 through two microbial pathways: (1) changes in urban matrix microbiomes decoupling of
471 plant-microbial interactions and (2) isolation-associated shifts in native microbiome
472 diversity underlying negative effects on plant productivity in native remnant habitats.
473 Below we further explore these novel results and highlight future directions.

474

475 *Fragmentation's role in shaping soil microbiomes through introduction of a novel habitat*
476 *matrix and reduced habitat connectivity.*

477 Urban microbiomes diverged in community composition, diversity, and putative functional
478 guilds, which suggests urbanization's filtering/selection on microbial communities can
479 occur in consistent ways across large swaths of a fragmented landscape. Other studies have
480 also found compositional differences between native and matrix habitats (Brinkmann et al.
481 2019; Chen et al. 2020; Mendes et al. 2015; Nakayama et al. 2019), but fewer found effects

482 on microbial diversity (but see Schneider et al. 2015; Tin et al. 2018). Native remnants in
483 our study had higher prokaryotic diversity due to much higher richness, suggesting habitat
484 heterogeneity may generate more microniches for prokaryotes on native fragments
485 (Groffman et al. 2014). In contrast, fungal diversity was lower in native habitats and was
486 primarily driven by decreased fungal evenness, suggesting native habitat supports more
487 dominant fungal taxa. The stronger difference in microbial diversity between native and
488 urban habitats that we detected compared to other studies may result from differences in
489 the type of matrix habitat(s) and the matrix contrast (i.e., how greatly the matrix habitat
490 differs from the natural habitat; Kupfer et al. 2016) investigated. The high-intensity urban
491 matrix surrounding native forests in our study represents substantial matrix contrast,
492 which could explain why we found stronger changes in diversity of urban microbiomes
493 than has been found in other systems where contrast is lower (e.g., agricultural fields:
494 Mendes et al. 2015; Nakayama et al. 2019).

495

496 Further, changes in urban microbiome composition and diversity appear to cause
497 functional guild shifts that can negatively impact primary producers. For instance, all five
498 putative pathogenic fungal families identified by random forest models as differentiating
499 urban and native habitats were higher in urban matrix, while putative symbiotic fungi were
500 higher in native habitats. This shift from symbiotrophs to pathotrophs in the urban matrix
501 was further supported by FUNGuild functional assignments. One other study found similar
502 guild trends when comparing tropical forests and agricultural matrix; however, effect sizes
503 were smaller, possibly due to lower matrix contrast (Brinkmann et al. 2019). Taken
504 together, our results emphasize how the introduction of novel matrix habitats across

505 fragmented landscapes could shape microbiomes with potential to impact plant-
506 microbiome interactions.

507

508 Our study also highlighted how microbiomes inside remaining native habitats change with
509 increasing habitat fragmentation. In fact, ~50% of variation in native fungal diversity was
510 explained by habitat connectivity alone, suggesting dispersal limitation (a key mechanism
511 for understanding microbial biogeography; Martiny et al. 2006; Mony et al. 2022) is
512 important in structuring microbial diversity across the landscape. While historically the
513 paradigm ‘everything is everywhere’ has been applied to microbes (Baas-Becking 1934),
514 increasingly evidence supports dispersal limitation as important for fungi (Kiesewetter &
515 Afkhami 2021; Vannette et al. 2016). For instance, Peay et al. 2010 found in a naturally-
516 fragmented landscape ectomycorrhizal fungi were dispersal limited, with species richness
517 decreasing ~50% at ~1km. Further, environmental filtering (another key component to
518 microbial biogeography; Martiny et al. 2006; Mony et al., 2022) is also contributing to
519 microbiome structure. For instance, despite dispersal limitation being unlikely within
520 fragment-urban pairs, there were significant differences in microbial diversity,
521 composition, and function between native habitat and adjacent urban matrix. Due to the
522 overwhelming amount of urban matrix in which native fragments are embedded (similar to
523 many urban environments; Angel et al. 2011), microbial inundation from the urban matrix
524 into native habitats is expected to be especially strong. Yet, we still find clear differentiation
525 of native habitat microbiomes from nearby urban matrix as well as less pathogenic/more
526 symbiotic microbes on native habitat fragments, supporting filtering-mediated resistance
527 of native microbiomes to the introduction of matrix habitats. The importance of

528 environmental filtering across the landscape was also emphasized by the large amount of
529 variation in microbial diversity explained by habitat attributes in native habitats. In
530 contrast, habitat attributes explained much less variation in urban microbiomes (~30%-
531 43% less than for native microbiomes), possibly due to urban matrix habitats being more
532 stochastic and/or disturbed (Nugent & Allison 2022). Our findings underscore the
533 importance of improving habitat connectivity and preserving abiotic and biotic conditions
534 that underpin environmental filtering in native fragments.

535

536 *Urbanization and disruption of beneficial plant-microbial interactions in the urban matrix*
537 *and remnant native habitats*

538 Given the rapidly increasing fragmentation of landscapes and the importance of
539 microbiomes for plant community health and dynamics (Berendsen et al. 2012; Fierer
540 2017), there is growing interest in how this relationship plays out in fragmented
541 landscapes. Observational studies have found microbial diversity is correlated with shifts
542 in plant community composition of revegetated/restored matrix habitat (Gellie et al. 2017;
543 Turley et al. 2020) and tied land conversion to microbial shifts in plant rhizospheres
544 (Estendorfer et al, 2017; McGee et al. 2020). For instance, tree rhizosphere communities
545 significantly differed for forests adjacent to rural/suburban versus urban matrices (Rosier
546 et al. 2021). By evaluating rhizosphere microbiomes, these studies infer which microbes
547 are likely interacting with which plants under different contexts, contributing towards
548 understanding microbe-plant interactions in fragmented landscapes. However, correlative
549 approaches cannot test whether changes in plant microbiomes are responsible for changes
550 in performance/composition of plant communities nor can they decouple microbial-

551 mediated effects and effects of other biotic and abiotic differences between habitat types.
552 Our study takes a step toward addressing this gap by experimentally manipulating
553 microbial communities from native and urban matrix habitats and assessing impacts on
554 plant community productivity reaching two main conclusions. First, urban microbiomes,
555 which are shifted in composition, diversity, and functions, did not confer productivity
556 benefits provided by native habitat microbiomes, suggesting that urban matrix disrupts
557 natural plant-microbial associations. Second, native microbiomes, which were shaped by
558 fragmentation-associated habitat attributes such as connectivity, play crucial beneficial
559 roles in native plant community productivity (here increasing it by nearly 300%).
560 Therefore, our results support that in addition to fragmentation shaping plant-microbiome
561 interactions through introduction of a novel urban matrix, fragmentation also shapes
562 microbiome effects on plant productivity through changes in habitat connectivity. In line
563 with this result, we recently demonstrated that connectivity-associated changes to
564 microbiomes can impact individual-level plant performance (Kiesewetter & Afkhami
565 2021), but this is the first experimental study showing that these changes in the
566 microbiome cascade up to impact plant community level properties. Taken together, our
567 results revealed how urbanization and habitat fragmentation are likely shaping hidden
568 microbial players in several ways that affect plant communities.

569

570 Despite plant communities' decreased reliance on urban microbiomes, it is important to
571 note that the consequences of introduced matrix habitats for plants can be complex,
572 altering many aspects of the abiotic and biotic environment organisms experience (Driscoll
573 et al. 2013; Kupfer et al. 2006). In our study, plant communities grown in urban soils,

574 regardless of microbial treatment, did better than those grown with native soils. Increased
575 productivity in urban mesocosms likely resulted from higher nutrients in urban matrix
576 soils (nitrogen: $F_{1,26}=3.63$, $P=0.068$; phosphorus: $F_{1,12}=22.72$, $P<0.001$), as commonly
577 reported for other urban environments (Grimm et al. 2008; Johnson & Munshi-South
578 2017). While the shift in urban microbiomes may itself have reduced microbially-conferred
579 productivity benefits (recall: increased pathotrophs and fewer symbiotrophs in urban
580 microbiomes), an alternative hypothesis is urban matrix-specific changes in context (i.e.,
581 increased nutrient availability) reduced the value of the microbiome to the plant
582 community, decreasing plant community reliance on urban microbiomes compared to
583 native microbiomes. In both cases, the urban matrix would be affecting plant-microbiome
584 interactions (i.e., by shifting microbiome makeup or by shifting its value to plant
585 communities). In reality, these 'alternatives' are not mutually exclusive and likely both
586 contribute to plant community productivity. In addition to differences in nutrient
587 availability, urban matrices are characterized by increased impervious land area (Johnson
588 & Munshi-South 2017), higher heavy metal concentration (Rodríguez Martin et al. 2015),
589 and higher temperatures (Grimm et al. 2008), which can reduce physical space for plant
590 communities and increase stress they experience. Additional stress in urban matrix may
591 generate greater reliance of the plant community on beneficial microbial partners
592 (Chandrasekaran et al. 2014; Meena et al. 2014) or could make the plant community more
593 susceptible to the negative effects of microbial pathogens (Pandey & Senthil-Kumar 2019).
594 Future work testing how key abiotic and biotic changes in the urban matrix modulate the
595 relationship between microbiomes and plant community dynamics (e.g., nutrient additions,

596 changes in other soil properties and sunlight environment, etc in native and urban soils)
597 will be an important next step.

598

599 *Conclusions and future directions*

600 In conclusion, our study not only illustrates how matrix introduction and habitat isolation
601 could drive meaningful differences in diversity, composition, and functions of microbiomes,
602 but also experimentally demonstrated for the first time how these changes can scale up to
603 strongly impact plant community productivity. From this work, we also identified four
604 areas for future research to further our understanding of plant-microbial interactions in
605 urbanized and fragmented environments. First, we postulate that matrix type and contrast
606 are important factors for determining fragmentation's consequences for microbial
607 diversity and composition with cascading effects on plants. Therefore, we advocate for
608 future work aimed at how multiple matrix types and contrasts (ideally within the same
609 system) impact microbiomes and their interactions with plant communities. Second, plant-
610 associating functional fungal guilds -- symbiotrophs and pathrotrophs -- differed in relative
611 abundance in urban and native habitats in our study, such that microbial communities in
612 the urban matrix appear to be functionally shifted towards less plant-beneficial
613 microbiomes. Therefore, detailed studies of microbial functional traits in matrix versus
614 native habitats would be especially valuable. In addition to functional assays,
615 metatranscriptomic approaches would be useful to gain community-wide profiles of
616 microbial functional shifts via changes in expression of functional genes in native and
617 urban matrix environments. Third, while our work focused on soil microbiomes, soils
618 contain complex communities, which also include other important and understudied soil

619 fauna (e.g., nematodes). As urbanization presents unique challenges for soils and their
620 communities, an important next step is to study how other members of the soil fauna and
621 their interactions with microbial communities respond to urbanized landscapes. Finally,
622 our work highlights how fragmentation can disrupt beneficial microbiome-plant
623 interactions through multiple pathways, spotlighting the need to study whether this change
624 will affect the resiliency of fragmented communities to other global change stressors, such
625 as climate change, as urbanization continues to increase globally (Seto et al. 2012).

626

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641

642 **Conflicts of Interest**

643 The authors declare no conflicts of interest.

644

645 **Author Contributions**

646 Kasey N. Kiesewetter and Michelle E. Afkhami developed the conceptual framework for the
647 project and experimental design. Kasey N. Kiesewetter performed field collections, lab
648 work, and sample processing with supervision from Michelle E. Afkhami. Leydiana Otano
649 performed lab work and experimental setup, with supervision from Michelle E. Afkhami
650 and Kasey N. Kiesewetter. Writing, editing, and reviewing were done by Kasey N.
651 Kiesewetter and Michelle E. Afkhami.

652

653 **Data Availability**

654 Demultiplexed sequencing data are available in the NCBI Sequence Read Archive (Bio-
655 Project ID: PRJNA938951; accession numbers: SAMN33450569 - SAMN33450598). All
656 experimental data script and data from the experiment and processed microbial data are
657 available on Zenodo ([10.5281/zenodo.7699344](https://zenodo.org/doi/10.5281/zenodo.7699344)).

658

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935 **Figure legends**

936 **Figure 1.** Map of historic Miami Rock Ridge pine rocklands (white layer) overlaid with
937 current pine rockland habitat (red layer) showing the 15 pine rockland fragments where
938 pairs of soil microbiomes were collected. Soils were collected in the core of each pine
939 rockland fragment and in the adjacent urban matrix. See Fig. S1 for more details.

940

941 **Figure 2.** Shannon diversity (**a,b**) and its components (species richness (**c,d**) and Pielou's
942 evenness (**e,f**)) in native and urban matrix habitat types. Prokaryotic diversity was higher
943 ($\chi^2(1,N=15)=3.94$; $P=0.047$; **a**) and fungal diversity was lower ($\chi^2(1,N=15)=9.13$, $P=0.003$;
944 **b**) in native habitats compared to urban matrix. Prokaryotic diversity was mainly driven by
945 differences in species richness ($\chi^2(1,N=15)=5.54$, $P=0.019$; **c**) rather than evenness
946 ($\chi^2(1,N=15)=1.16$, $P=0.281$; **e**). Fungal diversity was not driven by species richness
947 ($\chi^2(1,N=15)=0.032$, $P=0.858$; **d**) but rather fungal evenness ($\chi^2(1,N=15)=22.0$, $P<0.001$; **f**).
948 'n.s.' denotes a nonsignificant p-value, *denotes a p-value <0.05 , **denotes a p-value <0.01 ,
949 and *** denotes a p-value <0.001 . Error bars represent mean +/- standard error.

950

951 **Figure 3.** Bacterial (**a, b**) and fungal (**c, d**) community compositions differed in native core
952 habitats versus urban matrix habitats. The principle coordinate analyses (**a, c**) show native
953 core sites as green points and urban matrix sites as purple points, and points are labeled

954 with fragment identity abbreviations that correspond to the fragment identities listed in
955 Table S1. F statistics and p-values are reported from a PERMANOVA with a Bray-Curtis
956 distance matrix for either bacteria (**a**) or fungi (**c**) and the explanatory variables of habitat
957 type (i.e. native core versus urban matrix) and fragment identity. Bolded values indicate
958 significant p-value. For the bar graphs (**b, d**), each color represents the relative abundance
959 of a microbial class present in our data. Microbial taxa that could not be grouped as far
960 down as class, were grouped together into “unclassified” phylum or kingdom based on the
961 next most specific taxonomic order known for each OTU.

962

963 **Figure 4.** Habitat-associated shifts in fungal diversity and richness of the native
964 microbiomes explain variation in plant community productivity and richness. First,
965 variation in native microbiome diversity and composition (**a, b**) were explained by habitat
966 attributes of the landscape, including habitat connectivity and fragment area, while urban
967 microbiomes were not (**c, d**). This result is shown in graphs a-d, which show the strength of
968 habitat attributes’ relationships with diversity and richness of prokaryotes (blue bars) and
969 fungi (red bars) based on the attributes’ standardized regression slopes from best fit
970 models. Bars below zero represent a negative relationship, while bars above zero represent
971 positive relationships. Second, native fungal diversity and richness explained significant
972 variation in multiple plant performance metrics (solid red arrows) while urban diversity
973 did not. Arrows represent significant relationships between shifting microbial
974 diversity/richness and plant community productivity metrics (i.e., total biomass, shoot
975 biomass, and root biomass) and mesocosm richness. Solid arrows represent the positive
976 relationships we detected for live treatments (i.e., microbes were present) and faded

977 arrows represent the negative relationships we detected for sterile treatments (i.e.,
978 microbes were absent).

979

980 **Figure 5.** Native microbiomes increased community productivity, while urban
981 microbiomes did not significantly affect productivity. Overall productivity **(a)** was
982 significantly greater in mesocosms grown with the native habitat microbiomes compared
983 to mesocosms grown with sterilized native inoculum (total biomass: $F_{1,177}=17.5$, $P<0.001$),
984 while microbial effects of urban microbiomes on total biomass were weak ($F_{1,177}=3.25$,
985 $P=0.073$). Aboveground **(b)** and belowground **(c)** productivity showed the same outcome
986 with native microbiomes increasing both components of productivity (shoot mass:
987 $F_{1,177}=17.5$, $P<0.001$; root biomass: $F_{1,177}=12.0$, $P<0.001$) and urban microbiomes having
988 nonsignificant effects on both components (shoot mass: $F_{1,177}=2.44$, $P=0.120$; root mass:
989 $F_{1,177}=3.60$, $P=0.059$). Significance between microbial treatments within habitat type were
990 determined using planned contrasts. 'n.s.' denotes a nonsignificant p-value, 'n.s.¹', denotes a
991 marginally significant, p-value (0.05-0.1), *denotes a p-value <0.05 , **denotes a p-value $<$
992 0.01 , and *** denotes a p-value <0.001 . Error bars represent mean +/- standard error. M+
993 indicates microbial active/live treatment and M- indicates microbial sterile treatments.

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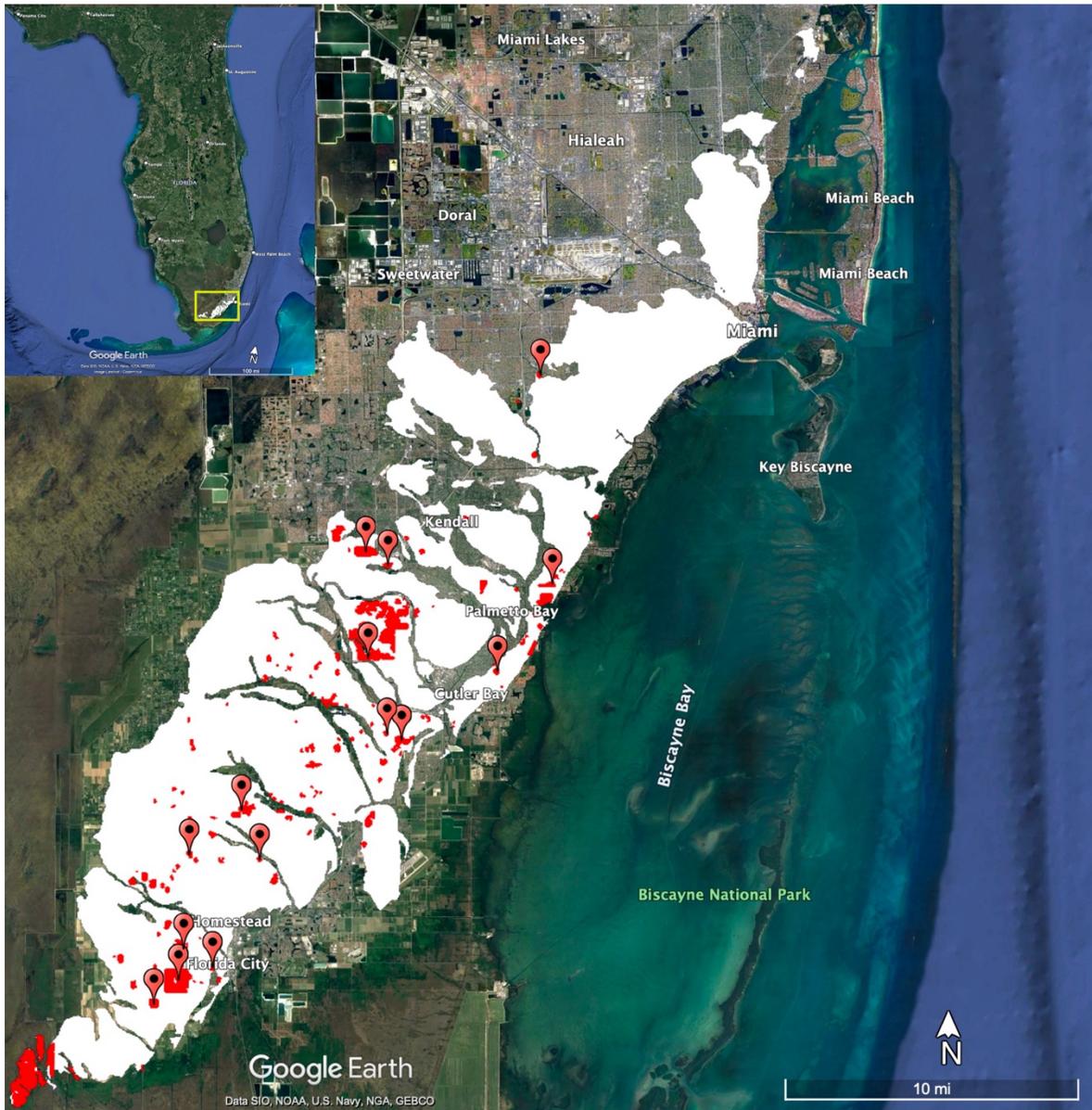
995 **Table 1.** Microbial families selected as important for distinguishing between native and
996 urban matrix habitat types by the random forest models with Boruta feature selection.

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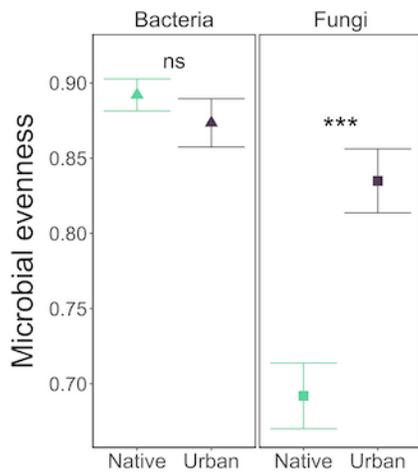
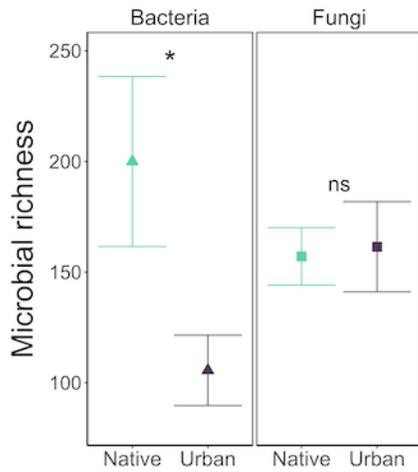
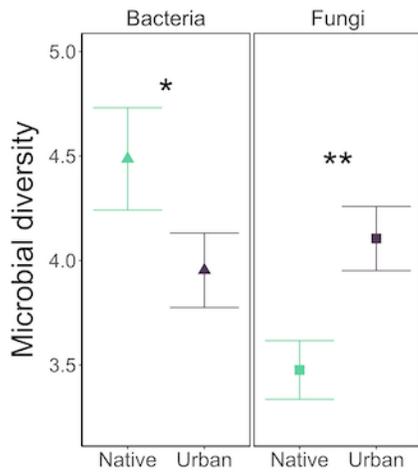
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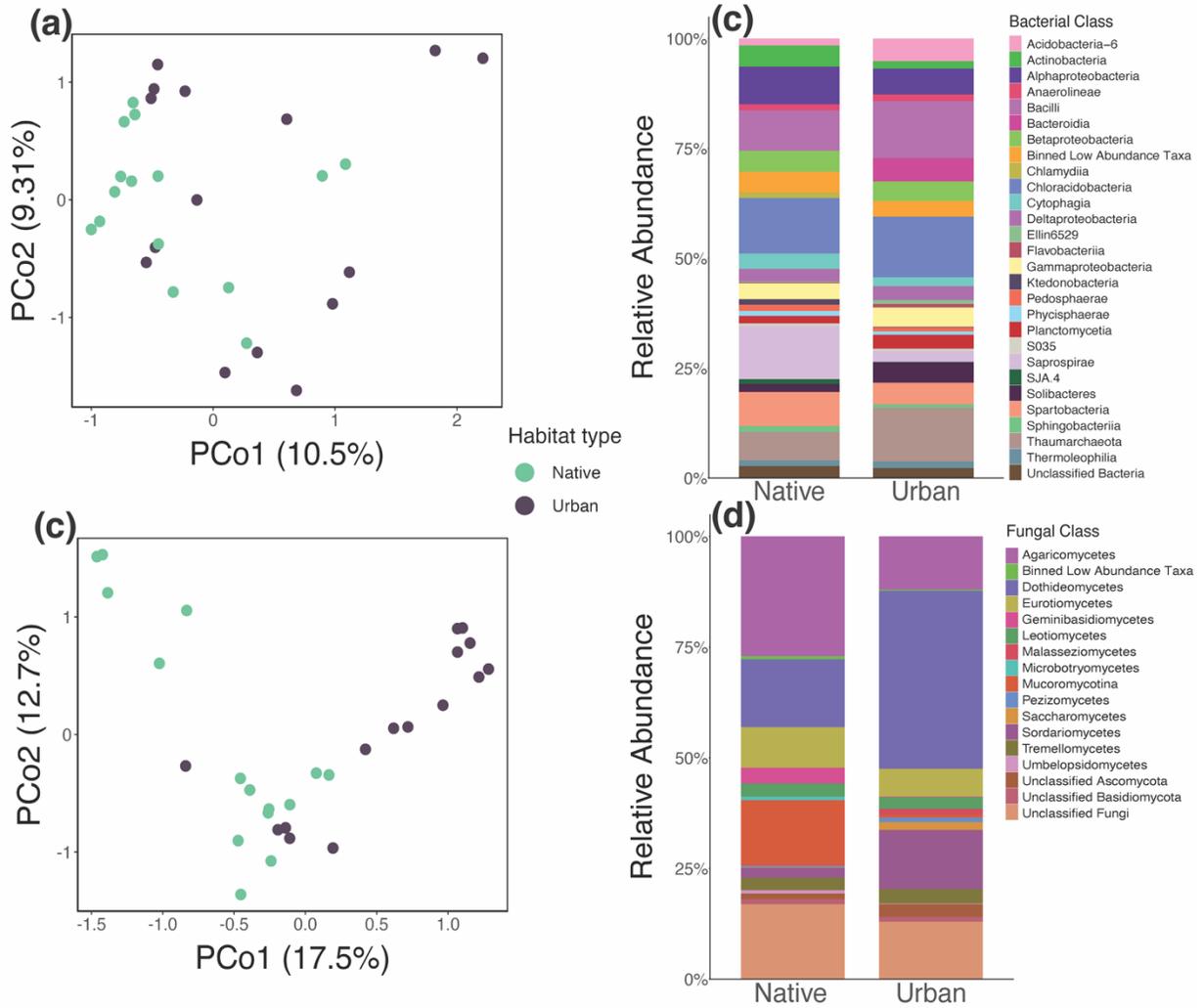
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1025 Figure 3.



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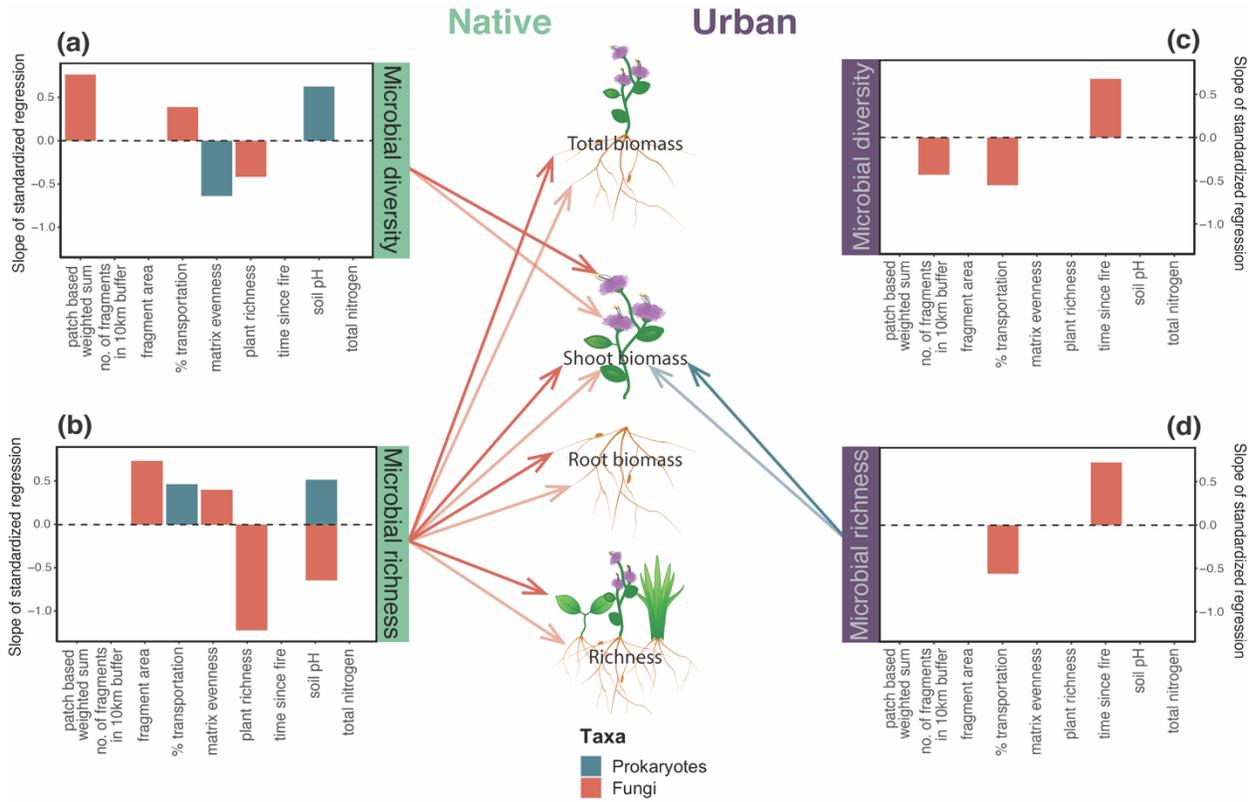
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1035 Figure 4.



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1046 Figure 5.

