Generalists and Specialists Determine the Trend and Rate of Soil Fungal Distance Decay of Similarity in a 20-ha Subtropical Forest

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Abstract: Fungi are an important component of microbial communities that serve a variety of important roles in nutrient cycling and are essential for plant nutrient uptake in forest soils. Distance decay of similarity (DDS) is one of the few ubiquitous phenomena in community ecology. However, the contribution of specialist and generalist fungal species in shaping DDS remains poorly investigated. Through removing operational taxonomic units (OTU) with low or high frequencies, we rigorously quantified the impact of specialists or generalists on the change in slope, initial similarity, and halving distance of DDS of undefined saprotroph, plant mutualist, and plant putative pathogen communities in a 20-ha subtropical evergreen forest plot in Yunnan Province, Southwest China. We hypothesized that (1) the soil fungal co-occurrence networks are different between the three fungal guilds; (2) specialists and generalists contribute to the spatial turnover and nestedness of beta diversity, respectively; and (3) the removal of specialists or generalists will have opposite effects on the change of slope, initial similarity, and halving distance of DDS. Co-occurrence network analysis showed that the undefined saprotroph network had a much more complicated structure than mutualist and pathogen networks. Ascomycota and Basidiomycota were the two most abundant phyla in soil fungal communities. We found that partly in line with our expectations, the change in initial similarity increased and decreased when removing specialists and generalists, respectively, but there was always one exception guild of out of the three communities for the change in slope and halving distance. We identified that such change was mainly due to the change in turnover and nestedness of beta diversity. Furthermore, the results show that species turnover rather than species nestedness drove fungal beta diversity across functional guilds for both specialists and generalists.

Keywords: distance decay of similarity; specialists; generalists; FUNGuild; Yunnan; subtropical forest; beta diversity; soil fungal communities

1. Introduction

Distance decay of similarity (DDS) is widely observed in various organisms [1] and has long been recognized as “the first law of geography” [2]. A pattern of DDS in a community is the consequence of joint effects of dispersal, selection, drift, and diversification [3]. Numerous studies have been implemented to unravel the relative importance of assembly processes through investigating DDS at various scales [4–7]. Recently, community assembly studies showed that the mechanisms underlying the distribution of generalists and specialists differ from each other [8–10]. However, the effect of generalist and specialist species on DDS has been rarely investigated, especially for microbial communities.

How generalist and specialist species alter DDS relationships has mixed evidence, potentially due to the little related research. A theoretical study showed that specialists
have relatively weak impacts on DDS compared to those of generalists [11]. On the contrary, Oono et al. [12] demonstrated that the DDS pattern of specialists was more similar to the DDS of whole foliar fungal endophytes than that by generalists. Zinger et al. [13] found a significant association between the removal of specialists and the change in DDS slope. However, abundance might not reflect species spatial distribution directly, especially for sequencing data that are compositional [14]. Thus, more knowledge about the underlying biogeographic patterns governing fungal community composition is required to support future predictions of ecosystem functioning [15].

The fungal kingdom is one of the most diverse groups on Earth [16] and accounts for a substantial proportion of the genetic diversity in the biosphere on Earth [17]. It plays a variety of crucial roles in maintaining species coexistence [18], nutrient cycling [19,20], and facilitating nutrient uptake by plants in terrestrial ecosystems. Co-occurrence network analysis has been widely used to study microbial interaction patterns [21]. Recently, this method, combined with next-generation sequencing, has been employed to explore the structure and assembly of complex fungal communities in forest soils [22]. Communities of fungi associated with plants can vary in composition or function when environments change, which has implications for fungal contributions to plant resilience [23]. A previous study suggested that geographic distance is a better predictor of fungal community distribution [24]. Understanding soil fungi fine-scale spatial distribution patterns may add to understanding community assembly mechanisms in soil microbes.

The fungal kingdom consists of diversified functional guilds [25] exhibiting different growth strategies and nutrient-acquisition capabilities [26]. Peay et al. [17] reported that the latitudinal gradient modes of several typical fungal guilds were significantly different from each other. The relative contribution of vegetation, soil, space, and climate varies across the guilds and account for variation in patterns [16]. However, the contribution of assembly processes is scale dependent [27]. Schröter et al. [28] clearly showed that the relative importance of assembly processes in driving fungal trophic guilds in the root mycobionte shifted from large scale to local scale. The community of specialist and generalist fungi in orchard soil was dominated by deterministic and stochastic processes, respectively [29]. Moreover, Jiao and Lu [30] found that the generalist fungal community was dominated by the stochastic process. In contrast, deterministic process primarily governs specialist fungal communities in agricultural soils. Therefore, how specialists and generalists influence DDS across different fungal guilds needs to be clarified at local scales.

To address the effects of specialists and generalists on DDS of multiple fungal guilds, soil samples were collected from a subtropical evergreen forest in Southwest China. Specifically, the following questions were addressed: (1) Are the soil fungal co-occurrence networks different between the three fungal guilds? (2) Which fungal taxa (specialists or generalists) contribute more to the DDS slope, initial similarity, and halving distance within each of the three fungal guilds? (3) How does specialists and generalists influence the nestedness and turnover beta diversity components within each of the three fungal guilds?

2. Materials and Methods

2.1. Study Site

This study was carried out in the 20 ha (500 m × 400 m) mid-mountain moist evergreen broadleaved forest plot in the Ailaoshan National Nature Reserve (24°32′ N, 101°01′ E), central Yunnan Province, Southwest China (Figure 1). The elevation of the plot ranges from 2472 to 2628 m above sea level. The vegetation type is subtropical evergreen broadleaf forest, dominated by the subtropical oak species Lithocarpus hancei, L. xylocarpus, Castanopsis wattii, and Schima noronhae [31]. The climate is a typical monsoon climate, with a 1931 mm mean annual precipitation, 85% of which falls from May to October [32]. The mean monthly temperature is 11.3 °C, ranging from 5.7 °C in January to 15.6 °C in July [33]. The soil is classified as an alfisol [34].
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Figure 1. The location and distribution of soil samples (left) in the 20 ha Ailaoshan forest dynamics plot (right) in Southwest China. The red crosses in the left panel represent the soil samples. Grey lines are contour lines.

2.2. Collection of Soil Fungal Samples

Soil samples were collected over a regular grid of points with 50 m in between in the 20 ha plot (Figure 1). Each grid point was paired with three additional sample points at any random selected distances out of 0.5 m, 2.5 m, 5 m, 10 m, 15 m, and 24.9 m in three random compass directions from the grid to capture variation in soil fungal composition at finer scales. For each sample, topsoil was collected from a depth between 0 and 10 cm. The soil included fine roots and comprised both the organic layer and top mineral soil. A total of 396 soil samples was taken in June 2017. The samples were collected in plastic bags and immediately transported to the laboratory on ice for analysis. All samples were stored at −80 °C prior to fungal DNA extraction. Soil fungal DNA were extracted using the MoBio PowerSoil™ DNA Isolation Kits (Bio Mio Lab Inc., Solana Beach, CA, USA) following the manufacturer’s instructions.

2.3. Sequencing and Sequence Data Processing

The soil samples were sequenced for fungal ITS2 region. The sequencing libraries were constructed using a previously described process [35], replacing the 16S specific primers with primers targeting fungal ITS2 region. Sequencing was conducted on the MiSeq platform in the University of Minnesota Genomic Center.

Read1 and Read2 sequencing data were processed using the FAST pipeline (https://github.com/ZeweiSong/FAST/, accessed on 31 January 2019), following the protocol using both reads for ITS2 (https://github.com/ZeweiSong/FAST/wiki/Fungal-ITS2-Pipeline-Using-Both-Reads, accessed on 31 January 2019). Illumina sequencing primers were first removed using cutadapt [36] and then merged using PEAR v0.9.8 [37]. In order to align with the UNITE database, we trimmed a 60 bp tail from all sequences. These 60 bp tails were trimmed in UNITE based on our own test using ITSx [38]. Sequences with ambiguous bases or homopolymers longer than nine bases were removed and sequences with a maximum
error rate > 1 were discarded. The sequences passing these quality-filtering steps were then clustered using NINJA-OPS (Knights Lab, Minneapolis, MN, USA) [39] with a similarity of 0.97. The resulting compositional data were randomly sampled to 10,000 sequence reads per sample. This was achieved using an iterative sampling process procedure in FAST, in which each sample was randomly sampled for 1000 times, and the draw closest to medium OTU richness was picked.

2.4. Soil Fungal Classification

Each fungal OTU was assigned to a functional guild using FUNGuild [25] to examine whether fungal functional guilds differed among soil fungal communities. FUNGuild v1.0 is a database hosted by Github (https://github.com/UMNFuN/FUNGuild, accessed on 30 May 2019). The database currently contains a total of 9476 entries, with 66% at the genus level and 34% at the species level. Only OTUs with a confidence ranking of “highly probable” or “probable” were retained in our analysis as per Cregger and Veach [40]. We classified our entries into three broad functional groupings referred to as trophic modes: pathotroph (i.e., plant pathogens), symbiotroph (i.e., mutualist), and saprotroph [16].

2.5. Data Analysis

Co-occurrence networks were constructed to investigate the soil fungal communities. The interactions between fungal taxa were determined through a network structure to decipher the complexity of fungal communities in the 20 ha forest plot. Co-occurrence networks of fungal communities were constructed based on OTU relative abundances and inferred by calculating the Spearman correlation matrix between OTUs. The network was explored using the igraph package [41]. To describe the network topology, the number of nodes and edges was calculated. In the network diagrams, each node represents an OTU indicating an individual taxon, whereas the edges between every two nodes indicates significant correlations between those two taxa.

To estimate distance–decay of similarity (DDS), the Bray–Curtis dissimilarity index was calculated to describe the pairwise dissimilarity between samples. DDS was assessed in a logarithmic transformed space to enhance the linear fitting, according to Nekola and White [42], using the formula:

\[
\log(S_{\text{com}}) = \log(a) + \beta \times \log(D)
\]  

(1)

where \(S_{\text{com}}\) is the community similarity, \(a\) is the intercept parameter, \(D\) is the geographic distance, and \(\beta\) is the slope of DDS. Pairwise geographic distances between samples were calculated using the x and y coordinates implemented in the vegan package [43] and were plotted against the pairwise Bray–Curtis dissimilarities using the plot function in base R.

The impacts of specialists and generalists on the distance–decay of similarity were assessed by removing minimum frequency and removing maximum frequency OTUs, respectively. To compare the estimates of parameters obtained using different approaches for these data, three distance–decay models were fitted: (1) linear regression of Delta log(\(a\)) or initial similarity, (2) linear regression of Delta beta, and (3) linear regression of the Delta halving distance (m). The halving distance (how much farther apart sites would have to be to halve the similarity) was calculated by noting similar halves whenever the separation distance, \(d\), was increased by the amount:

\[
d_{0.5} = -\log(0.5)/\beta = 0.693/\beta
\]  

(2)

where \(d_{0.5}\) denotes the halving distance [44]. Equation (2) does not depend on \(d\) itself, but only on the increase in \(d\).

Partitioning of beta diversity, as proposed by Baselga (2010) [45], was used for calculating beta diversity components. Turnover, nestedness, slope of DDS, and initial similarity were calculated to estimate the impact of specialists or generalists for the three functional guilds. All beta diversity measures were calculated using the betapart package [46]. Addi-
Partitioning of beta diversity, as proposed by Baselga (2010) [45], was used for calculating beta diversity components. Turnover, nestedness, slope of DDS, and initial similarity were calculated to estimate the impact of specialists or generalists for the three functional guilds. All beta diversity measures were calculated using the betapart package [46]. Additionally, triangular plots (simplex) were generated to describe the distributions of abundant OTUs concerning the partitive components of beta diversity for Jaccard similarity in all OTUs and the three functional guilds. All statistical analyses were performed in the R software version 4.0.3 (accessed on 6 June 2022) [47].

3. Results

3.1. The Co-Occurrence Network of Soil Fungal Communities

Across all the 396 soil samples, a total of 3182 fungal OTUs (singletons removed) were obtained from high-throughput sequencing datasets. Among them, 1,552,685 reads and 468 OTUs of undefined saprotrophs, 56,732 reads and 170 OTUs of plant pathogens, and 985,544 reads and 349 OTUs of mutualists were found. Ascomycota and Basidiomycota accounted for the largest proportion of nodes in soil fungal communities. The co-occurrence network of soil fungal communities was more complex in undefined saprotrophs than in mutualist and plant pathogens (Figure 2). The composition of nodes and edges also varied within each network. The undefined saprotroph was the most represented functional guild and was dominated by taxa from both Ascomycota and Basidiomycota (Figure 2a). The plant pathogen network had the least number of nodes and significant correlations among nodes and was dominated by Ascomycota (Figure 2b). Meanwhile, the mutualist network was relatively more complex than plant pathogen network and was dominated by members from Basidiomycota (Figure 2c).

Figure 2. Co-occurrence networks of soil fungal communities in undefined saprotrophs (a), plant pathogens (b) and mutualists (c). OTUs are represented as nodes and correlations as edges. The node sizes are correlated to the OTU abundances, node color indicates the corresponding taxonomic assignment at phylum level, and edge width indicates the correlation strength.

3.2. Distance-Decay Relationships on Soil Fungal Communities

The removal of specialists and generalists had opposite effects on the change in slope, initial similarity, and halving distance of DDS (Figure 3). The initial similarity, DDS slope, and halving distance all decreased when specialists were removed (Figure 3a–c). In contrast, the removal of generalists resulted in increased initial similarity, DDS slope, and halving distance (Figure 3d–f). Similarly, the removal of specialists and generalists also had opposite
effects on the change in slope, initial similarity, and halving distance of DDS of undefined soil saprotrophs, plant mutualists, and pathogen communities (Supplementary Material Figures S1–S3).

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### Figure 3.
Effect of removal of specialists (top) and generalists (bottom) on initial similarity (a,d), slope (b,e), and halving distance (c,f) of distance–decay of similarity.

3.3. Beta Diversity Partitioning

Turnover and nestedness contributions to beta diversity were similar in both specialists and generalists across functional guilds. Turnover was relatively higher in generalists than in specialists (Figure 4a), whereas nestedness was relatively higher in specialists than in generalists (Figure 4b). Nevertheless, turnover was higher and nestedness was lower for both specialists and generalists and turnover was the greater contributor to overall beta diversity across functional guilds for both specialists and generalists (Figure 4a,b). Slope of DDS was relatively higher in specialists than in generalists, except in pathogen specialists (Figure 4c). Meanwhile, initial similarity was relatively higher in specialists than in generalists in all functional guilds (Figure 4d).

Similarly, the triangular plots show that the beta diversity was high for the overall fungal community, as well as the individual functional guilds (Figure 5). Of the three components, species replacement dominated for the overall fungal community and all functional guilds. In the overall fungal community and undefined saprotrophs, the Jaccard similarity component was higher than the richness difference component. In contrast, the richness difference component was higher than the Jaccard similarity component for plant pathogens and mutualists. In all cases, the sum of species replacement and richness difference components was higher than the Jaccard similarity component, i.e., the beta diversity among samples was higher than the similarity (Figure 5).
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Figure 4. Comparison of turnover (a), nestedness (b), β (c), and log(a) (d) between the original fungal functional guild communities and communities where specialists and generalists had been removed. The boxplots show the distributions of the four features in each of the fungal functional guild communities. A red star stands for an observed feature in a community. p.sp: specialists were removed for the putative plant pathogen community; m.sp: specialists were removed for the mutualistic community; u.sp: generalists were removed for the undefined saprotroph community; p.co: generalists were removed for the mutualistic community; m.co: generalists were removed for the mutualistic community; u.co: generalists were removed for the undefined saprotroph community.

Figure 5. Triangular plots (simplex) of the beta diversity partition for all OTU and three functional guilds. RichDiff = richness difference; Repl = species replacement. The bold numbers are median values.

4. Discussion

Despite recent attempts to disentangle the distribution of fungal community assembly, few studies have focused on the contribution of soil fungal specialists and generalists in shaping distance–decay of similarity. Here, the co-occurrence networks and the characteristics of soil fungal specialists and generalists in terms of beta diversity and the underlying mechanisms of biogeographic patterns were investigated. The results show that (1) fungal co-occurrence networks showed significantly distinct structural properties in undefined saprotrophs, pathogens, and plant mutualists; (2) specialists and generalists contributed differently to the DDS slope, initial similarity, and halving distance within each of the three fungal guilds; and (3) turnover was the greater contributor to overall beta diversity across functional guilds for both specialists and generalists—however, generalists had a greater turnover rate than specialists within each of the three fungal guilds whereas specialists had a higher nestedness rate than generalists. These findings would bring new insights into how these two groups contribute to the diversity of the community and add knowledge on predicting soil fungal biogeographic community patterns among different fungal guilds.

4.1. Soil Fungal Community Diversity and Composition

Network analysis showed that fungal communities in the subtropical forest were non-randomly distributed. Fungal co-occurrence networks have considerably different structural properties in undefined saprotrophs, pathogens, and plant mutualists. The undefined saprotroph network showed a much more complicated structure than the plant mutualist and pathogen networks. It has been proposed that a highly connected network provides more functional redundancy [48]. Therefore, this suggests that the complex fungal network in undefined saprotrophs would lead to greater community stability and thus may enhance the decomposition of organic matter [49]. Ascomycota and Basidiomycota

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4.2. Distance–Decay Relationships of Soil Fungal Communities

Removal of generalist and specialist taxa had opposite effects on DDS. Numerous studies have reported that generalist and specialist microbial taxa typically have distinctive distribution patterns and functional traits [56–58]. Consequently, they could have unique, possibly contrasting, influence on distance-decay relationships of species. Generalists, which live in a wide range of environmental conditions, and specialists, which require more specific environmental conditions, are thought to be governed by contrasting assembly processes and thus contribute to overall diversity in different ways [59,60]. The contribution of generalists or specialists is still limited when attempting to predict and explain spatial patterns of biodiversity.

Our observations revealed the significant effects of the removal of specialists and generalists on distance-decay of similarity in soil fungal communities. We found that the removal of specialists from the entire community resulted in a decrease in distance-decay relationships. In contrast, the removal of generalists from the entire community resulted in an increase in distance-decay relationships. The removal of generalists had the effect of increasing the change in slope, initial similarity, and halving distance of DDS, indicating impacts of dispersal limitation [61]. Dispersal limitation is one of the most important stochastic processes influencing microbial community assembly [30]. In contrast, the
removal of specialists had the effect of decreasing the change in slope, initial similarity, and halving distance of DDS, and could have been influenced by deterministic processes. These results are consistent with a previous study that showed that the assembly of generalists was driven by stochastic processes due to their general indifference to changes in habitat conditions, whereas the assembly of specialists are more affected by deterministic processes due to their preferences for specific environmental conditions [62]. Our results are also consistent with a recent study [60] that also showed slightly weakened distance–decay relationships when specialists were removed and increase in the slope of distance–decay relationships with the removal of generalists. They found that generalists contributed more to stochastic processes in community assembly whereas specialists mainly contributed to the deterministic processes [60]. Due to dispersal limitation and environmental filtering, the removal of generalists and specialists having opposite effects on the change in slope, initial similarity, and halving distance of DDS may result in a change in the turnover of beta diversity.

4.3. Beta Diversity Partitioning

Species turnover almost entirely explained the fungal beta diversity rather than nestedness, indicating that the fungal beta diversity may mainly arise from the species turnover component. The specialists may have a greater turnover rate because they adapt to environmental changes more quickly [63]. Generalists are predicted to minimize species turnover and beta diversity, which represents the loss or replacement of species due to dispersal limitation and environmental filtering [60]. Recent studies have revealed that beta diversity and species turnover of fungal communities significantly increased with increasing geographic distance [64]. They also found that fungal beta diversity was almost entirely driven by species turnover rather than species nestedness [64]. We expected turnover to predominate because nestedness was much lower, but nestedness was also involved in contributing to beta diversity. We found that distance–decay relationships of fungal guilds in the subtropical forest of Southwest China are different. Such differences should be reflected in either turnover or nestedness in species and/or trait composition among the three functional guilds. Our results show that the impact of specialists and generalists contributed primarily to the spatial turnover and nestedness of soil fungal beta diversity in our study.

The triangular plots (simplex) showed that soil fungal communities among the samples were more dissimilar than similar. Beta diversity was high and was more associated with the highest species replacement rate for all OTU and three functional guilds. This suggests that no cosmopolitan generalist fungal taxa were present in most of the samples [65]. A high rate of fungal OTU replacement or turnover could indicate either restricted fungal niches and strong environmental filtering [66], strong interspecific competitive exclusion [65], or historical contingencies with priority effects that allow early colonizers to prevent latecomers from colonizing [67,68]. At local scales, habitat heterogeneity may be high in the subtropical forest dynamics plot and could be causing the high turnover patterns. It would have been ideal to identify biotic and abiotic factors shaping soil fungal community beta diversity, but those are currently outside of the scope of this research.

5. Conclusions

This study provides insights on local soil fungal distance–decay of similarity in a subtropical forest. Through next generation sequencing and co-occurrence network analysis, we described soil fungal communities from different functional guilds. In addition, by removing low- and high-frequency fungal OTUs, the respective effects of specialist and generalist fungi on DDS were quantified. The results demonstrated that the phyla Ascomycota and Basidiomycota dominated the entire fungal community and the undefined saprotroph. Ascomycota dominated plant pathogen communities and Basidiomycota dominated the mutualist communities. Undefined soil saprotroph co-occurrence networks were more complex than plant mutualist and pathogen networks. The removal of specialists or generalists
had significant and contrasting effects on the change in slope, initial similarity, and halving distance of soil fungal DDS. The removal of specialists resulted in the weakening whereas the removal of generalists resulted in the strengthening of soil fungal DDS. These observed patterns were consistent for the entire soil fungal community and across the fungal guilds. Lastly, fungal beta diversity was derived almost entirely from species turnover rather than from species nestedness across functional guilds for both specialists and generalists. Although it is difficult to disentangle the underlying mechanisms of soil fungal community assembly, the results could be used as a tool to study fungal biogeographic patterns through time and to predict future changes that might be damaging to forests.

Supplementary Materials: The following supporting information can be downloaded at: https://www.mdpi.com/article/10.3390/f13081188/s1, Figure S1. Effect of removal of specialists (top) and generalists (bottom) on initial similarity (a,d), slope (b,e), and halving distance (c,f) of distance–decay of similarity in soil undefined saprotrophs. Figure S2. Effect of removal of specialists (top) and generalists (bottom) on initial similarity (a,d), slope (b,e), and halving distance (c,f) of distance–decay of similarity in plant mutualists. Figure S3. Effect of removal of specialists (top) and generalists (bottom) on initial similarity (a,d), slope (b,e), and halving distance (c,f) of distance–decay of similarity in pathogens.

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