

1 **Field reduction of ectomycorrhizal fungi has cascading effects on soil microbial
2 communities and reduces the abundance of ectomycorrhizal symbiotic bacteria**
3

4 Louis Berrios¹✉, Kabir G. Peay^{1,2}
5

6 ¹Department of Biology, Stanford University, Stanford CA, 94305, USA
7

8 ²Department of Earth System Science, Stanford University, Stanford CA, 94305, USA
9

10 ✉ Corresponding author's email: berriosl@stanford.edu
11 ✉ Co-author's email: kpeay@stanford.edu
12

13
14 **ORCID IDs:** Louis Berrios (0000-0003-3888-9928); Kabir G Peay (0000-0002-7998-7412)
15

16
17 **Keywords:** bacteria; ectomycorrhizal fungi (EcMF); bacteria-EcMF-plant interactions
18

19
20 **Competing Interests:** The authors declare no competing interests.
21

22
23 **Data Availability Statement:** Raw sequence data are available from NCBI Short Read Archive
24 (PRJNA1170419). The code used to analyze these datasets can be found at
25 https://github.com/LouisBerrios/Gadgil_Project.
26

27
28 **Benefits-Sharing Statement:** Benefits from this research will be gained from depositing our
29 data and results on public databases.
30

31
32 **Funding Statement:** This work was funded, in part, by an NSF PRFB grant 2109481 awarded
33 to LB and NSF DEB Awards 1845544, 2021478 to KGP. This work was also funded by a
34 Stanford Doerr School of Sustainability Discovery Grant awarded to LB and KGP. The funders
35 did not contribute to the design of the experiments, data collection, analyses, decision to
36 publish, or the preparation of the manuscript.
37

46 **Abstract**

47 Specific interactions between bacteria and ectomycorrhizal fungi (EcMF) can benefit plant
48 health, and saprotrophic soil fungi represent a potentially antagonistic guild to these mutualisms.
49 Yet there is little field-derived experimental evidence showing how the relationship among these
50 three organismal groups manifests across time. To bridge this knowledge gap, we
51 experimentally reduced EcMF in forest soils and monitored both bacterial and fungal soil
52 communities over the course of a year. Our analyses demonstrate that soil trenching shifts the
53 community composition of fungal communities toward a greater abundance of taxa with
54 saprotrophic traits, and this shift is linked to a decrease in both EcMF and a common
55 ectomycorrhizal helper bacterial genus, *Burkholderia*, in a time-dependent manner. These
56 results not only reveal the temporal nature of a widespread tripartite symbiosis between
57 bacteria, EcMF, and a shared host tree, but they also refine our understanding of the commonly
58 referenced 'Gadgil effect' by illustrating the cascading effects of EcMF suppression and
59 implicating soil saprotrophic fungi as potential antagonists on bacterial-EcMF interactions.
60

61 **Introduction**

62 Bacteria and fungi represent two of the most dominant microbial groups in forest soils (Anthony
63 et al. 2023; Uroz et al. 2016), and their roles in forest microbiomes range from plant pathogen
64 suppression (Tahat et al. 2010; Weller et al. 2002) to biogeochemical cycling (Zhour et al.
65 2018), plant growth promotion (Hayat et al. 2010; Mohammadi et al. 2011), and inter-microbial
66 growth facilitation (Deveau et al. 2018). A major impediment to our understanding of forest
67 microbiomes, however, is that much of our mechanistic knowledge comes from laboratory
68 studies, which cannot capture the adaptive, complexity of forest soils (Tecon et al. 2019).
69 Similarly, most efforts have focused on the soil dynamics of one or two single microbial guilds,
70 such as bacteria, ectomycorrhizal fungi (EcMF), or saprotrophs (Awad et al. 2019; Hawkins et
71 al. 2023; Rousk J and Bååth 2011). In other cases, the focus has been to highlight distinct
72 community responses of soil bacteria and fungi, providing little insights into how these diverse
73 communities coalesce to exhibit paralleled, dependent responses (Li et al. 2020). These studies
74 have contributed key insights to our collective understanding of forest function, but they have
75 been unable to reconcile the potential interactions that occur between and among multiple
76 microbial guilds (Baldrian 2017).
77

78 One commonly referenced inter-guild interaction in forest soil ecology is the so-called 'Gadgil
79 effect' (Gadgil and Gadgil 1971,1975). It encompasses a fundamental process in forest ecology
80 (i.e., soil organic matter decomposition), whereby resource competition between EcMF and
81 saprotrophic fungi, for example, can drive changes in the rate of decomposition and thus affect
82 larger-scale soil biogeochemical cycles (Fernandez and Kennedy 2015). Efforts to understand
83 the mechanisms that govern the 'Gadgil effect' have proven informative to both our conceptual
84 and empirical models of how these microbial guilds coexist within forest microbiomes
85 (Fernandez et al. 2020; Smith and Wan 2019). Reports, for instance, have shown that in some
86 cases EcMF slow carbon cycling in forests by suppressing saprotrophic fungi (Averill and
87 Hawkes 2016), while others have shown that EcMF can drive priming effects that increase the
88 growth of soil saprotrophs (Choreño-Parra EM and Treseder 2024). Much research has
89 therefore demonstrated how EcMF and soil saprotrophic fungi interact, but little is known about
90 how bacteria factor into this equation (Lladó et al. 2017). Experiments primed to address the
91 anticipated global reductions in EcMF (Arnolds 1991; Querejeta et al. 2021; Steidinger et al.
92 2020) and the subsequent cascading effects on microbial communities (Berrios, Bogar et al.
93 2024; Delgado-Baquerizo et al. 2020) across time will be fundamental for enhancing the
94 accuracy of current Earth system models (Bonan and Doney 2018; Wieder et al. 2015).
95

96
97 Emerging evidence suggests that bacteria play a vital role in the success and maintenance of
98 ectomycorrhizal symbioses (Berrios, Venturini et al. 2024; Frey-Klett et al. 2007). For instance,
99 in a recent communication, we demonstrated that a small group of bacterial taxa can function as
100 indicators of plant host and EcMF health (Berrios et al. 2023) and even more recently
101 demonstrated that bacterial-EcMF interactions contribute to potential fungal pathogen
102 suppression (Berrios, Bogar et al. 2024). It remains unclear in the literature, however, how the
103 simultaneous reduction in EcMF and increase in soil saprotrophic fungi affect bacterial
104 communities – particularly bacterial genera that function as positive indicators of EcMF and
105 plant health. Moreover, how the potential interactions among these three microbial groups
106 manifest across time remains equally uncertain. For instance, it is likely that a reduction in
107 EcMF would allow soil saprotrophic fungi to predominate, since saprotrophs would then have
108 less competition for space and nutrients. But whether these cascading effects may also weaken
109 positive, synergistic interactions between EcMF and their helper bacteria over time remains
110 untested. Therefore, the extent of bacteria-EcMF symbiotrophy, the potential role of fungal
111 competitors, and the temporal scale that govern positive bacteria-EcMF interactions remain
112 unclear. Efforts to address these uncertainties will not only reveal the spatiotemporal dynamics
113 that govern a widespread tripartite symbiosis among bacteria, EcMF, and plants, but they will
114 refine our concepts surrounding the ‘Gadgil effect’ and lend insights to the how a predicted
115 global decline in EcMF may affect ecosystem functions (Steidinger et al. 2020; Willing et al.
116 2024).

117
118 To address these knowledge gaps, we conducted an experimental field manipulation in EcMF-
119 dominated forests across a one-year period and investigated how reductions to EcMF in forest
120 soils affected saprotrophic fungal communities and bacterial communities (with a targeted
121 analysis on positive indicators of EcMF and plant health). Specifically, we used a common soil
122 trenching approach to reduce the abundance of EcMF, which allowed us to test our hypotheses
123 that a reduction in EcMF would (1) have cascading effects on saprotrophic fungal communities
124 (particularly soil saprotrophs), (2) reduce the diversity of bacterial communities, and (3) alter the
125 spatiotemporal dynamics of EcMF helper bacteria (i.e., those that were previously identified as
126 positive indicators of EcMF abundance; see Berrios et al. 2023). Our data show that reducing
127 EcMF in forest soils benefits saprotrophic fungi, decreases bacterial richness, and drives down
128 the abundance of some EcMF helper bacteria in a time-dependent manner that also implicates
129 soil saprotrophic fungi as potential competitors of these bacteria-EcMF mutualisms. These
130 analyses therefore highlight the varying degrees of intimacy that exist between bacterial taxa,
131 EcMF, and soil saprotrophs – which refines our understanding of a widespread, positive
132 interaction among bacteria, EcMF, and plants and advances holistic concepts that surround the
133 ‘Gadgil effect’ and bacterial-fungal interactions.

134
135 **Methods**

136
137 **Field work and sampling strategy**

138 Field sampling was carried out at Point Reyes National Seashore (PRNS) in Marin County,
139 California, United States (38°04'N, 122°50'W) in October 2021, January 2022, April 2022, and
140 September 2022 (mean annual temperature = 16°C; mean annual precipitation = 91 cm). We
141 established three local sites in monodominant Bishop pine forests (determined by local
142 accessibility and forest structure) that were > 1 km apart to maximize spatial variability in fungal
143 communities. To reduce the abundance of ectomycorrhizal fungi, we established 10 paired plots
144 at each site (3 sites X 2 treatments X 10 replicates). Each plot pair consisted of two 1 m² plots
145 with one trenched plot (to reduce EcMF abundance) and one control plot (i.e., untrenched).
146 Trenching was initially done in June of 2021 to a depth of 40 cm using a large shovel to sever

147 root connections. These trenches were further maintained by retrenching with a handsaw once
148 every two weeks throughout the course of one year. Plot pairs were located 1 m from each
149 other and at least 5 m from any other plot pairs. Since severing roots can alter the influx of
150 carbon (Averill and Hawkes 2016), we began sampling microbial communities in October 2021
151 (~ 4 months after trenches were established). From this time, soil cores were collected from
152 each plot once every three months to capture seasonal variability and the temporal relationships
153 among soil microbes (20 cores X 3 sites X 4 sampling periods = 240 cores total). Corers were
154 surface sanitized between plots with 70% ethanol (EtOH), and soil washes were performed
155 between each sample collection to limit carryover. Soil cores were also taken at the furthest
156 point between paired trenched and untrenched plots to limit potential trenching effects on
157 untrenched plots. Each core was placed into a separate, clean plastic bag and stored at 4°C
158 within five hours of collection.

159 Since trenching can introduce unintended nutrient pulses, which could alter microbial
160 activity in soils (Averill and Hawkes, 2016), we controlled for these possible disturbance effects
161 by monitoring soil chemistry, soil water potential, and root ingrowth biomass at two control plots
162 per site – in addition to sampling microbial communities ~ 4 months after trenches were
163 established. These approaches together helped us (1) ensure that trenching soils did not cause
164 significant changes to measured soil chemistry and water availability parameters between
165 trenched and untrenched plots, (2) demonstrate the activity of ectomycorrhizal roots, and (3)
166 establish adequate time for the impact of these potential nutrient pulses to dissipate. Though
167 other trenching studies have tended to only investigate changes to water potential or total
168 nitrogen (Averill and Hawkes, 2016; Hansson et al. 2018; Siira-Pietikäinen et al. 2003), we
169 tested changes to 15 soil chemistry properties that include tests for changes to these typical
170 disturbance-related metrics. Given that EcMF are known to liberate carbon, nitrogen, and
171 phosphorus in most EcMF-dominated forests, our ability to disentangle whether soil
172 communities change in response to increased nutrients that EcMF provide or from new
173 colonizable structures (i.e., EcMF hyphae) is limited (See et al. 2022). Monitoring was done at
174 only two plots per site to avoid disturbing the litter decomposition bags that were installed in the
175 remaining eight plot pairs as part of a concurrent experiment. Soil chemistry was measured
176 using Plant Root Simulator (PRS™) probes equipped with anion and cation exchange resins
177 (Western Ag, Saskatoon, SK, Canada). Soil water content was monitored using a FieldScout
178 TDR 150 Soil Moisture Meter (Spectrum Technologies, Inc.), and root ingrowth was measured
179 by installing rigid mesh cores (20 cm depth) filled with pure sand. Root ingrowth cores were
180 harvested at the end of the year, and the biomass of ectomycorrhizal rootlets was recorded on
181 soil removed during core installation and in the core at the end of the experiment. All soil
182 chemistry, water potential, and root ingrowth biomass data can be found in Fig. S1.
183

184 **DNA extractions, library preparation, and amplicon sequencing**

185 We characterized microbial communities by first weighing a 0.25 g subsample of collected soil
186 material for DNA extractions, using quality controls appropriate for our system (Zinger et al.
187 2019). Specifically, our experiment included ten biological replicates sampled across two
188 treatment conditions (n=2, trenched vs. untrenched) at each of three sites (n=3) and collected at
189 four timepoint (n=4), for a total N = 240 (10 x 2 x 3 x 4). For each biological replicate we also
190 included two technical DNA extraction replicates that were pooled prior to PCR. After initial
191 collection in the field, 0.25 soil subsamples were stored at - 20°C until extractions were
192 performed, and all extractions were performed within one week of sample collection. DNA
193 extractions were conducted using a DNeasy PowerSoil Pro Kit according to the manufacturer's
194 specifications (Qiagen, Carlsbad, CA, USA). Extraction controls (i.e., negative controls) were
195 derived using Nanopure water as input extraction material. DNA extracts, including extraction
196 negative controls, PCR negative controls, and positive control from a previously published
197 synthetic mock community as a positive control for fungi (Palmer et al. 2018) and

198 ZymoBIOMICS Microbial Community Standard (ZYMO Research, USA) as a positive control for
199 bacteria were used as PCR templates to determine both fungal and bacterial community
200 composition. In a previous study we found that PCR replication had minimal effect on ecological
201 inference in fungal community metabarcoding (Smith and Peay 2014). Thus, we chose to use a
202 single technical PCR replicate to characterize each sample. First step PCR amplifications were
203 performed by mixing 1 μ l of template DNA, 5 μ l of MyTaqTM HS Red Mix, 3.2 μ l of PCR-grade
204 H₂O, and 0.4 μ l of each primer. The 515f (GTGYCAGCMGCCGCGTAA) / 806R
205 (GGACTACNVGGGTWTCTAAT) primer pair was used to amplify the V4 hypervariable region of
206 the bacterial 16S rRNA gene (Caporaso et al. 2011), and the KYO1
207 (CTHGGTCATTTAGAGGAASTAA) / KYO2 (TAGAGGAAGTAAAAGTCGTA) primer pair was
208 used to amplify the ITS1 subregion of the fungal rRNA operon (Toju et al. 2012). Extraction
209 controls and positive controls (using mock bacterial and fungal communities) were also used as
210 DNA templates for PCR amplifications. The PCR amplifications occurred in 96-well plates and
211 consisted of 35 cycles with a denaturation temperature of 95°C, an annealing temperature of
212 50°C (52.5°C for fungi), and an extension temperature of 72°C. Indexed tags for Illumina
213 sequencing were attached during a second step PCR amplification (Toju et al. 2016), which
214 consisted of eight cycles with identical temperature settings as the first step PCR. Gel
215 electrophoresis was used to confirm PCR products, and a magnetic bead purification method
216 using Sera-Mag SpeedBeads (MilliporeSigma, Munich, Germany) was used to clean PCR
217 products. A Qubit 3.0 fluorometer (Thermo Fisher, Waltham, MA, USA) was used to quantify
218 cleaned DNA. Lastly, we pooled our samples into two libraries at equimolar concentrations and
219 submitted them for 2 x 300 Illumina MiSeq sequencing at the Stanford Functional Genomics
220 Facility. Raw sequences are available from NCBI Short Read Archive (BioProject:
221 PRJNA1170419).

223 ***Bioinformatics and statistical analyses***

224 We received 9,148,012 demultiplexed reads (avg. per sample = 52,274) for our bacterial
225 dataset and 8,968,351 demultiplexed reads (avg. per sample = 49,541) for our fungal dataset.
226 After filtering, denoising, merging forward and reverse reads, and removing chimeric sequences
227 using the DADA2 workflow (Callahan et al. 2016), our bacterial dataset consisted of 3,697,587
228 reads (avg. per sample = 21,705), and our fungal dataset consisted of 5,211,759 reads (avg.
229 per sample = 29,470). The DADA2 workflow accuracy was evaluated by inferring and matching
230 mock community members (i.e., positive controls) to their expected sequences (residual error
231 rate = 0%). Taxonomic classifications were assigned to bacteria and fungi using the SILVA
232 (Quast et al. 2012) and UNITE (Nilsson et al. 2019) databases, respectively. Sequences
233 assigned to 'chloroplast,' 'mitochondria,' and 'archaea' were removed from our bacterial dataset.
234 For both our bacterial and fungal datasets, we removed taxa with less than 10 sequences, and
235 samples were rarefied to a read depth of 5,000 using the rarefy_even_depth function in the
236 phyloseq package in R (McMurdie and Holmes 2013). Since our negative controls contained
237 less than these sequence number cutoffs, they were removed during sub-setting and
238 rarefaction. After rarefaction, we retained 162 samples in our bacterial dataset and 159 samples
239 in our fungal dataset. Our bacterial dataset consisted of 3,480 amplicon sequence variants
240 (ASVs), and our fungal dataset consisted of 1,242 ASVs. Guild designations for fungi were
241 assigned at the genus level using the FungalTraits database (Pölme et al. 2020).

242 Statistical analyses were performed in R studio version 4.2.1 (Team RC). The phyloseq
243 package was used to derive alpha (α) and beta (β) diversity estimates (McMurdie and Homles
244 2013). The *microbiome* package was used to relativize our amplicon sequence abundance data
245 (Lahti and Shetty 2017). We tested for differences in bacterial (n=162) and fungal (n=159)
246 community composition between conditions (i.e., trenched versus untrenched soils) and as a
247 function of seasonal variation using permutational analyses of variance (PERMANOVA) as
248 implemented in the adonis2 function from the *vegan* package (Oksanen et al. 2007). The

249 betadisper function was used to test for homogeneity of variances. To investigate interactions
250 between bacteria and fungi, we used linear regression models to examine correlations between
251 the abundance of both bacteria and EcM fungi and bacteria and soil saprotrophic fungi from our
252 amplicon datasets. Since soil chemistry analyses demonstrated that trenching had no
253 statistically significant effect on soil moisture and only one of sixteen analyzed soil chemicals,
254 and our root ingrowth data demonstrated that trenching significantly restricted host roots for
255 EcMF (Fig. S1), the effects from trenching that we observed in our models are likely to be
256 primarily the result of biotic interactions. Our statistical models included trenching treatment,
257 sampling timepoint (i.e., seasonal variation), site location, and the interaction between each of
258 these categorical variables and the quantitative variables in question as predictor variables. For
259 models that sought to identify the statistical significance of tripartite interactions, we used
260 sequence abundance ratios since this approach integrates the direct relationship between
261 fungal guilds and bacterial sequences within and between samples by default. Each model was
262 generated with the lm function in the stats R package (Team RC). For partial regression plots,
263 we plotted the residuals of response variables (y-axis) by the residuals of all the co-variates that
264 contribute to the main predictor variable (x-axis). We determined the relative importance of
265 predictor variables in the final models as the r^2 contribution averaged over orderings among
266 regressors (the 'lmg' option in calc.relimpo from the relaimpo package; Grömping 2006). All
267 plots were generated using either base R or ggplot2 (Wickham and Wickham 2007). The code
268 used to analyze these datasets can be found at https://github.com/LouisBerrios/Gadgil_Project.
269

270 Results

272 *The effects of soil trenching and seasonality on fungal community composition*

273 Since trenching can introduce unintended soil disturbance (Averill and Hawkes 2016), we first
274 tested that our control plots did not differ substantially between trenching treatments. When we
275 analyzed soil chemistry, we found phosphorus concentrations were significantly elevated in
276 untrenched plots (relative to trenched plots) but that trenching had no significant effect on soil
277 moisture content or on the remaining 15 chemicals that were analyzed ($p > 0.05$; Fig. S1A-Q).
278 We also measured ectomycorrhizal root ingrowth masses and found that trenching effectively
279 reduced the amount of ectomycorrhizal colonized rootlets and had no significant effect on the
280 mass of other plant roots (Fig. S1R). Therefore, these control plot data suggest that detected
281 changes to microbial communities are likely not an artefact of our experimental set-up and that
282 our trenching approach effectively decreased the abundance of ectomycorrhizal fungi (EcMF).
283

284 To identify how soil trenching, seasonality, and the interaction between the two affected
285 the composition of soil fungal communities, we first derived α and β diversity estimates for whole
286 fungal communities. α -diversity (i.e., observed richness) estimates were marginally affected by
287 trenching ($F_{1,157} = 3.24$; $p = 0.073$) and statistically unaffected by season ($F_{3,155} = 1.78$; $p =$
288 0.154), and interactions between trenching and season were not observed (Table S1; Fig. S2).
289 The β -diversity of whole fungal communities was significantly changed by both trenching
290 (PERMANOVA: $F_{1,157} = 2.41$; $p = 0.004$) and seasonal variation (PERMANOVA: $F_{3,155} = 1.65$; $p =$
291 0.004), but no significant interaction between the two was observed (Table S2; Fig. S3A).
292 When we assessed individual changes to EcMF and saprotrophic fungi communities, however,
293 we found that trenching had a marginally significant effect on EcMF communities
294 (PERMANOVA: $F_{1,157} = 1.50$; $p = 0.068$; Table S3), with no seasonal effects detected
295 (PERMANOVA: $F_{3,155} = 0.904$; $p = 0.68$), whereas both trenching (PERMANOVA: $F_{1,157} = 3.16$; $p =$
296 0.003) and season (PERMANOVA: $F_{3,155} = 2.41$; $p = 0.001$) significantly changed saprotrophic
297 communities (Table S4; Fig. S3). Like whole fungal communities, no interaction between
298 trenching treatment and season was detected for either ectomycorrhizal or saprotrophic fungal
299 communities. In addition to these observed mean effects on composition, we also found
significant differences in compositional variance (i.e., sample to sample variability within

300 treatments) for whole fungal communities and saprotrophic fungal communities, whereby
301 heterogenous dispersions across trenching treatment and season indicated that untrenched
302 treatments and winter and spring seasons had the most compositional variance. In contrast,
303 EcMF communities were homogenously dispersed (Fig. S4). While PERMANOVA is relatively
304 unaffected by heterogeneity in dispersion (Anderson and Walsh 2013), the observed treatment
305 effects are likely due to changes in both mean and dispersion.

306 Next, we sought to understand how the relative abundance and taxonomic composition
307 of ectomycorrhizal and saprotrophic fungal communities changed, since soil trenching should
308 have significant effects on their abundance. On average, soil trenching decreased the relative
309 abundance of EcMF and increased the relative abundance of saprotrophic fungi (Fig. 1).
310 Analysis of variance (ANOVA) models further demonstrated that trenching and season had
311 significant effects on the abundance of these fungal guilds (Fig. 1B-C). However, like many of
312 our findings reported above, an interaction between trenching and season was not observed for
313 EcM (ANOVA: $F_{3,155} = 0.544$; $p = 0.653$) or saprotrophic fungal communities (ANOVA: $F_{3,155} =$
314 0.803 ; $p = 0.494$). In terms of taxonomy, a total of 30 EcM genera and 114 saprotrophic fungal
315 genera (i.e., dung, litter, soil, wood, and unspecified) were detected (Fig. S5; Table S5). We
316 observed variable inter-genus compositional shifts for EcMF communities (Fig. S4). Some
317 genera (e.g., *Hydnus* and *Inocybe*), for instance, were consistently suppressed in trenched
318 soils, while trenching increased the relative abundance of other genera (e.g., *Amanita* and
319 *Polyozellus*). For saprotrophic fungi, we focused our taxonomic analysis on soil saprotrophs
320 since these were the most abundant and trench-responsive saprotrophs in our dataset (Fig. 1A;
321 Table S5). Like EcMF genera, soil saprotrophs also displayed differing responses to trenching
322 at the genus-level (Fig. S5). The genus *Geminibasidium*, for example, remained more abundant
323 in trenched plots across each timepoint compared to untrenched plots, whereas *Saitozyma* –
324 the most dominant soil saprotrophic genus – was continually suppressed by soil trenching.
325 Collectively, these fungal community sequence data demonstrate that our soil trenching
326 methods effectively decreased the relative abundance of EcMF and increased the relative
327 abundance of saprotrophic fungi – though inter-guild and inter-genus variations were observed.
328

329 ***Bacterial community compositional changes caused by soil trenching and seasonality***

330 We derived estimates of bacterial richness (i.e., α -diversity) and β -diversity to understand how
331 soil trenching, seasonality, and the interaction between these two variables impacted bacterial
332 community structure. Both trenching and season had significant effects on bacterial richness,
333 whereby richness was typically highest in untrenched soils (Fig. 2A; Table S6) and followed a
334 sinusoidal curve over time on average, with peaks in winter and summer (Fig. S6). When we
335 assessed changes to β -diversity, we found that trenching (PERMANOVA: $F_{1,160} = 1.82$; $p =$
336 0.019) and season (PERMANOVA: $F_{3,158} = 2.88$; $p = 0.001$) had significant effects – though no
337 interaction between these variables was detected (PERMANOVA: $F_{3,158} = 0.916$; $p = 0.667$). A
338 beta dispersion tests suggested that the communities compared between trenching conditions
339 were dispersed homogeneously (Fig. S7A), whereas communities across seasons were
340 dispersed heterogeneously with the greatest distance to centroids occurring during winter and
341 spring ($F_{3,158} = 4.98$; $p = 0.002$; Fig. S7; Table S7B).

342 Taxonomically, trends at the phyla-level and finer hierarchical levels were evident. For
343 instance, the phylum Acidobacteria was suppressed by soil trenching, whereas – relative to
344 untrenched plots – the phylum Verrucomicrobiota was elevated (Fig. 2C). Other phyla exhibited
345 seasonal trends but exhibited no obvious changes due to trenching (e.g., Acinobacteriota
346 elevated in Fall and Summer sampling periods and Bacteroidota elevated in Winter and Spring).
347 At the genus level, we observed 349 bacterial genera in our dataset. To narrow our genus-level
348 investigation, we targeted 11 bacterial genera that can be indicators of EcMF relative
349 abundance (Berrios et al. 2023). Most of these genera, however, did not show strong responses
350 to soil trenching and seasonal variation, but a few genera appeared to be either depleted (i.e.,

351 *Burkholderia*), enriched (i.e., *Acidibacter*), or seasonally affected (i.e., *Puia*) due to trenching
352 (Fig. S8). Together, these data converge on the notion that reducing EcMF in forest soils
353 decreases bacterial richness and alters the dynamics of a small group of bacteria.
354

355 **Relationships between EcMF, soil saprotrophs, bacterial richness, and *Burkholderia* species**
356 Because soil trenching had significant effects on EcMF, saprotrophic fungi (particularly soil
357 saprotrophs), and bacterial communities (Fig. 1 and Fig. 2), we wanted to further investigate the
358 links between these microbes. To these ends, we first constructed linear regression models that
359 used EcMF relative sequence abundance as predictor variables for bacterial richness. We found
360 that EcMF sequence abundance correlated positively with bacterial richness and functioned as
361 significant predictors of bacterial richness ($F_{1,80} = 4.88$; $p = 0.03$), though trenching alone
362 explained much of the predictive strength in our model ($F_{1,80} = 7.51$; $p = 0.008$; Fig. 3A; Table
363 S8). In addition, we detected an interaction between EcMF abundance and season/sampling
364 timepoint ($F_{3,78} = 5.06$; $p = 0.003$). These observations, paired with the fact that (1) soil
365 saprotrophs and EcMF compete for resources and resultantly drive each other's spatiotemporal
366 dynamics (Fernandez and Kennedy 2016) and (2) soil saprotrophs were both the most
367 abundant saprotrophic fungal guild in our dataset and were most sensitive to seasonal and
368 trenching effects (Fig. 1A), led us to investigate whether soil saprotroph abundance could
369 predict bacterial richness. Our linear regression analysis demonstrated that soil saprotrophic
370 fungi correlated negatively with bacterial richness and that both saprotroph sequence
371 abundance and the effect of trenching were significant predictors of bacterial richness (Fig. 3B).
372 In each of our models (Fig. 3), season also appeared to have a significant effect, with detectable
373 interactions between season and EcMF abundance (Table S8) and between season and soil
374 saprotroph abundance (Table S9). Site location was not a significant factor for either model.
375 Because relative abundance data can present some interdependence between measured
376 variables, we next used the ratio of EcMF and soil saprotrophic fungi relative abundances as
377 predictors of bacterial richness. Our regression model (Table S10) demonstrated that indeed the
378 ratio of EcMF and soil saprotrophs functioned as a significant predictor of bacterial richness
379 ($F_{1,80} = 14.37$; $p = 0.0003$), and the individual r^2 contributions of season ($r^2 = 11\%$), EcMF:soil
380 saprotroph ratios ($r^2 = 7.5\%$), and EcMF:soil saprotroph ratios as a function of time ($r^2 = 11\%$)
381 totaled roughly that of what we observed in our regression models without ratios.

382 To understand how soil trenching and season alter the relationships between EcMF and
383 bacteria in the genus *Burkholderia* – which we have found to be a positive indicator of tree
384 seedling and EcMF health in this system (Berrios et al. 2023) – we first examined the degree to
385 which significant decreases in the relative abundance of *Burkholderia* species were attributable
386 to soil trenching and season. We found that a significant decrease in *Burkholderia* species
387 manifested after one year post trench establishment (Fig. 4A). To further enhance our
388 understanding of these relationships, we constructed a linear regression model (predictor
389 variables = *Burkholderia* relative sequence abundance, trenching treatment, sampling season,
390 and sampling site; response variable = EcMF relative sequence abundance). Though site
391 location was a significant predictor of EcMF sequence abundance, our model indicated that
392 *Burkholderia* sequence abundance was also a significant predictor (Fig. 4B). However, since the
393 effect of trenching has potentially cascading effects on the abundance of soil saprotrophic fungi
394 and *Burkholderia* sequences over time (Fig. 3), we built in soil saprotroph, sampling
395 timepoint/season, and sampling site interaction terms (Table S11). Indeed, this refinement
396 improved overall model strength by 14% (i.e., increased r^2 from 35% to 49%), detected a
397 statistical interaction between *Burkholderia* and soil saprotrophs ($F_{1,80} = 14.10$; $p = 0.0003$), and
398 reiterated the time-dependent response we observed (Fig. 4A) for the potential interactions
399 between *Burkholderia* and EcMF ($F_{3,78} = 4.10$; $p = 0.009$). Therefore, these data sum to illustrate
400 that a reduction in EcMF alters the spatiotemporal dynamics of soil saprotrophs, which has

401 varied effects on bacterial communities and elicits a time-dependent response for a common
402 bacterial symbiont of EcMF and pine trees.

403 404 Discussion

405 How we have come to understand the complexity of forest microbiomes has been derived
406 primarily through the observed interactions of one or two microbial guilds – an approach that
407 has both answered and left open many questions. Multiple lines of evidence, for instance, have
408 shown that competition between EcMF and saprotrophic fungi can lead to changes in soil
409 nutrient economies (Averill and Hawkes 2016; Bending 2003; Gadgil and Gadgil 1971,1975).
410 This phenomenon, commonly called the ‘Gadgil effect,’ can have substantial impacts on forest
411 functions (Fernandez and Kennedy 2016; Simard et al. 1997), but no experimental field
412 evidence has shown how the competition between EcMF and saprotrophic fungi may impact the
413 spatiotemporal dynamics of soil bacteria – particularly bacteria that support EcMF and plant
414 health. With this knowledge gap in mind, we conducted a year-long trenching experiment in a
415 tractable forest system (Peay and Bruns 2014; Peay et al. 2010) to understand how a reduction
416 in EcMF alters the spatiotemporal dynamics of bacteria, EcMF, and saprotrophic fungi. Our data
417 demonstrate that a reduction in EcMF not only alters the community composition of bacteria and
418 fungi, but this reduction increases soil saprotrophic fungi and significantly reduces the relative
419 abundance of ectomycorrhizal helper bacteria in the genus *Burkholderia* (Berrios et al. 2023)
420 after about one year (see Fig. 5 for a conceptual diagram). This longitudinal field study therefore
421 refines our view of a long-standing phenomenon that occurs in EcMF-dominated forest soils
422 (i.e., the ‘Gadgil effect’), providing a nuanced understanding of the factors that contribute to
423 symbioses between bacteria and fungi.
424

425 Consistent with previous findings (Fernandez et al. 2020), our data show that soil trenching
426 drives down the relative abundance of EcMF and enriches for soil saprotrophic fungi (Fig. 1; Fig.
427 S1). How this occurs has received considerable attention (Lang et al. 2021; Malik 2019;
428 Sterkenberg et al. 2010), and the degree to which soil disturbance, changes in soil moisture
429 content, the release of organic substrates post-trenching, and direct interactions between EcMF
430 and soil saprotrophs influence the ‘Gadgil effect’ have been difficult parameters to untangle in
431 complex systems (Averill and Hawkes 2016; Brzostek et al. 2015; Lindahl et al. 2010). Although
432 our analyses cannot separate the magnitude and direction for each of these potential variables,
433 our trenching controls suggest that the observed shifts in community structure are not simply
434 artefacts of treatment induced changes in the soil environment (Fig. S1), and our regression
435 models likewise capture the effects of trenching as a function of seasonal variability (Fig. 1; Fig.
436 S3-S5; Table S1-S4). The detection of seasonal variability is key here because it suggests that
437 we detected changes in microbial activity and not simply relic DNA. In addition, experiments
438 primed to address the disturbance effects of soil trenching – perhaps the largest potential
439 confounder of these experiments – have shown that the often-used alternative method (i.e.,
440 using mesh bags to exclude EcMF) generates similar results to those of soil trenching
441 experiments (Averill and Hawkes 2016). Therefore, it is unlikely that the observations we report
442 herein are simply artefacts of soil disturbance – though decreased soil phosphorus availability in
443 trenched plots cannot be omitted as a potential contributor. In line with this notion is the fact that
444 our analyses are unable to detect whether and to what degree organic matter that EcMF liberate
445 in soils, hyphal biomass as a new colonizable niche, or a combination of the two facilitated the
446 changes to microbial communities (See et al. 2022). Further, the mechanisms that drive the
447 standard view of the ‘Gadgil effect’ are largely out of the scope of this communication. For
448 example, it is likely that the inter-genus compositional changes of EcMF are linked to
449 competition-colonization trade-offs within the EcMF guild (Fig. S5), and these interactions could
450 play a role in how soil saprotrophic fungi respond to a reduction in EcMF. Nonetheless, we
451

452 instead discuss the intra- and inter-guild changes of fungal communities only to the extent that
453 allows us to situate our primary focus (i.e., the spatiotemporal dynamics of bacterial
454 communities) in the context of inter-guild fungal competition and bacterial-fungal interactions.
455

456 Though seasonality affected the composition of whole fungal communities and saprotrophic
457 fungal communities, observed shifts in EcMF communities were less evident (Table S2-S4; Fig.
458 S3-S4). Some reports have noted significant seasonal effects on EcMF – both in terms of
459 mycelial biomass in soils (Wallander et al. 2001) and host root colonization (Soudzilovskaia et
460 al. 2015; Swaty et al. 1998). Others, however, have observed marginal seasonal effects on
461 EcMF communities, noting instead strong effects of season on saprotrophic fungal communities
462 (Gorfer et al. 2021) – such as those that we observed during the spring for soil saprotrophic
463 fungi, which may correspond to a concomitant shift to warmer, wetter weather (Fig. 1). These
464 discrepancies are likely a product of local, climatic and edaphic factors (Beidler et al. 2023;
465 Gong et al. 2022), and the lack of detectable seasonal effects on EcMF communities in our
466 system likely reflects the relatively consistent seasonal nature of the environment in an
467 evergreen, coastal Mediterranean ecosystem. The significant degree of temporal shifts that we
468 observe for other fungal guilds, however, may be a result of comparatively fast growth rates
469 (i.e., slower for EcMF and higher for other guilds), which may also factor into the potential
470 competitive play between EcMF and other fungal guilds. Therefore, it is not too surprising that
471 we do not observe large, seasonal shifts of EcMF composition. But why EcMF abundances
472 were greater in colder months than warmer months (Fig. 1A) is likely due to the greater
473 physiological activity of Bishop pine trees during the colder months in California, which may be
474 caused, in part, by greater precipitation. It's also known that EcMF communities can have
475 species-specific responses to temperature (Koizumi and Nara 2020), and our data illustrate that
476 EcMF also respond differentially at the genus level (e.g., *Cortinarius* exclusively detected during
477 fall and winter; see Fig. S5). These intra-guild dynamics are also likely driven by similar intra-
478 guild dynamics that shape saprotrophic fungi, which together ultimately shape the inter-guild
479 changes that we observed. Nevertheless, a closer examination of how seasonality impacts
480 competition-colonization tradeoffs within EcMF communities could help clarify these presumably
481 context-dependent observations (Kennedy et al. 2011; Moeller and Peay 2016; Smith et al.
482 2018).
483

484 When we investigated how soil trenching and seasonality affected bacterial communities, we
485 found that these factors had variable effects (Fig. 2; Fig. S6; Fig. S7). Trenching decreased
486 bacterial richness, and seasonal effects were observed, but we did not observe a detectable
487 interaction between the two factors (Fig. 2A; Table S6). These observations are consistent with
488 recent field (Berrios et al. 2023) and greenhouse (Berrios, Bogar et al. 2024) studies, and the
489 absence of an interaction between soil trenching and sampling season suggests that seasonal
490 variability neither intensifies nor suppresses the effects we observed from trenching. In terms of
491 bacterial β -diversity, seasonal variation was the strongest and most significant driver, and – like
492 our observations on bacterial richness – we did not detect an interaction between soil trenching
493 and sampling season (Fig. 2B; Fig. S7; Table S7). Previous reports have shown that EcMF
494 enrich a small group of bacterial taxa (Berrios et al. 2023; Berrios, Bogar et al. 2024; Bowen GD
495 and Theodorou 1979; Nguyen and Bruns 2015), so it is therefore not surprising that only small,
496 community-wide differences in bacterial composition were detected (Fig. 2; Fig. S7). In line with
497 this notion, taxonomic shifts in bacterial communities were also moderate (Fig. 2C). On
498 average, the effect of trenching decreased the relative abundance of acidophilic (e.g.,
499 *Acidibacter*) or acidotolerant (e.g., *Burkholderia*) bacterial taxa and elevated the relative
500 abundance of Verrucomicrobiota (Fig. 2C; Fig. S8). These results are likely associated with a
501 decrease in EcMF and an increase in saprotrophic fungi in trenched plots (Fig. 1), since EcMF
502 tend to foster a lower pH, and saprotrophic fungi favor a relatively more neutral pH (Yamanaka

503 2003). However, they do not fully capture the inter-phyla dynamics of bacteria since some
504 Verrucomicrobiota are also acidophilic (Dunfield et al. 2007). In addition, many previously
505 reported indicator bacterial taxa of EcMF relative abundance (Berrios et al. 2023; Berrios, Bogar
506 et al. 2024; Nguyen and Bruns 2015) were relatively unaffected by soil trenching (Fig. S8),
507 suggesting that (1) many of these taxa may thrive on the necromass of EcMF (Maillard et al.
508 2024); (2) there was still sufficient EcMF available for them to find a niche, and (3) soil
509 saprotrophs play a significant role in the spatiotemporal dynamics of these bacterial-EcMF
510 interactions (Fig. 3).

511 The bacterial genus *Burkholderia* has been linked to enhanced ectomycorrhizal fungal
512 colonization and pine seedling growth (Berrios et al. 2023), and it has been a temporally stable
513 bacterium in Bishop pine forests for decades (Nguyen and Bruns 2015). Therefore, we sought
514 to uncover how decreasing EcMF in forest soils affected the abundance of this genus over time.
515 Our data demonstrated that soil trenching decreased the relative abundance of *Burkholderia*
516 over time (Fig. 4A) and that *Burkholderia* sequence abundance functioned as a positive
517 predictor of EcMF sequence abundance (Fig. 4B), which aligns with previous field and
518 greenhouse observations (Berrios et al. 2023; Nguyen and Bruns 2015). Past analyses,
519 however, were unable to reconcile the spatiotemporal dynamics of this ectomycorrhizal helper
520 bacteria and moreover had not situated these findings in a longitudinal study within the complex,
521 adaptive environment of forest soils. Our analyses here not only demonstrate that the reduction
522 of EcMF (and likely active, healthy plant host roots) drives down the abundance of *Burkholderia*
523 species (Fig. 4), but they implicate soil fungal saprotrophs as a potentially antagonistic factor for
524 the proliferation of these species ($F_{1,80} = 14.10$; $p = 0.0003$; Table S8). Why it took a year before
525 trenching effects significantly impacted these bacteria may be explained by the fact that our soil
526 trenching methods can be viewed as a 'knock-down' rather than a 'knock-out' of EcMF in this
527 system. The most parsimonious explanation is therefore that *Burkholderia* species were still
528 deriving benefits from EcMF in trenched plots, though direct necromass uptake is unlikely for
529 this genus given a lack of detectable chitinase activity in a previous transcriptomic study
530 (Berrios et al. 2023) and a general lack of genes that encode chitinases in many publicly
531 available *Burkholderia* genomes (Berrios 2022a). In contrast, a less conservative explanation is
532 that the suppression of EcMF in trenched plots derepressed soil saprotrophic fungi, which
533 resultantly attenuated the tight association between EcMF and *Burkholderia* through resource
534 competition. The exact mechanism for this bacterial-fungal competition is unclear, but a recent
535 analysis demonstrated that soil fungi are up to four times more efficient at obtaining complex
536 compounds in the soil (Wang and Kuzyakov 2024). Though direct biotic warfare (e.g., antibiotic
537 production) may also explain these observations. Nevertheless, these findings do clarify the
538 temporal relationships among a previously reported widespread tripartite relationship among
539 bacteria, EcMF, and a shared plant host (Berrios et al. 2023), but further studies that use
540 additional manipulative and quantitative approaches would help strengthen these observations.
541

542 The field of bacterial-fungal interactions continues to emerge as an important yet neglected sub-
543 discipline in microbial ecology and holds promise for both fundamental and applied research in
544 forest ecology (Artursson et al. 2006; Berrios 2022b; Berrios, Venturini et al. 2024; Frey-Klett et
545 al. 2011; Manfredini et al. 2021). These observations, coupled with the critical role of EcMF for
546 terrestrial carbon cycling (Hawkins et al. 2023) and the anticipated global warming-fueled rise in
547 plant pathogenic soil saprotrophs (Delgado-Baquerizo et al. 2020), suggest that more efforts to
548 understand the interactions between bacteria, EcMF, and saprotrophs are warranted and timely.
549 In this study, we leverage a tractable forest system to understand the spatiotemporal response
550 of bacterial communities and previously reported ectomycorrhizal helper bacteria to a reduction
551 of EcMF. Our data demonstrate that a reduction in EcMF has cascading effects on saprotrophic
552 fungal communities and contributes to a time-dependent reduction in the genus *Burkholderia* –

554 bacteria that function as positive indicators of EcMF and tree health (Berrios et al. 2023). With
555 predictions that point to EcMF diversity loss (Arnolds 1991; Steidinger et al. 2020), a
556 predominance of saprotrophic fungi (Delgado-Baquerizo et al. 2020), and climate-related
557 mismatches between soil microbiota and their plant hosts (Rudgers et al. 2020), it will become
558 even more essential to understand the temporal dynamics that govern multipartite relationships
559 between plants and microbes (Berrios and Rentsch 2022; Gschwend et al. 2021). Our data
560 collectively highlight that a loss of EcMF will not only cause a loss of bacterial biodiversity, but it
561 will also cause a loss of bacteria that benefit plant hosts and symbiotic fungi – which may further
562 exacerbate competition between EcMF and soil saprotrophic fungi. These observations
563 therefore begin to clarify the temporal dynamics of common EcMF-bacteria interactions and
564 contribute a nuanced view of how bacterial communities factor into the potential interactions
565 surrounding the so-called ‘Gadgil effect.’

566
567 It will be essential moving forward to understand how bacteria – especially mycorrhizal helper
568 bacteria – both facilitate and interfere with competition between EcMF and saprotrophic fungi.
569 From our data, it is clear that the degree of symbiotropy between bacteria and EcMF differ
570 across space, time, and taxonomic designation (Fig. 2-4), which could potentially impact the
571 magnitude, direction, type, and outcome of interactions among various bacterial, fungal, and
572 plant groups. A major consideration therefore is the focal metric or proxy being measured and
573 what those system-wide outcomes might be at varying scales. In our system, EcMF are key for
574 the establishment and maintenance of the host tree, and we have identified bacterial taxa that
575 support ecological interactions of ectomycorrhizal symbiosis. But how these multipartite
576 relationships control processes that link carbon budgets, fire ecology, and climate resilience,
577 and how they can be applied to further conservation efforts remains less apparent (Bisbing et al.
578 2023; Taylor et al. 2020). Though our data cannot identify, for example, changes to soil
579 respiration or carbon deposition rates, efforts to gauge these responses should be undertaken
580 because they could serve as important metrics to determine system-wide outcomes of
581 multipartite relationships (Averill and Hawkes, 2016). Given the innumerable context
582 dependencies that shape plant-soil interactions, it is likely that the level of ‘benefit’ and
583 ‘detiment’ that soil microbial communities provide and incur, respectively, changes over time
584 and with respect to community context (Berrios 2022b; Fukami 2015). Experiments that
585 investigate the eco-evolutionary relationships among bacteria, EcMF, saprotrophs, and host
586 trees will help organize these context dependencies and clarify our predictions of how forests
587 may change in the future. How we then integrate this information into Earth system models
588 (Bonan and Doney 2018) and leverage them for forest conservation efforts remains a primary
589 research agenda that could reveal key insights into soil nutrient economies (Wieder et al. 2015),
590 atmospheric exchange processes (Sutton et al. 2007), and land management strategies
591 (Pongratz et al. 2018) as we traverse further into the Anthropocene.

592
593

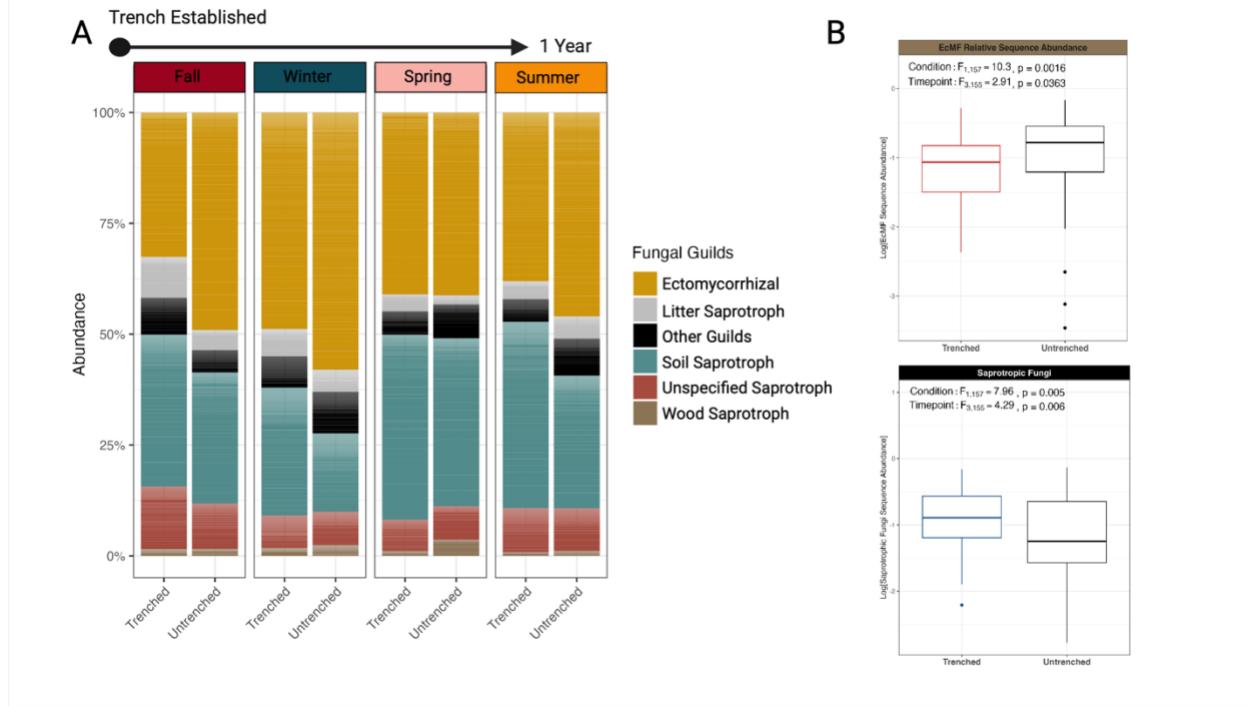
594 **Funding/Acknowledgements:** This work was funded, in part, by an NSF PRFB grant 2109481
595 awarded to LB and NSF DEB Awards 1845544, 2021478 to KGP. This work was also funded by
596 a Stanford Doerr School of Sustainability Discovery Grant awarded to LB and KGP. The funders
597 did not contribute to the design of the experiments, data collection, analyses, decision to
598 publish, or the preparation of the manuscript.

599

600 **Author Contributions:** LB designed the experiments, collected and processed the samples,
601 analyzed the data, and wrote the manuscript. KGP established the soil trenches and maintained
602 them throughout the experiment with the help of LB. KGP and LB edited the manuscript.

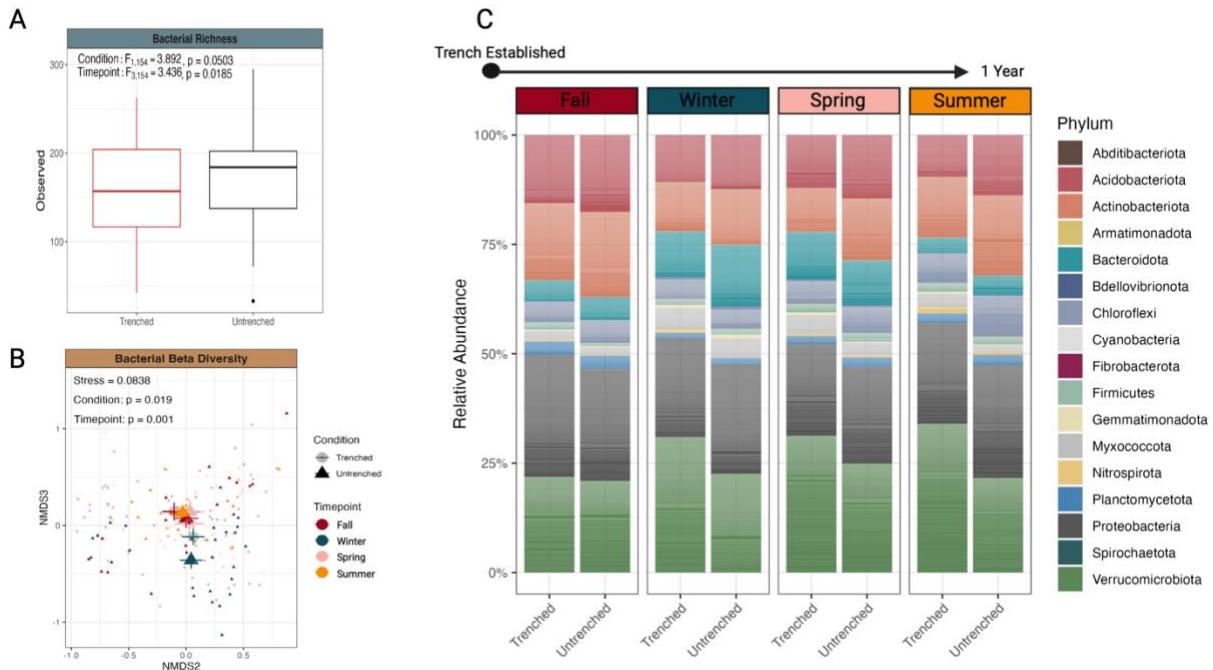
603
604

605 **Figures and Legends**



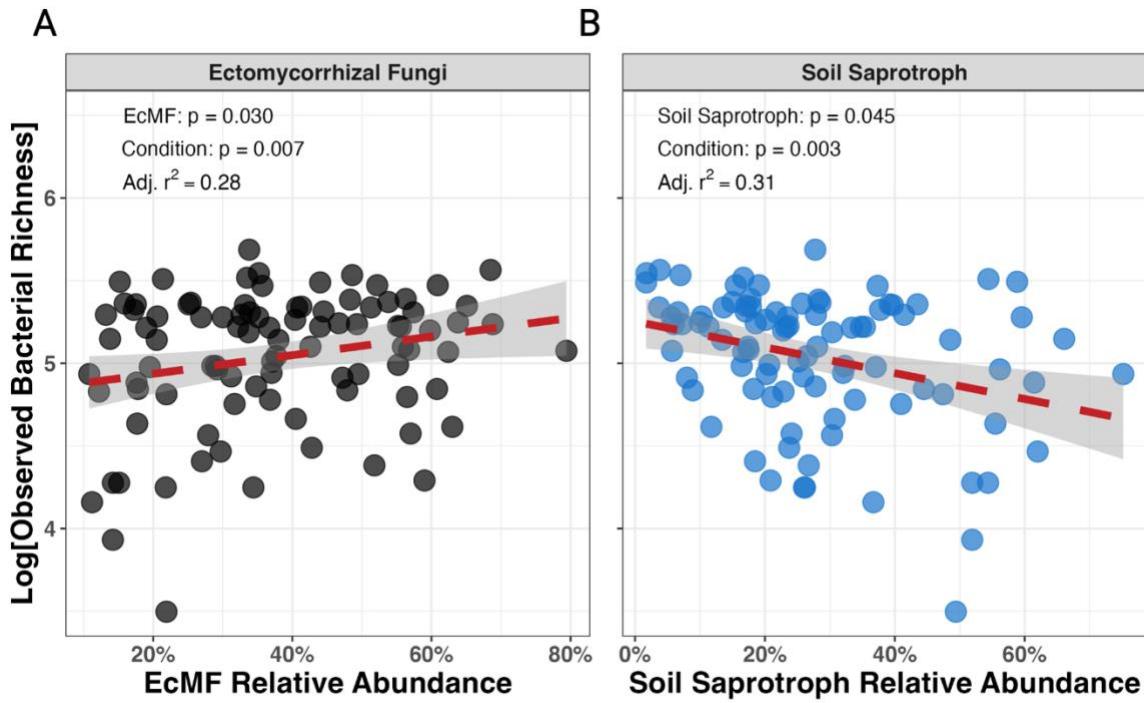
606
607
608 **Fig. 1. Compositional changes of fungal taxa. (A)** Stacked bar plot showing the
609 compositional changes of fungal guilds across time. The plot is faceted by season (timepoint of
610 sampling) and grouped by condition (i.e., trenched or untrenched). **(B)** Ectomycorrhizal fungi
611 (EcMF) relative sequence abundance (top) and saprotrophic fungi relative sequence abundance
612 (bottom) comparisons between trenched and untrenched soils. ANOVA model statistic outputs
613 are shown in each panel.

614
615
616
617



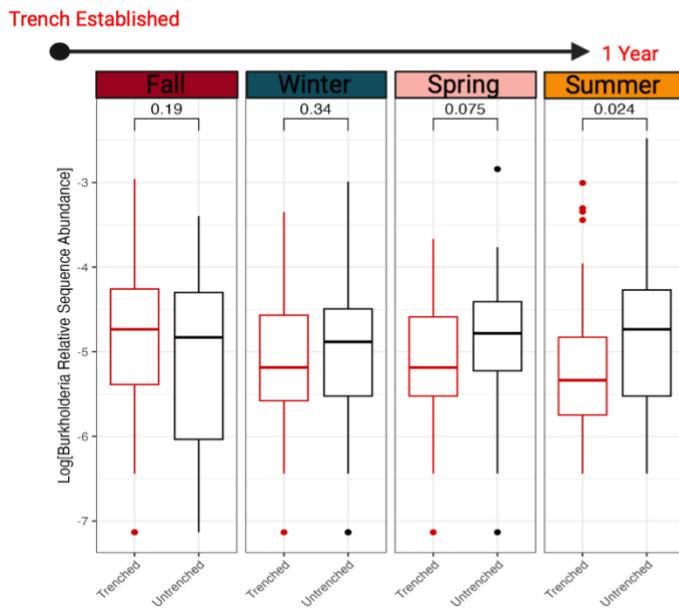
618
619
620
621
622
623
624
625
626
627
628
629

Fig. 2. Compositional changes of bacterial taxa. (A) Observed bacterial richness (i.e., α -diversity) comparisons between trenched and untrenched soils. ANOVA model outputs are shown for the effect of condition (i.e., trenched or untrenched soils) and sampling timepoint (i.e., season). See Table S6 for the complete model output. (B) Non-metric multidimensional scaling (NMDS) analysis that illustrates the changes of bacterial β -diversity as a function of condition and sampling timepoint. Centroids for each condition are shown with emanating standard error lines, and PERMANOVA model statistics are shown in the top left corner. See Table S7 for the complete model output. (C) Stacked bar plot showing the compositional changes of bacterial phyla across time. The plot is faceted by season (timepoint of sampling) and grouped by condition. Note that the focal bacterial genus in our analysis hereafter (i.e., *Burkholderia*) belongs to the phylum Proteobacteria.

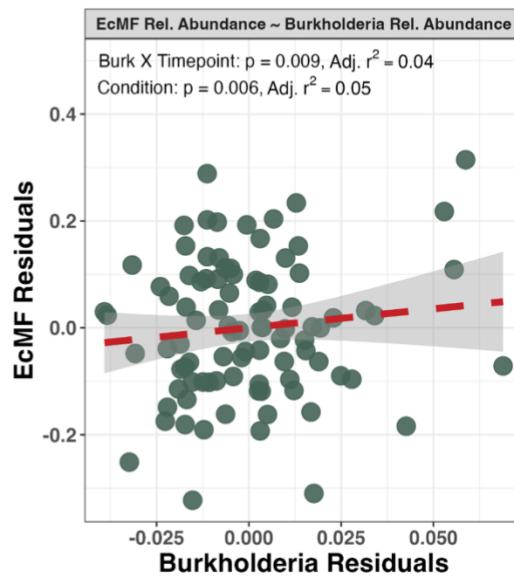


630
 631 **Fig. 3. Relationships between ectomycorrhizal fungi (EcMF), soil saprotrophs, and**
 632 **bacterial richness.** (A) Correlation between observed bacterial richness and EcMF relative
 633 sequence abundance. Linear regression model statistics are shown. (B) Correlation between
 634 observed bacterial richness and soil saprotrophs relative sequence abundance. The 95%
 635 confidence interval is shown in gray, and linear regression model statistics are shown. Full
 636 model outputs are shown in Table S8 and Table S9.
 637
 638
 639

A



B

640
641

642 **Fig. 4. Relationships between ectomycorrhizal fungi and the bacterial genus**
643 ***Burkholderia*.** (A) Log-transformed relative sequence abundance of amplicon sequence
644 variants (ASVs) matched to the genus *Burkholderia*. Abundance values are grouped by soil
645 condition (i.e., trenched and untrenched) and further faceted by sampling season (as observed
646 in Marin County, CA, USA). P values for t-test statistics are shown above each pairwise
647 comparison. Statistics were performed in R using the `stat_compare_means()` function in the
648 `ggbpbr` package (Kassambara 2023). (B) Linear regression analysis investigating the
649 relationship between EcMF relative sequence abundance and *Burkholderia* sequence
650 abundance. Residuals from linear regression outputs are shown. Partial regression r^2 values
651 were derived using the `relaimpo` package in R and are shown for each listed predictor value.
652 The 95% confidence interval is shown in gray, and model output statistics are shown in the
653 panel. See Table S10 for the complete model output statistics.

654

655

656

657

658

659

660

661

662

663

664

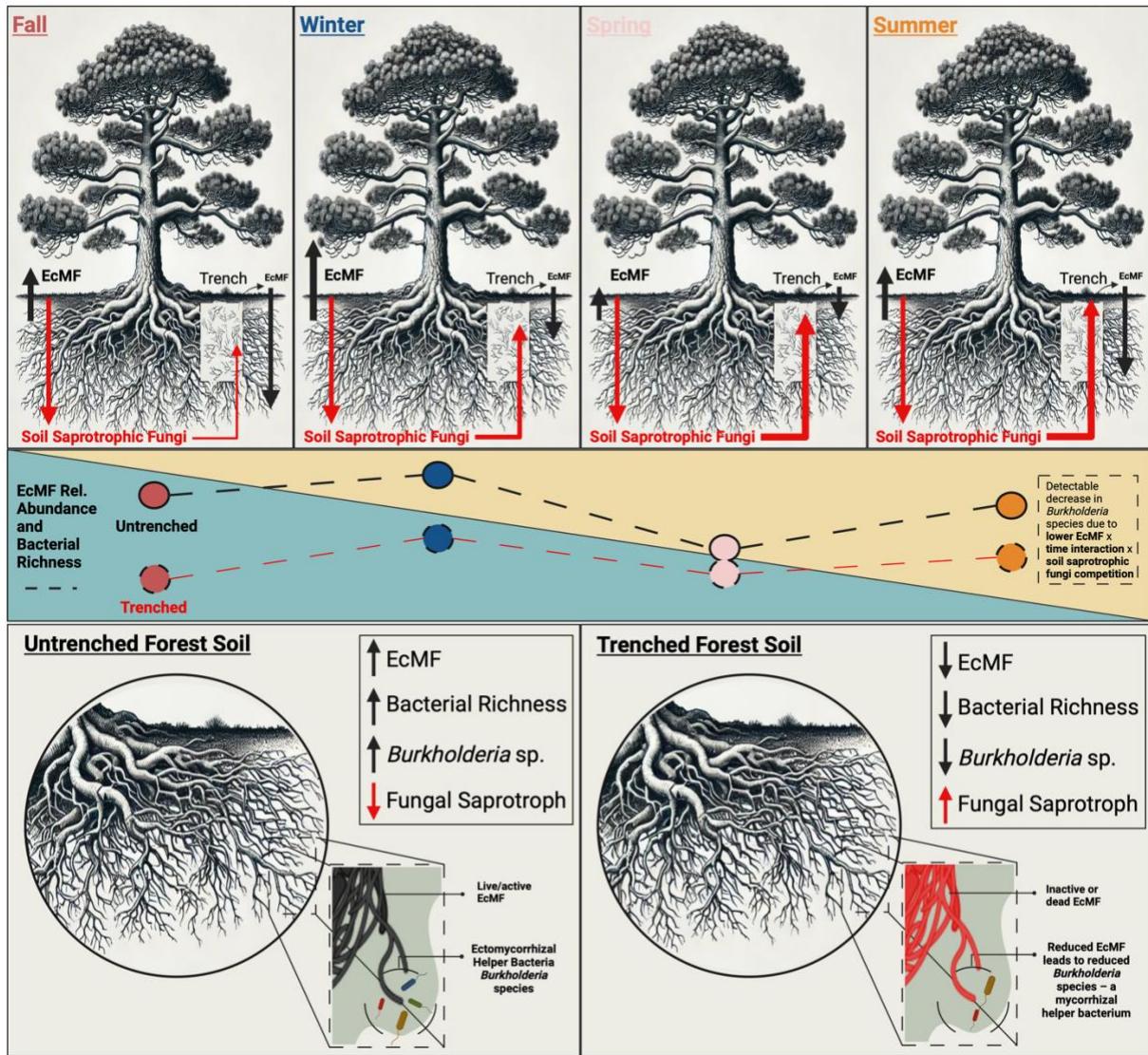
665

666

667

668

669



670
671

Fig. 5. Conceptual diagram of the spatiotemporal dynamics of bacteria-fungi interactions.

The relative increases and degrees of EcMF, soil saprotrophic fungi, and EcMF helper bacteria (i.e., *Burkholderia* species) across time and with respect to forest soil trenching are illustrated. Black arrows correspond to EcMF, and red arrow correspond to soil saprotrophic fungi. In general, we see that EcMF relative abundances are increased in untrenched plots compared to trenched plots and that soil saprotrophic fungi relative abundances are decreased in untrenched plots (relative to trenched plots). The thickness and direction of arrows correspond to the relative change in abundance (i.e., thicker arrows = stronger relative change and thinner arrows = weaker, less detectable relative change). For example, the relative abundance of EcMF in untrenched plots was observed to increase strongly in the winter, whereas a moderate rise in soil saprotrophic fungi was observed in the winter, followed by a strong increase in the spring. We also illustrate the coupled fluctuations of EcMF relative abundance, bacterial richness, and amplicon sequence variants (ASVs) matched to *Burkholderia* species throughout the seasons (shown by the line-connected dots), where dots higher on our scale correspond to greater relative abundance and vice versa. Lastly, the general pattern of microbial fluctuations is shown to illustrate the differences in soil microbiota (particularly the changes to the mycorrhizal helper

688 bacterial genus, *Burkholderia*) between untrenched and trenched forest soils. Raw data for this
689 visual summary can be found throughout the main text and supplementary data, and the
690 depictions represented here function solely to capture the primary results from our analyses.
691
692

693 References

1. Anderson MJ, Walsh DC (2013) PERMANOVA, ANOSIM, and the Mantel test in the face of heterogeneous dispersions: what null hypothesis are you testing?. *Ecological Monographs* 83(4):557-74.
2. Anthony MA, Bender SF, van der Heijden MG (2023) Enumerating soil biodiversity. *Proceedings of the National Academy of Sciences* 120(33):e2304663120.
3. Arnolds EE (1991) Decline of ectomycorrhizal fungi in Europe. *Agriculture, Ecosystems & Environment* 35(2-3):209-44.
4. Artursson V, Finlay RD, Jansson JK (2006) Interactions between arbuscular mycorrhizal fungi and bacteria and their potential for stimulating plant growth. *Environmental Microbiology* 8(1):1-0.
5. Averill C (2016) Slowed decomposition in ectomycorrhizal ecosystems is independent of plant chemistry. *Soil Biology and Biochemistry* 102:52-4.
6. Averill C, Hawkes CV (2016) Ectomycorrhizal fungi slow soil carbon cycling. *Ecology Letters* 19(8):937-47.
7. Awad A, Majcherczyk A, Schall P, Schröter K, Schöning I, Schrumpf M, Ehbrecht M, Boch S, Kahl T, Bauhus J, Seidel D (2019) Ectomycorrhizal and saprotrophic soil fungal biomass are driven by different factors and vary among broadleaf and coniferous temperate forests. *Soil Biology and Biochemistry* 131:9-18.
8. Baldrian P (2017) Forest microbiome: diversity, complexity and dynamics. *FEMS Microbiology Reviews* 41(2):109-30.
9. Beidler KV, Powers JS, Dupuy-Rada JM, Hulshof C, Medvigh D, Pizano C, Salgado-Negret B, Van Bloem SJ, Vargas G G, Waring BG, Kennedy PG (2023) Seasonality regulates the structure and biogeochemical impact of ectomycorrhizal fungal communities across environmentally divergent neotropical dry forests. *Journal of Ecology* 111(8):1598-613.
10. Bending GD (2003) Litter decomposition, ectomycorrhizal roots and the 'Gadgil' effect. *New Phytologist* 158(2).
11. Berrios L (2022a) Examining the genomic features of human and plant-associated *Burkholderia* strains. *Archives of Microbiology* 204(6):335.
12. Berrios L (2022b) The genus *Caulobacter* and its role in plant microbiomes. *World Journal of Microbiology and Biotechnology* 38(3):43.
13. Berrios L, Bogar GD, Bogar LM, Venturini AM, Willing CE, Del Rio A, Ansell TB, Zemaitis K, Velickovic M, Velickovic D, Pellitier PT, Yeam J, Hutchinson C, Bloodsworth K, Lipton MS, Peay KG (2024) Ectomycorrhizal fungi alter soil food webs and the functional potential of bacterial communities. *Msystems* 8:e00369-24.
14. Berrios L, Rentsch JD (2022) Linking reactive oxygen species (ROS) to abiotic and biotic feedbacks in plant microbiomes: The dose makes the poison. *International Journal of Molecular Sciences* 23(8):4402.
15. Berrios L, Venturini AM, Ansell TB, Tok E, Johnson W, Willing CE, Peay KG (2024) Co-inoculations of bacteria and mycorrhizal fungi often drive additive plant growth responses. *ISME Communications* 4(1):ycae104.
16. Berrios L, Yeam J, Holm L, Robinson W, Pellitier PT, Chin ML, Henkel TW, Peay KG (2023) Positive interactions between mycorrhizal fungi and bacteria are widespread and benefit plant growth. *Current Biology* 33(14):2878-87.
17. Bisbing SM, Urza AK, York RA, Hankin LE, Putz TR (2023) Persistent, viable seedbank buffers serotinous bishop pine over a broad fire return interval. *Fire Ecology* 19(1):35.
18. Bonan GB, Doney SC (2018) Climate, ecosystems, and planetary futures: The challenge to predict life in Earth system models. *Science* 359(6375):eaam8328.
19. Bowen GD, Theodorou C (1979) Interactions between bacteria and ectomycorrhizal fungi. *Soil Biology and Biochemistry* 11(2):119-26.

743 20. Brzostek ER, Dragoni D, Brown ZA, Phillips RP (2015) Mycorrhizal type determines the
744 magnitude and direction of root-induced changes in decomposition in a temperate forest. *New*
745 *Phytologist* 206(4):1274-82.

746 21. Caporaso JG, Lauber CL, Walters WA, Berg-Lyons D, Lozupone CA, Turnbaugh PJ, Fierer N,
747 Knight R (2011) Global patterns of 16S rRNA diversity at a depth of millions of sequences per
748 sample. *Proc Natl Acad Sci U S A* 108 Suppl 1:4516–4522.
749 <https://doi.org/10.1073/pnas.1000080107>

750 22. Choréño-Parra EM, Treseder KK (2024) Mycorrhizal fungi modify decomposition: a meta-
751 analysis. *New Phytologist*.

752 23. Delgado-Baquerizo M, Guerra CA, Cano-Díaz C, Egidi E, Wang JT, Eisenhauer N, Singh BK,
753 Maestre FT (2020) The proportion of soil-borne pathogens increases with warming at the global
754 scale. *Nature Climate Change* 10(6):550-4.

755 24. Dunfield PF, Yuryev A, Senin P, Smirnova AV, Stott MB, Hou S, Ly B, Saw JH, Zhou Z, Ren Y,
756 Wang J (2007) Methane oxidation by an extremely acidophilic bacterium of the phylum
757 *Verrucomicrobia*. *Nature* 450(7171):879-82.

758 25. Ekblad A, Wallander H, Godbold DL, Cruz C, Johnson D, Baldrian P, Björk RG, Epron D,
759 Kieliszewska-Rokicka B, Kjøller R, Kraigher H (2013) The production and turnover of
760 extramatrical mycelium of ectomycorrhizal fungi in forest soils: role in carbon cycling. *Plant and*
761 *Soil* 366:1-27.

762 26. Fernandez CW, Kennedy PG (2016) Revisiting the ‘Gadgil effect’: do interguild fungal interactions
763 control carbon cycling in forest soils?. *New Phytologist* 209(4):1382-94.

764 27. Fernandez CW, See CR, Kennedy PG (2020) Decelerated carbon cycling by ectomycorrhizal
765 fungi is controlled by substrate quality and community composition. *New Phytologist* 226(2):569-
766 82.

767 28. Frey-Klett P, Burlinson P, Deveau A, Barret M, Tarkka M, Sarniguet A (2011) Bacterial-fungal
768 interactions: hyphens between agricultural, clinical, environmental, and food microbiologists.
769 *Microbiology and Molecular Biology Reviews* 75(4):583-609.

770 29. Frey-Klett P, Garbaye J, Tarkka M (2007) The mycorrhiza helper bacteria revisited. *New*
771 *Phytologist* 176(1):22-36.

772 30. Fukami T (2015) Historical contingency in community assembly: integrating niches, species
773 pools, and priority effects. *Annual Review of Ecology, Evolution, and Systematics* 46(1):1-23.

774 31. Gadgil RL, Gadgil PD (1971) Mycorrhiza and litter decomposition. *Nature* 10;233(5315):133-.

775 32. Gadgil PD, Gadgil RL (1975) Suppression of litter decomposition by mycorrhizal roots of *Pinus*
776 *radiata*.

777 33. Gong S, Feng B, Jian SP, Wang GS, Ge ZW, Yang ZL (2022) Elevation matters more than
778 season in shaping the heterogeneity of soil and root associated ectomycorrhizal fungal
779 community. *Microbiology Spectrum* 10(1):e01950-21.

780 34. Gorfer M, Mayer M, Berger H, Rewald B, Tallian C, Matthews B, Sandén H, Katzensteiner K,
781 Godbold DL (2021) High fungal diversity but low seasonal dynamics and ectomycorrhizal
782 abundance in a Mountain Beech forest. *Microbial Ecology* 82(1):243-56.

783 35. Grömping U (2007) Relative importance for linear regression in R: the package relaimpo. *Journal*
784 *of Statistical Software* 17:1-27.

785 36. Gschwend F, Hartmann M, Hug AS, Enkerli J, Gubler A, Frey B, Meuli RG, Widmer F (2021)
786 Long-term stability of soil bacterial and fungal community structures revealed in their abundant
787 and rare fractions. *Molecular Ecology* 30(17):4305-20.

788 37. Hansson LJ, Ring E, Franko MA, Gärdenäs AI (2018) Soil temperature and water content
789 dynamics after disc trenching a sub-xeric Scots pine clearcut in central Sweden. *Geoderma*
790 327:85-96.

791 38. Hawkins HJ, Cargill RI, Van Nuland ME, Hagen SC, Field KJ, Sheldrake M, Soudzilovskaia NA,
792 Kiers ET (2023) Mycorrhizal mycelium as a global carbon pool. *Current Biology* 33(11):R560-73.

793 39. Hayat R, Ali S, Amara U, Khalid R, Ahmed I (2010) Soil beneficial bacteria and their role in plant
794 growth promotion: a review. *Annals of Microbiology* 60:579-98.

795 40. Kassambara A (2023) *ggnpubr: 'ggplot2' Based Publication Ready Plots*. R package version
796 0.6.0, <https://rpkgs.datanovia.com/ggnpubr/>.

797 41. Kennedy PG, Higgins LM, Rogers RH, Weber MG (2019) Colonization-competition tradeoffs as a
798 mechanism driving successional dynamics in ectomycorrhizal fungal communities. *Plos one*
799 6(9):e25126.

800 42. Koizumi T, Nara K (2020) Ectomycorrhizal fungal communities in ice-age relict forests of *Pinus*
801 *pumila* on nine mountains correspond to summer temperature. *The ISME Journal* 14(1):189-201.

802 43. Lahti L, Shetty S (2017) microbiome R package. doi: 10.18129/B9.bioc.microbiome

803 44. Lang AK, Jevon FV, Vietorizs CR, Ayres MP, Hatala Matthes J (2021) Fine roots and mycorrhizal
804 fungi accelerate leaf litter decomposition in a northern hardwood forest regardless of dominant
805 tree mycorrhizal associations. *New Phytologist* 230(1):316-26.

806 45. Li SP, Wang P, Chen Y, Wilson MC, Yang X, Ma C, Lu J, Chen XY, Wu J, Shu WS, Jiang L
807 (2020) Island biogeography of soil bacteria and fungi: similar patterns, but different mechanisms.
808 *The ISME Journal* 14(7):1886-96.

809 46. Lindahl BD, De Boer W, Finlay RD (2010) Disruption of root carbon transport into forest humus
810 stimulates fungal opportunists at the expense of mycorrhizal fungi. *The ISME Journal* 4(7):872-
811 81.

812 47. Lladó S, López-Mondéjar R, Baldrian P (2017) Forest soil bacteria: diversity, involvement in
813 ecosystem processes, and response to global change. *Microbiology and Molecular Biology
Reviews* 81(2):10-128.

815 48. Maillard F, Colin Y, Viotti C, Buée M, Brunner I, Brabcová V, Kohout P, Baldrian P, Kennedy PG
816 (2024) A cryptically diverse microbial community drives organic matter decomposition in forests.
817 *Applied Soil Ecology* 193:105148.

818 49. Malik RJ (2019) No “Gadgil effect”: Temperate tree roots and soil lithology are effective predictors
819 of wood decomposition. *Forest Pathology* 49(3):e12506.

820 50. Manfredini A, Malusà E, Costa C, Pallottino F, Mocali S, Pinzari F, Canfora L (2021) Current
821 methods, common practices, and perspectives in tracking and monitoring bioinoculants in soil.
822 *Frontiers in Microbiology* 12:698491.

823 51. McMurdie PJ, Holmes S (2013) phyloseq: an R package for reproducible interactive analysis and
824 graphics of microbiome census data. *PloS One* 8(4):e61217.

825 52. Moeller HV, Peay KG (2016) Competition-function tradeoffs in ectomycorrhizal fungi. *PeerJ*
826 4:e2270.

827 53. Mohammadi K, Khalesro S, Sohrabi Y, Heidari G (2011) A review: beneficial effects of the
828 mycorrhizal fungi for plant growth. *J. Appl. Environ. Biol. Sci* 1(9):310-9.

829 54. Oksanen J, Kindt R, Legendre P, O'Hara B, Stevens MH, Oksanen MJ, Suggs MA (2007) The
830 vegan package. *Community Ecology Package* 10(631-637):719.

831 55. Nilsson RH, Larsson KH, Taylor AF, Bengtsson-Palme J, Jeppesen TS, Schigel D, Kennedy P,
832 Picard K, Glöckner FO, Tedersoo L, Saar I (2019) The UNITE database for molecular
833 identification of fungi: handling dark taxa and parallel taxonomic classifications. *Nucleic Acids
Research* 47(D1):D259-64.

835 56. Palmer JM, Jusino MA, Banik MT, Lindner DL (2018) Non-biological synthetic spike-in controls
836 and the AMPtk software pipeline improve mycobiome data. *PeerJ* 6:e4925.

837 57. Peay KG, Bruns TD (2014) Spore dispersal of basidiomycete fungi at the landscape scale is
838 driven by stochastic and deterministic processes and generates variability in plant-fungal
839 interactions. *New Phytologist* 204(1):180-91.

840 58. Peay KG, Garbelotto M, Bruns TD (2010) Evidence of dispersal limitation in soil microorganisms:
841 isolation reduces species richness on mycorrhizal tree islands. *Ecology* 91(12):3631-40.

842 59. Pölmé S, Abarenkov K, Henrik Nilsson R, Lindahl BD, Clemmensen KE, Kauserud H, Nguyen N,
843 Kjøller R, Bates ST, Baldrian P, Frøslev TG (2020) FungalTraits: a user-friendly traits database of
844 fungi and fungus-like stramenopiles. *Fungal Diversity* 105(1):1-6.

845 60. Pongratz J, Dolman H, Don A, Erb KH, Fuchs R, Herold M, Jones C, Kuemmerle T, Luyssaert S,
846 Meyfroidt P, Naudts K (2018) Models meet data: Challenges and opportunities in implementing
847 land management in Earth system models. *Global Change Biology* 24(4):1470-87.

848 61. Quast C, Pruesse E, Yilmaz P, Gerken J, Schweer T, Yarza P, Peplies J, Glöckner FO (2012) The
849 SILVA ribosomal RNA gene database project: improved data processing and web-based tools.
850 *Nucleic Acids Research* 41(D1):D590-6.

851 62. Querejeta JI, Schlaepi K, López-García Á, Ondoño S, Prieto I, van Der Heijden MG, del Mar
852 Alguacil M (2021) Lower relative abundance of ectomycorrhizal fungi under a warmer and drier
853 climate is linked to enhanced soil organic matter decomposition. *New Phytologist* 232(3):1399-
854 413.

855 63. Rousk J, Bååth E (2011) Growth of saprotrophic fungi and bacteria in soil. *FEMS Microbiology
856 Ecology* 78(1):17-30.

857 64. Rudgers JA, Afkhami ME, Bell-Dereske L, Chung YA, Crawford KM, Kivlin SN, Mann MA, Nuñez
858 MA (2020) Climate disruption of plant-microbe interactions. *Annual Review of Ecology, Evolution,
859 and Systematics* 51(1):561-86.

860 65. See CR, Keller AB, Hobbie SE, Kennedy PG, Weber PK, Pett-Ridge J (2022) Hyphae move
861 matter and microbes to mineral microsites: Integrating the hyphosphere into conceptual models of
862 soil organic matter stabilization. *Global Change Biology* 28(8):2527-40.

863 66. Siira-Pietikäinen A, Haimi J, Fritze H (2003) Organisms, decomposition, and growth of pine
864 seedlings in boreal forest soil affected by sod cutting and trenching. *Biology and Fertility of Soils*
865 37:163-74.

866 67. Simard SW, Perry DA, Smith JE, Molina R (1997) Effects of soil trenching on occurrence of
867 ectomycorrhizas on *Pseudotsuga menziesii* seedlings grown in mature forests of *Betula
868 papyrifera* and *Pseudotsuga menziesii*. *The New Phytologist* 136(2):327-40.

869 68. Smith DP, Peay KG (2014) Sequence depth, not PCR replication, improves ecological inference
870 from next generation DNA sequencing. *PLoS one* 9(2):e90234.

871 69. Smith GR, Steidinger BS, Bruns TD, Peay KG (2018) Competition–colonization tradeoffs
872 structure fungal diversity. *The ISME Journal* 12(7):1758-67.

873 70. Smith GR, Wan J (2019) Resource-ratio theory predicts mycorrhizal control of litter
874 decomposition. *New Phytologist* 223(3):1595-606.

875 71. Soudzilovskaia NA, Douma JC, Akhmetzhanova AA, van Bodegom PM, Cornwell WK, Moens EJ,
876 Treseder KK, Tibbott M, Wang YP, Cornelissen JH (2015) Global patterns of plant root
877 colonization intensity by mycorrhizal fungi explained by climate and soil chemistry. *Global
878 Ecology and Biogeography* 24(3):371-82.

879 72. Steidinger BS, Bhatnagar JM, Vilgalys R, Taylor JW, Qin C, Zhu K, Bruns TD, Peay KG (2020)
880 Ectomycorrhizal fungal diversity predicted to substantially decline due to climate changes in North
881 American Pinaceae forests. *Journal of Biogeography* 47(3):772-82.

882 73. Sterkenburg E, Clemmensen KE, Ekblad A, Finlay RD, Lindahl BD (2018) Contrasting effects of
883 ectomycorrhizal fungi on early and late stage decomposition in a boreal forest. *The ISME Journal*
884 12(9):2187-97.

885 74. Sutton MA, Nemitz E, Erisman JW, Beier C, Bahl KB, Cellier P, De Vries W, Cotrufo F, Skiba U, Di
886 Marco C, Jones S (2007) Challenges in quantifying biosphere–atmosphere exchange of nitrogen
887 species. *Environmental Pollution* 150(1):125-39.

888 75. Swaty RL, Gehring CA, Van Ert M, Theimer TC, Keim P, Whitham TG (1998) Temporal variation
889 in temperature and rainfall differentially affects ectomycorrhizal colonization at two contrasting
890 sites. *The New Phytologist* 139(4):733-9.

891 76. Tahat MM, Kamaruzaman S, Othman R (2010) Mycorrhizal fungi as a biocontrol agent. *Plant
892 Pathology Journal* 9(4):198-207.

893 77. Taylor A, Biswas T, Randall JM, Klausmeyer K, Cohen B (2020) Parched pines: a quantitative
894 comparison of two multi-year droughts and associated mass mortalities of bishop pine (*Pinus
895 muricata*) on Santa Cruz Island, California. *Remote Sensing in Ecology and Conservation*
896 6(1):20-34.

897 78. Team RC. R: A language and environment for statistical computing.

898 79. Tecon R, Mitri S, Ciccarese D, Or D, van der Meer JR, Johnson DR (2019) Bridging the holistic-
899 reductionist divide in microbial ecology. *MSystems* 4(1):10-128.

900 80. Toju H, Kishida O, Katayama N, Takagi K (2016) Networks depicting the fine-scale co-
901 occurrences of fungi in soil horizons. *PLoS ONE* 11:e0165987.

<https://doi.org/10.1371/journal.pone.0165987>

903 81. Toju H, Tanabe AS, Yamamoto S, Sato H (2012) High-coverage ITS primers for the DNA-based
904 identification of ascomycetes and basidiomycetes in environmental samples. *PLoS one*
905 7(7):e40863.

906 82. Uroz S, Buée M, Deveau A, Mieszkin S, Martin F (2016) Ecology of the forest microbiome:
907 highlights of temperate and boreal ecosystems. *Soil Biology and Biochemistry* 103:471-88.

908 83. Wallander H, Nilsson LO, Hagerberg D, Bååth E (2001) Estimation of the biomass and seasonal
909 growth of external mycelium of ectomycorrhizal fungi in the field. *New Phytologist* 151(3):753-60.

910 84. Wang C, Kuzyakov Y (2024) Mechanisms and implications of bacterial-fungal competition for soil
911 resources. *ISME J* wrae073. doi: 10.1093/ismej/wrae073.

912 85. Weller DM, Raaijmakers JM, Gardener BB, Thomashow LS (2002) Microbial populations
913 responsible for specific soil suppressiveness to plant pathogens. *Annual Review of
914 Phytopathology* 40(1):309-48.

915 86. Wickham H, Wickham MH (2007) The ggplot package. URL: <https://cran.r-project.org/web/packages/ggplot2/index.html>.

916 87. Wieder WR, Allison SD, Davidson EA, Georgiou K, Hararuk O, He Y, Hopkins F, Luo Y, Smith MJ,
917 Sulman B, Todd-Brown K (2015) Explicitly representing soil microbial processes in Earth system
918 models. *Global Biogeochemical Cycles* 29(10):1782-800.

919 88. Willing CE, Pellitier PT, Van Nuland ME, Alvarez-Manjarrez J, Berrios L, Chin KN, Villa LM, Yeam
920 JJ, Bourque SD, Tripp W, Leshyk VO, Peay KG (2024) A risk assessment framework for the
921 future of forest microbiomes in a changing climate. *Nature Climate Change* 29:1-4.

922 89. Yamanaka T (2003) The effect of pH on the growth of saprotrophic and ectomycorrhizal ammonia
923 fungi in vitro. *Mycologia* 95(4):584-9.

924 90. Zhou Z, Wang C, Luo Y (2018) Effects of forest degradation on microbial communities and soil
925 carbon cycling: a global meta-analysis. *Global Ecology and Biogeography* 27(1):110-24.

926 91. Zinger L, Bonin A, Alsos IG, Bálint M, Bik H, Boyer F, Chariton AA, Creer S, Coissac E, Deagle
927 BE, De Barba M (2019) DNA metabarcoding—Need for robust experimental designs to draw
928 sound ecological conclusions. *Molecular Ecology* 28(8):1857-62.

929

930