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1	Changes in fungal communities across a forest disturbance gradient
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3	Lingling Shi ^{a, b, c} , Gbadamassi G.O. Dossa ^d , Ekananda Paudel ^e , Huadong Zang ^f , Jianchu Xu
4	^{a, b} , Rhett D. Harrison ^g #
5	
6	Key Laboratory for Tree diversity and Biogeography of East Asia, Kunming Institute of
7	Botany, Chinese Academy of Sciences, Kunming, Yunnan, China ^a ;
8	World Agroforestry Centre, East and Central Asia, Kunming, Yunnan, China ^b ;
9	Biogeochemistry of Agroecosystems, Georg-August-University of Goettingen, Goettingen,
10	Germany ^c ;
11	CAS Key Laboratory of Tropical Forest Ecology, Xishuangbanna Tropical Botanical Garden,
12	Chinese Academy of Sciences, Menglun, Mengla 666303, Yunnan, China ^d ,
13	Nepali Academy of Science and Technology, Khumaltar, Lalitpur, Nepal ^e ;
14	College of Agronomy and Biotechnology, China Agriculture Univiersity, Beijing, China,
15	100193 ^f ;
16	World Agroforestry Centre, East & Southern Africa Region, 13 Elm Road, Woodlands,
17	Lusaka, Zambia ^g
18	
19	Running Head: Forest disturbance affect soil fungal communities
20	
21	#Address correspondence to Rhett D. Harrison, r.harrison@cgiar.org.
22	

23 Abstract

24 Deforestation has a substantial impact on above ground biodiversity, but the response of 25 below ground soil fungi remains poorly understood. In a tropical montane rainforest in 26 southwestern China, plots were established along a forest degradation gradient ranging from 27 mature and regenerated forests to open land to examine the impacts of forest degradation and 28 deforestation on ecosystem diversity and function. Here, we evaluate the changes in below 29 ground fungal diversity and community composition using a metabarcoding approach. Soil 30 saprotrophic fungal richness declined with increasing forest disturbance. For example, 31 Penicillium spp. (Phosphorus (P) solubilizing fungi) dominated in mature forest, but were 32 less abundant in regenerating forest and showed the lowest abundance in open land sites. 33 Conversely, the abundance of facultative pathogenic fungi increased along the disturbance 34 gradient. The decline in soil saprophytic fungi may be a direct result of forest disturbance or 35 it may be associated with increased availability of soil phosphorus indirectly through an 36 increase in soil pH. The increase in abundance of facultative pathogenic fungi may be related 37 to reduce competition with saprotrophic fungi, changes in microclimate or increased spore 38 rain. These results demonstrate a loss of dominant P solubilizing saprotrophic fungi along the 39 disturbance gradient, indicated a change from soil P limitation in mature tropical forest to soil 40 C limitation in deforested sites. The increased prevalence of pathogenic fungi may inhibit 41 plant succession following deforestation. Overall, this research demonstrates that soil fungi 42 can be used as a sensitive indicator for soil health to evaluate the consequences of forest 43 disturbance.

45 Importance

46	The soil fungal functional group changes in response to forest disturbance indicated a close
47	interaction between the above-ground plant community and the below-ground soil biological
48	community. Soil saprotrophic fungi declined in relative abundance with increasing forest
49	disturbance. At the same time, the relative abundance of facultative pathogenic fungi
50	increased. The loss of saprotrophic fungal richness and abundance may have been a direct
51	result of forest disturbance or an indirect result of changes in soil pH and soil P. Furthermore,
52	the dominant P solubilizing saprotrophic fungi was replaced by diverse facultative pathogenic
53	fungi, which have weaker C decomposition ability. These changes potentially indicate a shift
54	from soil phosphate limitation to carbon limitation following deforestation. This study
55	suggests that changes in fungal functional group composition can be used as an indicator of
56	the effects of forest disturbance on soil carbon and nutrients.
57	
58	Key words
59	illumina sequencing, forest disturbance, fungal functional groups, tropical forest, P extraction,
60	soil health.

61

62 **Introduction**

Habitat disturbance and land use intensification are the principle drivers of global biodiversity loss in terrestrial ecosystems (1). Degradation of natural forest induces serious damage to the soil, such as negative changes in particle aggregation, erosion and nutrient leaching, and the loss of several organisms below ground that provide important ecosystem functions (2, 3). Recently, the response of soil microbial communities to forest disturbance have been widely discussed and attracted increased attention (4).

69 Soil fungal communities are key components involved in soil biogeochemical cycling. 70 They are affected by changes in plant community composition and in turn affect plant growth 71 (5). According to their different functions, soil fungi can be separated into saprotrophic, 72 symbiotic and pathogenic fungi. Soil organic matter and plant litter are predominantly 73 decomposed by saprotrophic fungi, and consequently these fungi provide significant soil 74 carbon resources to support plant growth in most forest ecosystems (6). Plant growth in 75 tropical forests is commonly limited by P availability (7, 8). Therefore, some fungal groups 76 can promote plant growth by increasing availability of this nutrient. For example, arbuscular 77 mycorrhizal fungi (AMF) in tropical and subtropical forests can help their host tree absorb P 78 from soil (9). Some rhizosphere fungi produce organic acids to solubilize phosphates, 79 including *Penicillium* and *Aspergillus* spp, thereby promoting plant growth in tropical forests. 80 Pathogenic fungi typical have negative plant-fungi interactions, which inhibit plant growth 81 and change the plant community composition and diversity (10). Understanding the changes 82 in the abundance of these functional groups under forest disturbance is important for 83 understanding the stability and resilience of the ecosystem.

Forest disturbance changes vegetation characteristics (e.g., plant biomass, species composition and canopy structure) and thus exerts substantial impacts on soil properties (e.g., soil C, elemental stoichiometry and pH) (11, 12). Such changes can affect soil fungal

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Applied and Environmental Microbioloay 87 attributes. A global meta-analysis found that forest degradation reduces soil C and N content, 88 increases soil pH and increases C decomposition rates. The study also found a decrease in 89 soil fungal biomass in disturbed sites, but increased species diversity (13). Consistent with 90 previous studies, they further point out that the changes in soil pH were significantly 91 correlated with changes in soil fungal community composition (14). Besides loss of host 92 species, forest disturbance can also cause the direct loss of mycorrhizal fungi and rhizosphere 93 fungi (15). However, the relative strength of these factors in determining fungal community 94 composition in tropical systems is not well understood. Accurate predictions of the impacts of 95 forest disturbance depend on identifying edaphic factors associated with significant shifts in 96 fungal communities.

97 The effects of canopy opening on understory environments can be particularly pronounced in tropical and subtropical forests, due to the increase in light, precipitation and 98 99 also temperature. Previous studies indicate that limited canopy opening can increase spatial 100 variance in soil nutrients and carbon, and C decomposition rates may be higher under canopy 101 gaps (16). In addition, canopy openings can enable colonization by new plant species and 102 fungal spores (17). The scale of a canopy opening is an important factor in driving these 103 environmental changes. Small canopy openings have been found to benefit forest 104 development in some studies (18). Several studies of soil fungi have concentrated on newly 105 created or drastically disturbed forest habitats, such as post flooding, fire or logging, and 106 focus on changes in tree species composition or soil properties (19, 20). However, it is rare 107 for studies to investigate how soil fungi respond to long-term forest disturbances associated 108 with canopy opening. Knowledge on how soil fungi respond to the canopy opening may help 109 us to understand the effects of changes in above-ground vegetation on below-ground 110 biological processes.

111 In a tropical monsoon rainforest in southwest China, we set up a series of sampling plots

112 across a forest disturbance gradient (Fig S1). In this forest, the disturbance was caused by 113 shifting cultivation and the expansion of traditional tea plantations. Our previous studies 114 investigated changes in tree and liana community composition (21), litter inputs and litter 115 decomposition (22), insect diversity and composition (23) and soil and leaf litter mesofauna 116 community composition (22) across the disturbance gradient. We found significant changes 117 in above-ground (plant and micro-climate) and below-ground environments (soil), Belecourt 118 e al 2016 together with a significant decline in rates of litter decomposition along the 119 disturbance gradient (24). However, inferences concerning soil biochemical cycling were 120 hard to predict, although we expect changes in the fungal composition particularly as a 121 consequence of drastically lower C inputs in deforested sites. The current study expands on 122 our previous work by using Illumina-sequencing to examine fungal diversity and community 123 composition along the forest disturbance gradient, and identified the factors driving fungal 124 community change. We had three aims in this study: (1) We examined whether forest 125 disturbance triggers taxonomical or functional changes in the soil fungi community. It has 126 been shown that highly degraded sites are nutrient limited. Hence, we hypothesized that soil 127 fungi will decline in their diversity and change composition along the disturbance gradient; (2) 128 We evaluated the effects of biotic (i.e., plant) and abiotic (i.e., soil properties and 129 microclimate) factors on the diversity and composition of associated fungal groups. We 130 hypothesized that biotic factors contribute to most of the variation in fungal characteristics 131 than abiotic factors due to the close linkage between above- and below-ground communities; 132 (3) We aimed to explore the potential role of soil fungi as indicator of soil heath under forest 133 disturbance. 134

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136 **Results**

137 Soil fungal diversity different with forest disturbance

138 Forest disturbance significantly reduced saprotrophic fungal richness, while facultative 139 fungal richness increased along the disturbance gradient. As a result the highest total fungal 140 richness occurred in deforested open land sites (Fig. 1). Saprotrophic fungal species richness 141 decline by 20% along the disturbance gradient from mature forest to deforested sites. With 142 the reduced abundance of saprotrophic fungi, there was a substantial increase in the 143 proportion of facultative fungi, most of which harbored pathogenetic ability (Fig. 1 and Fig. 144 S2). Total fungi species richness was most strongly associated with soil P concentration (P =145 0.017; $P_{adi} = 0.051$) (Table 2). Structural equation modeling also suggested that forest 146 disturbance has a direct negative effect on changes of soil saprotrophic fungi (Estimate (β) = -147 0.61, P < 0.01 (Fig. 2). Soil saprotrophic fungal abundance was strongly negatively 148 correlated with soil P concentration ($\beta = -0.66$, P < 0.01) and positively correlated with soil C 149 concentration ($\beta = 0.37$, P < 0.01) (Fig. 2). Soil P was indirectly affected by forest 150 disturbance thought changes in soil pH (Fig. 2). In total, environmental factors explained 41% 151 of the variance in the abundance of saprotrophic fungi (Fig. 2). These results suggested that 152 edaphic variables rather than plant community properties were the best predictors of fungal 153 richness under forest disturbance.

154 Fungal taxonomic and functional composition changed with forest disturbances

Fungal communities in deforested sites were significantly different from those in forest sites (mature and regenerating forests) in both wet and dry seasons (Fig. 3). A procrustes analysis revealed that the composition of fungal communities was significantly correlated among wet and dry season, but the overall dissimilarity among communities increased from wet to dry season (Fig. 3). PERMANOVA analysis indicated that forest disturbance ($R^2 =$ 0.07; *P* < 0.00) and seasonal change ($R^2 = 0.03$; *P* < 0.00) both had significant effects on soil

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161 fungal community composition. NMDS vectors analysis revealed that total fungal community composition was affected by soil pH ($R^2 = 0.35$; P < 0.00), Mn concentration ($R^2 = 0.34$; P < 0.04) 162 0.00), Fe concentration ($R^2 = 0.19$; P = 0.01), Ca concentration ($R^2 = 0.13$; P = 0.02), soil 163 C:N ($R^2 = 0.16$; P = 0.01) and soil P concentration ($R^2 = 0.13$; P = 0.03) (Table S2). 164 165 Forest disturbance reduced the relative abundance of total saprotrophic fungi, but

increased those fungi with pathotrophic ability within saprotrophic fungi (Fig. 4a). Strictly 166 167 saprotrophic fungi have an important role in decomposition and dominated in forest sites (Fig. 168 S2). For example, Penicillium, a saprotrophic fungi, dominated in mature forest, was less 169 abundant in regenerating forest, and least abundant in open land, regardless of the season (Fig. 170 S3). In contrast, weakly saprotrophic fungi increased in abundance in disturbed forest sites 171 (Fig. S2). For example, Cryptococcus, a common soil saprophytic yeast with a weak 172 pathotrophic ability, was more abundant in regenerating forest than in mature forest (Fig. S3). 173 Finally, the percentage of pathotrophic fungi within the saprotrophic fungi was two times 174 higher in open land sites than in forests (Fig. 4b). Fusarium is a common plant pathogenic 175 fungal group and was abundant in open land samples, but rare in regenerating and mature 176 forests (Fig. S3). Didymella, another fungal plant pathogen, was more abundant in mature 177 forest and open land than in regenerating forest in the wet season, but in the dry season it was 178 only abundant in open land (Fig. S3).

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DISCUSSION 180

181 Forest disturbance changes above- and below-ground biodiversity

182 In our study site, tree species richness was found to decline in response to forest 183 disturbance, along with associated changes in community composition and structure (25). 184 Large trees were removed in the forest in order to open the canopy for understory planting 185 with tea (26). In contrast, understory vegetation (e.g., shrub and grass) increased their

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186 diversity due to the increased availability of light (27). Therefore, at least in regenerating 187 forest, the total plant diversity actually increased and provided more diverse habitat and 188 substrates for soil fungal community, especially for the symbiotrophic fungi and saprotrophic 189 opportunities (28). Saprotrophs comprised 40% of the total fungal species richness, while 190 mycorrhizal fungi made up less than 10%. Such fungal community composition can be 191 explained by differences between bulk soil and rhizosphere soil. AMF dominate in monsoon 192 tropical forest, but are mainly concentrated in rhizosphere soil. Therefore, the effects of forest 193 disturbance on the bulk soil fungal community are mainly reflected in changes in 194 saprotrophic fungi.

195 Decline in dominant saprotrophic fungi with increasing forest disturbance

196 Dominant saprotrophic fungi declined in diversity and abundance with increasing forest 197 disturbance and these changes were apparently driven by changes in soil properties rather 198 than changes in vegetation (such as plant diversity). Soil pH was significantly higher in open 199 land sites, as compared to mature or regenerating forest sites. However, soil pH may not 200 directly affect fungal community structure (14, 29, 30). Rather, changes in plant community 201 and soil pH contributed to the differentiation of local soil P concentrations, which were 202 significantly correlated with saprotrophic and pathogenic fungi abundance. The soil microbial 203 P extractor, Penicillium, dominated fungal communities in forests (31). It is commonly 204 known that soil P is a limiting factor in many tropical forests (32). Plant root exudates, 205 mycorrhizal fungi and soil saprotrophic P extractor fungi can increase the P availability to 206 plants in tropical forests (33). Soil pH in tropical forests is typically low (about 4-5), and 207 raising soil pH (5-6) can increase the release of P and its availability to plants and other biota 208 (34). The relative abundance of soil P extractor fungi declined from mature forest to 209 regenerating forest to open land, suggesting an increase in P availability along the disturbance 210 gradient (35). As well as soil pH and soil P concentration, soil fungal composition and Applied and Environmental

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211 richness were also correlated with soil Fe and Mn concentration, which are important cations 212 affecting microbial enzyme activities in litter and soil organic matter decomposition (36). The 213 reduced species richness of saprotrophic fungi might further indicate a decrease in litter 214 decomposition capacity in disturbed sites, supporting our previous report of declining 215 decomposition rates along the disturbance gradient in this forest (22). Together with higher 216 soil P availability but lower decomposition rates, soil C availability declined along the 217 disturbance gradient and may be a new limiting factor for soil microbial decomposition (37).

218 Increased canopy opening caused by forest disturbance can lead to higher variation in 219 microclimate patterns for precipitation, temperature and light (38, 39). Fluctuations in 220 microclimate are significant in monsoon tropical areas. However, the canopy can buffer the 221 influence of changes, providing a more stable environment with lower light and precipitation 222 availability (40). In a previous study at our study site, we found that the soil temperature and 223 moisture were significantly higher in open land sites than in primary (mature) and 224 regenerating forests, especially during the dry season (21, 22) (Table 1). As expected in this 225 seasonal tropical ecosystem, the change from wet to dry periods exerted a strong influence on 226 the structure of fungal communities (41, 42). For example, the dominant fungal genus, 227 *Penicillium*, was represented in both seasons but with much higher relative abundance in the 228 dry season (Fig. S3). We speculate that high soil moisture in the wet season may limit the 229 growth of some fungi, including *Penicillium*, by increasing the abundance of anaerobic 230 microsites (41, 43). Furthermore, seasonal variation in the fungal community composition 231 was substantially more in open land as compared to forest sites, which may result from the 232 buffering effects of canopy cover (44, 45). In addition, the seasonal changes affected the 233 saprotrophic fungi more than other groups. In tropical forests, saprotrophic fungi most live in 234 the litter layer (46). The litter and surface soil layers are most prone to variation in above-235 ground microclimatic conditions, such as prolonged dryness during the dry season.

239 Deforestation contributes to increase facultatively pathogenic fungi

240 Facultative pathogenic fungi were found to make up large proportion of soil fungal 241 community in disturbed sites, especially in deforested sites. We did not find significant 242 effects of any specific single environmental factors correlated with these changes. The 243 changes of these facultative pathogenic fungi might suggest a complex interaction between 244 soil and plants. For example, increased understory vegetation increased the heterogeneity of 245 litter and root composition, which may provide diverse ecological niches for pathogenic fungi. 246 Additionally, increased light may induce saprotrophic fungi to express parthenogenesis (47). 247 Besides changes in microclimate, canopy opening can also afford opportunities for free 248 fungal spores in the air to be deposited on the soil (48). The air above the tropical forest 249 canopy is full of fungal spores, especially of plant pathogenic fungi (49). These airborne 250 fungal spores could be deposited to soil in canopy gaps (50). Further analysis on the co-251 occurrence of fungal species among habitats (Fig. S4) suggested that more unique species 252 appeared in deforested sites, especially in the wet season. Furthermore, we found a higher 253 abundance of pathogenic fungi in open land sites, and most of these species belong to wind 254 transported species. Pathogens and other symbiotic fungi that infect above-ground plant parts 255 commonly disperse as airborne spores (51). For example, *Cryptococcus* and *Didymella* have 256 been reported as saprophytic pathogens and have been transported worldwide by wind (52, 257 53). Studies have also found that high light levels trigger pathogenicity of these fungi while 258 low light favor endosymbiotic development, which constrains recruitment of endophyte-259 infested seedlings to the shaded understory through limiting survival of seedlings in direct 260 sunlight (52, 54, 55). Hence, canopy opening may not only introduce new pathogenic fungi,

but also induced their parthenogenesis.

262 Soil fungi can be used as an indicator of soil heath in forest disturbances

263 Previous studies on forest disturbance have mainly discussed changes in vegetation 264 (especially the loss of functional plant species, such as N fixing trees) or soil properties (soil 265 C and N) (56–58). However, it is often difficult to detect the changes in soil nutrient status. 266 Hence, scientists have been trying to use soil microbial functional groups to detect soil 267 nutrient limitation, because the soil microbial community is much more sensitive to soil 268 nutrient limitation than plants (59, 60). Our results indicated a close correlation between 269 changes in soil P with dominant soil fungal species, and suggest that dominant soil fungal 270 groups can be used as bio-markers to predict the condition of limiting soil nutrients (61). The 271 increase in pathogenic fungi may have a negative impact on the rate of forest succession (62). 272 Additionally, soil fungal community composition changed seasonally, and these changes 273 were more significant in deforested areas than in forests (63, 64). These results support the 274 notion that changes in the composition and diversity of soil fungi not only indicate changes in 275 the soil environment, but also contribute to the effects of forest disturbance on ecosystem 276 function.

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279 MATERIALS AND METHODS

280 Forest disturbance history. Our research was conducted in Mengsong, Xishuangbanna, 281 SW China (UTM/WGS84: 47Q 656355 E, 2377646 N, 1100-1900 m asl). The climate is 282 strongly seasonal with 80% of the rainfall occurring over six months from May to October. 283 Annual mean precipitation varies from 1600 to 1800 mm (65). Forest in the area has been 284 classified as seasonal tropical montane rain forest, which grades into seasonal evergreen 285 broad-leaved forest on hill slopes and ridges (66). The rain forest contains many floristic 286 elements in common with rain forests throughout Asia, although Dipterocarps are absent. The 287 evergreen broadleaf forest is floristically similar to more seasonal forests to the north, with 288 many species of Fagaceae and Lauraceae in the canopy.

289 The primary forests here have a density canopy covering, but the canopy structure has 290 often been changed due to long-term farming activities. The local farmers in this area 291 commonly cut down trees in the forest to increase light availability to understory tea 292 plantations. These openings may extend through time to complete deforestation. In these 293 plantations human activities, including fertilization and frequent harvesting, cause serious 294 disturbance to the environment (67). In the past farmers also practiced slash-and-burn 295 agriculture but a logging ban in the 1980s stopped this activity. So nowadays the landscape 296 has patches of forest at various stages of regrowth, as well as mature forests.

Plot design and sample collection. During 2010 to 2013, 28 sampling plots were established using a stratified random approach that resulted in 10 mature forest plots, 12 regenerating forest plots and 6 open habitat plots interspersed across the landscape (22) (Fig. S1). Samples from each sub-plot were pooled together into one sample to represent this plot. Soil samples were collected in June 2012 (wet season) and February 2013 (dry season), immediately after litter fall and during the period of the highest expected microbial activity. Fresh litter and twigs were removed from the surface and soil cores of 10 cm depth were Applied and Environ<u>mental</u>

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cycle of 10 min at 72°C. PCR products were pooled and their relative quantity was estimated purified by use of Qubit 2.0 Fluorometer (Invitrogen), and the Qubit dsDNA HS Assay Kit

PCR amplification. DNA was extracted from 0.5 g of soil per sample using the Soil DNA Isolation Kit (MoBio, Carlsbad, CA, USA) according to the manufacturer's protocols. PCR was performed using forward primers (ITS1) and degenerate reverse primer ITS2aR (68).

soil characteristics and plant properties were investigated by previous authors (22).

taken in the A layer by gently pounding metal rings into the ground. The samples were

transported to the laboratory in sterile plastic bags on ice and stored overnight at 4°C.

Approximately 20 g of moist subsample were stored at -20° C for subsequent analysis. The

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311 The PCR cocktail comprised 0.6 μ l DNA, 0.5 μ l each of the primers (20 μ M), 5 μ l 5× HOT

312 MOLPol Blend Master Mix (Molegene, Germany) and 13.4 µl double-distilled water. PCR

313 was carried out in four replicates in the following thermocycling conditions: an initial 15 min

314 at 95°C, followed by 30 cycles of 95°C for 30 s, 55°C for 30s, 72 °C for 1 min, and a final

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316 by running 2µl DNA on 1% agarose gel for 15 min. DNA samples yielding no visible band or

317 a strong band were re-amplified using 35 and 25 cycles instead. We also used negative (for

318 DNA extraction and PCR) and positive controls throughout the experiment. Amplicons were

320 (Invitrogen). Purified amplicons were subjected to normalization of quantity by use of

321 SequalPrep Normalization Plate Kit (Invitrogen, Carlsbad, CA, USA) following the

322 manufacturer's instructions. Sequencing was carried out on an Illumina MiSeq sequencer at

323 the Research and Testing Laboratory Inc., U.S.A. Although all sequencing runs in this study

324 were paired-end, only the forward reads were analyzed for the purposes of this study.

325 **Microbial community analysis.** Pyrosequencing resulted in 1174278 reads with a 326 median length of 512 base pairs (bp). Raw Illumina fastq files were de-multiplexed, quality-327 filtered, and taxonomic analyzed using QIIME (v. 1.4.0-dev) workflow using IPython

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329 picking and diversity analyses stages. In the first stage, reads were filtered using settings 330 described in manual, as modulated by the parameters (p), (q), (r), and (n) described in (22). In 331 the second stage, OTUs were assigned using the QIIME UCLUST13 wrapper, with a 332 threshold of 97% pairwise nucleotide sequence identity (97% ID), and the cluster centroid for 333 each OTU was chosen as the OTU representative sequence (70). During the taxonomic 334 analysis stage, OTU representative sequences were then classified taxonomically using non-335 default reference database from UNITE databases (71), filtered at 97% ID, using a 0.80 336 confidence threshold for taxonomic assignment. Furthermore, we assigned each fungal genus, 337 family or order to functional categories using the FUNGuild website (72). If different 338 lifestyles were present in specific genera, we chose the dominant group (> 75% of species

assigned to a specific category) or considered its ecology unknown (< 75%) levels (Table S1).

Statistical analyses. All the datasets were rarefy to 1000 per sample, using the function '*rarefy*' in R package 'vegan'(73), to reduce differences in sequencing depth. We chose to analyze richness and community composition in groups that were represented by at least 450 OTUs (fungi, Ascomycota, Basidiomycota, saprotrophic fungi, mycorrhizal fungi and pathogenic fungi). For richness analyses of soil fungi, we counted the OTU richness using the function '*diversity*' in R package 'vegan', and standardized the OTU richness using the function '*scale*' in R package 'vegan'(73).

Concentrations of soil nutrients and vegetation measurements were logarithm or square-root transformed prior to analyses to improve the distribution of residuals and reduce non-linearity. To disentangle the effects of edaphic and floristic variables on residual richness of soil fungi, individual variables were subjected to multiple regression model selection based on the corrected Akaike Information Criterion (AIC). The components of best models were forward-selected to determine their adjusted coefficients of determination as implemented in the 'vegan' package in R (73). The effects of forest disturbance and season change on fungal

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species richness data were statistically evaluated by one-way ANOVA (assumptions were tested by Levene's test for homogeneity of variances and Chi-square test for normality). When groups were significantly different, ANOVAs were followed with Tukey's HSD test. When *P* values ≤ 0.05 , examined values were considered to be significantly different. Bonferroni correction was used to adjust the *P* value in multiple comparisons.

359 We used Structural Equation Models (SEM) using Amos ver.22 (SPSS, Chicago, IL, 360 USA) to determine the direct and indirect paths between forest disturbance, environmental 361 predictors and richness of mycorrhizal fungi and saprotrophic fungi. Based on the results of 362 best variable indicators selection, we chose to include soil variables (soil pH and P 363 concentration), plant diversity (Shannon diversity index) and saprotrophic groups into model 364 construction. We tested all direct and indirect relations among exogenous and endogenous 365 variables. Then the fit of models was maximized based on both chi-square test and root mean 366 square error of approximation and Comparative Fit Index. Bootstrapping is preferred to the 367 classical maximum likelihood estimation in these cases because in bootstrapping probability 368 assessments are not based on the assumption that the data match a particular theoretical 369 distribution. There is no single universally accepted test of overall goodness of fit for SEM, applicable in all situations regardless of sample size or data distribution. Here we used the χ^2 370 test (χ^2 ; the model has a good fit when χ^2 is low (~ ≤ 2) and *P* is high (traditionally ≥ 0.05)) 371 372 and the root MSE of approximation (RMSEA; the model has a good fit when RMSEA is low 373 $(\sim \le 0.05)$ and P is high (traditionally > 0.05)). In addition, and because some variables were 374 not normal distributed, we confirmed the fit of the model using the Bollen-Stine bootstrap 375 test (the model has a good fit when the *P* value is high (traditionally > 0.10) (74).

Fungal community composition was analyzed using Global Nonmetric Multidimensional Scaling (GNMDS). The effects of forest disturbance and seasonal change were analyzed using multivariate analysis of variance (PERMANOVA) with the '*adonis*' Accepted Manuscript Posted Online

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379 function in package 'vegan'. The effects of edaphic and floristic variables on community 380 composition of soil organisms were determined based on either "Bray-Curtis" dissimilarity 381 after abundances were "Hellinger transformed", and excluding OTUs that occurred in a 382 single sample. We used the function 'envfit' to fit environmental variables while plotting the 383 non-metric multidimensional scaling (NMDS) ordination with 'metaMDS' result (75). To test 384 the correlation in community composition among soil fungi in wet and dry season, we 385 calculated the bidirectional Procrustes correlation coefficient using the 'procrustes' function 386 with 5000 permutations as implemented in the 'vegan' package. All statistical analyses were 387 carried out with the R software v3.0.2 (76).

388 Accession number(s). The raw sequencing reads were submitted to the NCBI Sequence 389 Read Archive (SRA) under the Project no. PRJNA412774, available at 390 http://www.ncbi.nlm.nih.gov/sra/, accessions no. from SRR6125802 to SRR6125608.

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1.

403	3	Huenneke LF, Jackson RB, Kinzig A, Leemans R, Lodge DM, Mooney HA,
404	ł	Oesterheld M, Poff NLR, Sykes MT, Walker BH, Walker M, Wall DH. 2000. Global
405	5	biodiversity scenarios for the year 2100. Science 287:1770-1774.
406	5	https://doi.org/10.1126/science.287.5459.1770
407	7 2.	Martínez ML, Pérez-Maqueo O, Vázquez G, Castillo-Campos G, García-Franco J,
408	}	Mehltreter K, Equihua M, Landgrave R. 2009. Effects of land use change on
409)	biodiversity and ecosystem services in tropical montane cloud forests of Mexico.
41()	For Ecol Manage 258:1856–1863. https://doi.org/10.1016/j.foreco.2009.02.023
411	l 3.	Tilman D, Cassman KG, Matson PA, Naylor R, Polasky S. 2002. Agricultural
412	2	sustainability and intensive production practices. Nature.
413	3	https://doi.org/10.1038/nature01014.
414	ł 4.	Monkai J, Hyde KD, Xu J, Mortimer PE. 2017. Diversity and ecology of soil fungal
415	5	communities in rubber plantations. Fungal Biol Rev 31:1–11.
416	5	https://doi.org/10.1016/j.fbr.2016.08.003.
417	7 5.	Baldrian P. 2017. Forest microbiome: Diversity, complexity and dynamics. FEMS
418	3	Microbiol Rev 41:109–130. https://doi.org/10.1093/femsre/fuw040.
419) 6.	Talbot JM, Bruns TD, Smith DP, Branco S, Glassman SI, Erlandson S, Vilgalys R,
420)	Peay KG. 2013. Independent roles of ectomycorrhizal and saprotrophic
421	L	communities in soil organic matter decomposition. Soil Biol Biochem 57:282–291.
422	2	https://doi.org/10.1016/j.soilbio.2012.10.004.
423	37.	LeBauer DS, Treseder KK. 2008. Nitrogen limitation of net primary productivity in
424	ł	terrestrial ecosystems is globally distributed. Ecology 89:371–379.
425	5	https://doi.org/10.1890/06-2057.1.
426	Ď	

Sala OE, Chapin FS, Armesto JJ, Berlow E, Bloomfield J, Dirzo R, Huber-Sanwald E,

Applied and Environmental Microbiology

427	8.	Feller IC, McKee KL, Whigham DF, O'Neill JP. 2003. Nitrogen vs. phosphorus
428		limitation across an ecotonal gradient in a mangrove forest. Biogeochemistry
429		62:145–175. https://doi.org/10.1023/A:1021166010892
430	9.	Smith SE. 2003. Mycorrhizal Fungi Can Dominate Phosphate Supply to Plants
431		Irrespective of Growth Responses. Plant Physiol 133:16–20.
432		https://doi.org/10.1104/pp.103.024380
433	10.	Bagchi R, Gallery RE, Gripenberg S, Gurr SJ, Narayan L, Addis CE, Freckleton RP,
434		Lewis OT. 2014. Pathogens and insect herbivores drive rainforest plant diversity
435		and composition. Nature 506:85–88. https://doi.org/10.1038/nature12911.
436	11.	Rydgren K, Økland RH, Hestmark G. 2004. Disturbance severity and community
437		resilience in a boreal forest. Ecology 85:1906–1915. https://doi.org/10.1890/03-
438		0276.
439	12.	Chazdon RL. 2003. Tropical forest recovery: Legacies of human impact and
440		natural disturbances. Perspect Plant Ecol Evol Syst. 6: 51-71.
441		https://doi.org/10.1078/1433-8319-00042.
442	13.	Zhou Z, Wang C, Luo Y. 2018. Effects of forest degradation on microbial
443		communities and soil carbon cycling: A global meta-analysis. Glob Ecol Biogeogr
444		27:110–124. https://doi.org/10.1111/geb.12663.
445	14.	Rousk J, Brookes PC, Bååth E. 2009. Contrasting soil pH effects on fungal and
446		bacterial growth suggest functional redundancy in carbon mineralization. Appl
447		Environ Microbiol 75:1589–1596. https://doi.org/10.1128/AEM.02775-08.
448	15.	van der Heijden MGA, Martin FM, Selosse MA, Sanders IR. 2015. Mycorrhizal
449		ecology and evolution: The past, the present, and the future. New Phytol
450		205:1406–1423. https://doi.org/10.1111/nph.13288.
451	16.	Ediriweera S, Singhakumara BMP, Ashton MS. 2008. Variation in canopy structure,

452		light and soil nutrition across elevation of a Sri Lankan tropical rain forest. For
453		Ecol Manage 256:1339–1349. https://doi.org/10.1016/j.foreco.2008.06.035.
454	17.	Lodge DJ, Cantrell SA, González G. 2014. Effects of canopy opening and debris
455		deposition on fungal connectivity, phosphorus movement between litter cohorts
456		and mass loss. For Ecol Manage 332:11–21.
457		https://doi.org/10.1016/j.foreco.2014.03.002
458	18.	Chen JM, Cihlar J. 1995. Plant canopy gap-size analysis theory for improving
459		optical measurements of leaf-area index. Appl Opt 34:6211.
460		https://doi.org/10.1364/A0.34.006211.
461	19.	Allison SD, Treseder KK. 2008. Warming and drying suppress microbial activity
462		and carbon cycling in boreal forest soils. Glob Chang Biol 14:2898–2909.
463		https://doi.org/10.1111/j.1365-2486.2008.01716.x.
464	20.	Sun H, Santalahti M, Pumpanen J, Köster K, Berninger F, Raffaello T, Jumpponen A,
465		Asiegbu FO, Heinonsalo J. 2015. Fungal community shifts in structure and
466		function across a boreal forest fire chronosequence. Appl Environ Microbiol
467		81:7869–7880. https://doi.org/10.1128/AEM.02063-15.
468	21.	Paudel E, Dossa GGO, Xu J, Harrison RD. 2015. Litterfall and nutrient return along
469		a disturbance gradient in a tropical montane forest. For Ecol Manage 353:97–106.
470		https://doi.org/10.1016/j.foreco.2015.05.028.
471	22.	Paudel E, Dossa GGO, De BleCourt M, Beckschafer P, Xu J, Harrison RD. 2015.
472		Quantifying the factors affecting leaf litter decomposition across a tropical forest
473		disturbance gradient. Ecosphere 6:1-20. https://doi.org/10.1890/ES15-00112.1.
474	23.	Zhang K, Lin S, Ji Y, Yang C, Wang X, Yang C, Wang H, Jiang H, Harrison RD, Yu DW.
475		2016. Plant diversity accurately predicts insect diversity in two tropical
476		landscapes. Mol Ecol. https://doi.org/10.1111/mec.13770

477	24.	Bödeker ITM, Nygren CMR, Taylor AFS, Olson Å, Lindahl BD. 2009. ClassII
478		peroxidase-encoding genes are present in a phylogenetically wide range of
479		ectomycorrhizal fungi. ISME J 3:1387–1395.
480		https://doi.org/10.1038/ismej.2009.77.
481	25.	González G, Lodge DJ. 2017. Soil biology research across latitude, elevation and
482		disturbance gradients: A review of forest studies from Puerto Rico during the past
483		25 years. Forests 8:178-183. https://doi.org/10.3390/f8060178.
484	26.	Bongers F, Poorter L, Hawthorne WD, Sheil D. 2009. The intermediate
485		disturbance hypothesis applies to tropical forests, but disturbance contributes
486		little to tree diversity. Ecol Lett 12:798–805. https://doi.org/10.1111/j.1461-
487		0248.2009.01329.x.
488	27.	Dantas V de L, Hirota M, Oliveira RS, Pausas JG. 2016. Disturbance maintains
489		alternative biome states. Ecol Lett 19:12–19. https://doi.org/10.1111/ele.12537.
490	28.	Shi LL, Mortimer PE, Ferry Slik JW, Zou XM, Xu J, Feng WT, Qiao L. 2014. Variation
491		in forest soil fungal diversity along a latitudinal gradient. Fungal Divers 64:305–
492		315. https://doi.org/10.1007/s13225-013-0270-5.
493	29.	Altizer S, Ostfeld RS, Johnson PTJ, Kutz S, Harvell CD, Lindahl BD, Menkis A, James
494		TY, Thu PQ, Suija A, Smith ME, Sharp C, Saluveer E, Saitta A, Rosas M, Riit T,
495		Ratkowsky D, Pritsch K, Põldmaa K, Piepenbring M, Phosri C, Peterson M, Parts K,
496		Pärtel K, Otsing E, Nouhra E, Njouonkou AL, Nilsson RH, Morgado LN, Mayor J,
497		May TW, Majuakim L, Lodge DJ, Lee SS, Larsson K-H, Kohout P, Hosaka K, Hiiesalu
498		I, Henkel TW, Harend H, Guo L, Greslebin A, Grelet G, Geml J, Gates G, Dunstan W,
499		Dunk C, Drenkhan R, Dearnaley J, Kesel A De, Dang T, Chen X, Buegger F, Brearley
500		FQ, Bonito G, Anslan S, Abell S, Abarenkov K. 2013. Climate Change and Infectious
501		Diseases: From Evidence to a Predictive Framework. Science 341:514–519.

Applied and Environmental Microbiology

502		https://doi.org/10.1126/science.aai8212.
503	30.	Rousk J, Bååth E, Brookes PC, Lauber CL, Lozupone C, Caporaso JG, Knight R,
504		Fierer N. 2010. Soil bacterial and fungal communities across a pH gradient in an
505		arable soil. ISME J 4:1340–1351. https://doi.org/10.1038/ismej.2010.58.
506	31.	Tedersoo L, Bahram M, Jairus T, Bechem E, Chinoya S, Mpumba R, Leal M,
507		Randrianjohany E, Razafimandimbison S, Sadam A, Naadel T, Kãljalg U. 2011.
508		Spatial structure and the effects of host and soil environments on communities of
509		ectomycorrhizal fungi in wooded savannas and rain forests of Continental Africa
510		and Madagascar. Mol Ecol 20:3071–3080. https://doi.org/10.1111/j.1365-
511		294X.2011.05145.x.
512	32.	Alvarez-Clare S, Mack MC, Brooks M. 2013. A direct test of nitrogen and
513		phosphorus limitation to net primary productivity in a lowland tropical wet forest.
514		Ecology 94:1540–1551. https://doi.org/10.1890/12-2128.1.
515	33.	Hinsinger P. 2001. Bioavailability of soil inorganic P in the rhizosphere as affected
516		by root-induced chemical changes: A review, p. 173–195. In Plant and Soil.
517		https://doi.org/10.1023/A:1013351617532.
518	34.	Kassen R. 2002. The experimental evolution of specialists, generalists, and the
519		maintenance of diversity. J Evol Biol 15:173–190.
520		https://doi.org/10.1046/j.1420-9101.2002.00377.x.
521	35.	Cotrufo MF, Wallenstein MD, Boot CM, Denef K, Paul E. 2013. The Microbial
522		Efficiency-Matrix Stabilization (MEMS) framework integrates plant litter
523		decomposition with soil organic matter stabilization: Do labile plant inputs form
524		stable soil organic matter? Glob Chang Biol 19:988–995.
525		https://doi.org/10.1111/gcb.12113.
526		

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527	36.	Camenzind T, Hättenschwiler S, Treseder KK, Lehmann A, Rillig MC. 2018.
528		Nutrient limitation of soil microbial processes in tropical forests. Ecol Monogr
529		88:4-21. https://doi.org/10.1002/ecm.1279.
530	37.	Wisz MS, Pottier J, Kissling WD, Pellissier L, Lenoir J, Damgaard CF, Dormann CF,
531		Forchhammer MC, Grytnes JA, Guisan A, Heikkinen RK, Høye TT, Kühn I, Luoto M,
532		Maiorano L, Nilsson MC, Normand S, Öckinger E, Schmidt NM, Termansen M,
533		Timmermann A, Wardle DA, Aastrup P, Svenning JC. 2013. The role of biotic
534		interactions in shaping distributions and realised assemblages of species:
535		Implications for species distribution modelling. Biol Rev 88:15–30.
536		https://doi.org/10.1111/j.1469-185X.2012.00235.x.
537	38.	Carlson DW, Groot A. 1997. Microclimate of clear-cut, forest interior, and small
538		openings in trembling aspen forest. Agric For Meteorol 87:313–329.
539		https://doi.org/10.1016/S0168-1923(95)02305-4.
540	39.	Pohlman CL, Turton SM, Goosem M. 2007. Edge effects of linear canopy openings
541		on tropical rain forest understory microclimate. Biotropica 39:62–71.
542		https://doi.org/10.1111/j.1744-7429.2006.00238.x.
543	40.	Cao M, Zhang J. 1997. Tree species diversity of tropical forest vegetation in
544		Xishuangbanna, SW China. Biodivers Conserv 6:995–1006.
545		https://doi.org/10.1023/A:1018367630923.
546	41.	Denef K, Six J, Bossuyt H, Frey SD, Elliott ET, Merckx R, Paustian K. 2001. Influence
547		of dry-wet cycles on the interrelationship between aggregate, particulate organic
548		matter, and microbial community dynamics. Soil Biol Biochem 33:1599–1611.
549		https://doi.org/10.1016/S0038-0717(01)00076-1.
550	42.	Cavagnaro TR. 2016. Soil moisture legacy effects: Impacts on soil nutrients, plants
551		and mycorrhizal responsiveness. Soil Biol Biochem 95:173–179.

552		https://doi.org/10.1016/j.soilbio.2015.12.016
553	43.	Schadt CW, Classen AT. 2007. Soil Microbiology, Ecology, and BiochemistrySoil
554		Science Society of America Journal. 4: 1420-1432.
555		https://doi.org/10.2136/sssaj2007.0017br.
556	44.	Smith AP, Marín-Spiotta E, Balser T. 2015. Successional and seasonal variations in
557		soil and litter microbial community structure and function during tropical
558		postagricultural forest regeneration: A multiyear study. Glob Chang Biol 21:3532–
559		3547. https://doi.org/10.1111/gcb.12947.
560	45.	Castaño C, Lindahl BD, Alday JG, Hagenbo A, Martínez de Aragón J, Parladé J, Pera J,
561		Bonet JA. 2018. Soil microclimate changes affect soil fungal communities in a
562		Mediterranean pine forest. New Phytol 220:1211–1221.
563		https://doi.org/10.1111/nph.15205.
564	46.	Wang FC, Fang XM, Ding ZQ, Wan SZ, Chen FS. 2016. Effects of understory plant
565		root growth into the litter layer on the leaf litter decomposition of two woody
566		species in a subtropical forest. For Ecol Manage 364:39–45.
567		https://doi.org/10.1016/j.foreco.2016.01.003.
568	47.	Remén C, Fransson P, Persson T. 2010. Population responses of oribatids and
569		enchytraeids to ectomycorrhizal and saprotrophic fungi in plant-soil microcosms.
570		Soil Biol Biochem. https://doi.org/10.1016/j.soilbio.2010.02.017
571	48.	Adams BJ, Schnitzer SA, Yanoviak SP. 2017. Trees as islands: canopy ant species
572		richness increases with the size of liana-free trees in a Neotropical forest.
573		Ecography (Cop) 40:1067–1075. https://doi.org/10.1111/ecog.02608.
574	49.	Lodge DJ, Cantrell S. 1995. Fungal communities in wet tropical forests: variation
575		in time and space. Can J Bot 73:1391–1398. https://doi.org/10.1139/b95-402.
576		

577	50.	Nakamura A, Kitching RL, Cao M, Creedy TJ, Fayle TM, Freiberg M, Hewitt CN,
578		Itioka T, Koh LP, Ma K, Malhi Y, Mitchell A, Novotny V, Ozanne CMP, Song L, Wang
579		H, Ashton LA. 2017. Forests and Their Canopies: Achievements and Horizons in
580		Canopy Science. Trends Ecol Evol 32: 438-451.
581		https://doi.org/10.1016/j.tree.2017.02.020.
582	51.	Aylor DE. 1990. The role of intermittent wind dispersal of fungal pathogens. Annu
583		Rev Phytopathol 28:73–92
584		https://doi.org/10.1146/annurev.py.28.090190.000445.
585	52.	Arm F. 1996. Keine forschung ohne ethik! Krankenpfl Soins Infirm 89:87–95.
586		https://doi.org/10.1016/j.funeco.2017.03.002.
587	53.	May RC, Stone NRH, Wiesner DL, Bicanic T, Nielsen K. 2016. Cryptococcus: From
588		environmental saprophyte to global pathogen. Nat Rev Microbiol 14:106–117.
589		https://doi.org/10.1038/nrmicro.2015.6.
590	54.	Álvarez-Loayza P, White JF, Torres MS, Balslev H, Kristiansen T, Svenning JC, Gil N.
591		2011. Light converts endosymbiotic fungus to pathogen, influencing seedling
592		survival and niche-space filling of a common tropical tree, Iriartea deltoidea. PLoS
593		One 6:e16386. https://doi.org/10.1371/journal.pone.0016386.
594	55.	Jhorar OP, Butler DR, Mathauda SS. 1998. Effects of leaf wetness duration, relative
595		humidity, light and dark on infection and sporulation by Didymella rabiei on
596		chickpea. Plant Pathol 47:586–594. https://doi.org/10.1046/j.1365-
597		3059.1998.0280a.x.
598	56.	Stork NE, Srivastava DS, Eggleton P, Hodda M, Lawson G, Leakey RRB, Watt AD.
599		2017. Consistency of effects of tropical-forest disturbance on species composition
600		and richness relative to use of indicator taxa. Conserv Biol 31:924–933.
601		https://doi.org/10.1111/cobi.12883.

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602	57.	Solar RR de C, Barlow J, Andersen AN, Schoereder JH, Berenguer E, Ferreira JN,
603		Gardner TA. 2016. Biodiversity consequences of land-use change and forest
604		disturbance in the Amazon: A multi-scale assessment using ant communities. Biol
605		Conserv 197:98–107. https://doi.org/10.1016/j.biocon.2016.03.005.
606	58.	Cohen WB, Yang Z, Stehman S V., Schroeder TA, Bell DM, Masek JG, Huang C, Meigs
607		GW. 2016. Forest disturbance across the conterminous United States from 1985-
608		2012: The emerging dominance of forest decline. For Ecol Manage 360:242–252.
609		https://doi.org/10.1016/j.foreco.2015.10.042.
610	59.	Ferris H, Tuomisto H. 2015. Unearthing the role of biological diversity in soil
611		health. Soil Biol Biochem 85:101–109.
612		https://doi.org/10.1016/j.soilbio.2015.02.037.
613	60.	Trivedi P, Delgado-Baquerizo M, Anderson IC, Singh BK. 2016. Response of Soil
614		Properties and Microbial Communities to Agriculture: Implications for Primary
615		Productivity and Soil Health Indicators. Front Plant Sci 7:990.
616		https://doi.org/10.3389/fpls.2016.00990.
617	61.	Burns JH, Anacker BL, Strauss SY, Burke DJ. 2015. Soil microbial community
618		variation correlates most strongly with plant species identity, followed by soil
619		chemistry, spatial location and plant genus. AoB Plants 7:plv030-plv030.
620		https://doi.org/10.1093/aobpla/plv030.
621	62.	Ampt EA, van Ruijven J, Raaijmakers JM, Termorshuizen AJ, Mommer L. 2018.
622		Linking ecology and plant pathology to unravel the importance of soil-borne
623		fungal pathogens in species-rich grasslands. Eur J Plant Pathol 1–16.
624		https://doi.org/10.1007/s10658-018-1573-x.
625	63.	Castillo BT, Nave LE, Le Moine JM, James TY, Nadelhoffer KJ. 2018. Impacts of
626		experimentally accelerated forest succession on belowground plant and fungal

Applied and Environmental Microbiology

627		communities. Soil Biol Biochem 125:44–53.
628	64.	Liao H, Huang F, Li D, Kang L, Chen B, Zhou T, Peng S. 2018. Soil microbes regulate
629		forest succession in a subtropical ecosystem in China: evidence from a mesocosm
630		experiment. Plant Soil 430:277–289. https://doi.org/10.1007/s11104-018-3733-
631		3.
632	65.	Xu J, Lebel L, Sturgeon J. 2009. Functional links between biodiversity, livelihoods,
633		and culture in a hani swidden landscape in southwest china. Ecol Soc 14.
634		https://doi.org/10.5751/ES-02916-140220.
635	66.	Zhu H, Shi JP, Zhao CJ. 2005. Species composition, physiognomy and plant
636		diversity of the tropical montane evergreen broad-leaved forest in southern
637		Yunnan. Biodivers Conserv 14:2855–2870. https://doi.org/10.1007/s10531-
638		004-8220-x.
639	67.	Sheil D, Burslem DFRP. 2003. Disturbing hypotheses in tropical forests. Trends
640		Ecol Evol. 18: 18-26. https://doi.org/10.1016/S0169-5347(02)00005-8.
641	68.	White TJ, Bruns T, Lee S, Taylor J. 1990. Amplification and Direct Sequencing of
642		Fungal Ribosomal Rna Genes for Phylogenetics, p. 315–322. In PCR Protocols.
643		https://doi.org/10.1016/B978-0-12-372180-8.50042-1.
644	69.	Ragan-Kelley B, Walters WA, McDonald D, Riley J, Granger BE, Gonzalez A, Knight
645		R, Perez F, Caporaso JG. 2013. Collaborative cloud-enabled tools allow rapid,
646		reproducible biological insights. ISME J. 7: 461-464.
647		https://doi.org/10.1038/ismej.2012.123.
648	70.	Rideout JR, He Y, Navas-Molina JA, Walters WA, Ursell LK, Gibbons SM, Chase J,
649		McDonald D, Gonzalez A, Robbins-Pianka A, Clemente JC, Gilbert JA, Huse SM,
650		Zhou H-W, Knight R, Caporaso JG. 2014. Subsampled open-reference clustering
651		creates consistent, comprehensive OTU definitions and scales to billions of

652		sequences. PeerJ 2:e545. https://doi.org/10.7717/peerj.545.
653	71.	Abarenkov K, Nilsson RH, Larsson KH, Alexander IJ, Eberhardt U, Erland S,
654		Høiland K, Kjøller R, Larsson E, Pennanen T, Sen R, Taylor AFS, Tedersoo L, Ursing
655		BM, Vrålstad T, Liimatainen K, Peintner U, Kõljalg U. 2010. The UNITE database
656		for molecular identification of fungi - recent updates and future perspectives.
657		New Phytol. 186: 281-285. https://doi.org/10.1111/j.1469-8137.2009.03160.x.
658	72.	Abarenkov K, Nilsson RH, Larsson KH, Alexander IJ, Eberhardt U, Erland S,
659		Høiland K, Kjøller R, Larsson E, Pennanen T, Sen R, Taylor AFS, Tedersoo L, Ursing
660		BM, Vrålstad T, Liimatainen K, Peintner U, Kõljalg U. 2010. The UNITE database
661		for molecular identification of fungi - recent updates and future perspectives.
662		New Phytol.186: 281-285. https://doi.org/10.1111/j.1469-8137.2009.03160.x.
663	73.	Oksanen, J 2017. Vegan: ecological diversity. R Package Version 2.4-4.
664		https://doi.org/ 10.1029/2006JF000545.
665	74.	Maronna RA, Martin RD, Yohai VJ. 2006. Robust Statistics: Theory and Methods.
666		Robust Stat Theory Methods 1–403. https://doi.org/10.1002/0470010940.
667	75.	O'Connor RJ. 1988. Multivariate analysis of ecological communities. Trends Ecol
668		Evol 3:121. https://doi.org/10.1016/0169-5347(88)90124-3.
669	76.	Pinheiro J, Bates D, DebRoy S, Sarkar D. 2014. R Development Core Team. 2014.
670		nlme: linear and nonlinear mixed effects models. R package version 3.1-117. R
671		Found Stat Comput Vienna, Austria.
672		
	 653 654 655 656 659 660 661 662 663 664 665 666 667 668 669 670 671 	6533 71. 654 - 655 - 656 72. 657 72. 659 72. 660 - 661 - 662 73. 664 74. 666 74. 666 75. 6668 76. 6670 76. 6701 -



FIG 1 The total fungal diversity (a and b) and dominant functional group (c and d) response to different land cover types after forest disturbance. Saprotrophic fungi were the dominant fungal group in these land cover types. Facultative fungi include several fungal groups that have multi-tropic modes, such as Pathotroph-Saprotroph, Pathotroph-Saprotroph-Symbiotroph and Pathotroph-Symbiotroph (for details refer to supplements Fig S2). MAT = mature forest, REG = regenerating forest, OPE = open land. Within each panel, different letters indicate a significant difference.



FIG 2 Structural equation model demonstrating the direct and indirect effects of forest disturbance, edaphic and floristic variables on species richness of saprotrophic fungi. The model explained 41% of the variance in abundance of saprotrophic fungi among samples. Red and black arrows indicate positive and negative relationships, respectively. The width of arrows is proportional to the strength of path coefficients. Numbers above arrows indicate standardized path coefficients. Dashed blue lines indicate tested hypotheses that were not significant.



FIG 3 (A) Non-metric multidimensional scaling (NMDS) plot of the fungal community composition (relative abundance data were Hellinger transformed) in three land cover types along a forest disturbance gradient. The ellipses represent the group mean standard error. Red indicates dry season composition and green wet season composition. Circles = mature forest; Triangles = regenerating forest; squares = open land (deforested). (B) Procrustes analysis of seasonal change (from wet season to dry season) of soil fungal community composition based on the NMDS plot. There was a highly significant correlation between the wet season community composition and dry season community composition across sites ($R^2 = 0.41$, P = 0.01). However, as indicated by the increased spread of the points (most arrows point away from centre), the fungal communities were more dissimilar in dry season than that in wet season.





Table 1. Mean (SD) daily maximum air temperature (Temp.), soil water content (Soil water), relative humidity (RH) and median photosynthetically active radiation (PAR) for 3 months in the middle of the wet (June–August) season in 2012 and dry (February–April) season in 2013. Data were recorded in the understory at three sites along a forest-disturbance gradient representing MAT = mature forest, REG = regenerating forest and OPE = open land. For PAR, readings 1 hr either side of the solar noon were used.

Season		Temp (°C)	Soil water (m ³ m ⁻³)	RH (%)	PAR (µE)
Wet	MAT	$20.6^{a}(1.5)$	0.12 (0.04)	98.0 (2.1)	9.8 ^a (4.5)
	REG	$21.5^{a}(1.7)$	0.29 (0.01)	98.5 (1.7)	31.1 ^a (14.1)
	OPE	23.0 ^b (3.0)	0.22 (0.05)	96.7 (2.7)	668.2 ^b (498.5)
Dry	MAT	22.3a (2.4)	0.04 (0.05)	64.6 (15.6)	20.4 ^a (12.7)
	REG	$25-6^{a}(3.0)$	0.05 (0.02)	72.2 (16.1)	$20.5^{a}(6.2)$
	OPE	29.2 ^b (3.0)	0.06 (0.04)	66.6 (14.7)	1419.0 ^b (391.2)

Table 2 Best regression models fungal richness for total fungi and waprotrophic fungi. NA:

Not Available"

	Total fungi				Saprotrophic fungi				
	Estimate	SE	Р	$P_{ m adj}$	Estimate	SE	Р	$P_{ m adj}$	
Soil total P	-4.191e- ⁰¹	1.698e ⁻⁰¹	0.017	0.051	-3.684e ⁻⁰¹	1.229e ⁻⁰¹	0.004	0.012	
Tree diversity			NA	NA	-3.408e ⁻⁰¹	1.229e ⁻⁰¹	0.008	0.023	
Soil Fe	3.049e ⁻⁰¹	1.698e ⁻⁰¹	0.078	0.235			NA	NA	