

1 Evidence for subsoil specialization in arbuscular mycorrhizal 2 fungi

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9 1. Introduction

10 Arbuscular mycorrhizal (AM) fungi belong to the monophyletic subphylum
11 Glomeromycotina (Spatafora et al., 2016) and form a symbiotic relationship with most land
12 plants (Brundrett and Tedersoo, 2018). These fungi can increase plant productivity
13 (Lekberg and Koide, 2005), enhance nutrient uptake (Smith and Smith, 2011), promote soil
14 aggregation (Leifheit et al., 2014), boost pathogen protection (Veresoglou and Rillig,
15 2012), and are therefore considered important factors in agriculture. AM fungal
16 communities in arable land have been characterized both with spore identification
17 techniques (Antunes et al., 2012; Köhl et al., 2014) and molecular methods (e.g. Alguacil et
18 al., 2008; Van Geel et al., 2017) but with few exceptions, existing information is limited to
19 the first 30 cm of the soil profile. Subsoil (i.e. beneath the plough layer) AM fungal
20 communities, however, differ from those in topsoil in diversity, species composition and
21 community structure (Muleta et al., 2008; Oehl et al., 2005; Yang et al., 2010) and even
22 exhibit contrasting patterns of distribution at higher taxonomic levels (Sosa-Hernández et
23 al., 2018). We hypothesize that these differences are caused by Grinnellian ecological
24 specialization (Devictor et al., 2010), i. e. top- and subsoil represent two different
25 environments to which particular AM taxa have adapted.

26 A recent study by Roy et al. (2017) used high-throughput Illumina sequencing to analyze
27 AM fungal communities in a series of agricultural fields in western Germany forming a re-
28 cultivation chronosequence (hereafter referred to as “chronosequence fields”). In short,
29 following mining operations, pits were closed and restored with local soil and after a 3-year
30 period of alfalfa (*Medicago sativa*) cultivation reconverted to conventional agriculture. The
31 restoration was carried out with a mixture of former agricultural soil and loess parent
32 material from various depths. Therefore, we assume that directly after conversion AM
33 fungal communities from different depths experience a community coalescence event

34 (Rillig et al., 2015), i.e. taxa from different depths are mixed in the newly deposited top
35 layers. This event provides excellent opportunity to trace the fate of subsoil-specific AM
36 fungal taxa along the re-cultivation chronosequence, which allows testing our hypothesis of
37 ecological specialization of certain AM fungal taxa to deep soil layers. In a recent study,
38 we characterized AM fungal communities in an agricultural field both in top (10-30cm
39 deep) and subsoil (60-75 cm deep) (Sosa-Hernández et al., 2018), hereafter referred to as
40 “unmixed field”. We identified subsoil and topsoil indicator AM fungal taxa. Here, we
41 traced those taxa along the chronosequence fields. According to our hypothesis, AM fungal
42 taxa identified as subsoil indicators would decrease in abundance in the topsoil along the
43 chronosequence as a function of time since the mixing event occurred, while taxa identified
44 as topsoil indicators will maintain their abundance, and ii) early mixed community would
45 resemble subsoil communities and this similarity would decrease through time.

46

47 **2. Material and methods**

48 **2.1 Study sites and sequencing**

49 Both study sites are located in the southwest of the state of North Rhine-Westphalia,
50 Germany, and in both soil has been characterized as Haplic Luvisol (FAO, 1998). The
51 chronosequence fields (Roy et al., 2017) consist of a re-cultivation chronosequence after
52 open mining, comprising 10 fields. The newly deposited soil profile is about 2 m deep and
53 consists of a mixture of the previous soil (1 m deep) and loess substrate in a 1:5 ratio. For
54 the first three years after the mixing event fields are covered permanently with alfalfa
55 (hereafter referred to as phase 1), for the two following years barley (*Hordeum vulgare*)
56 was cultivated (hereafter referred to as phase 2) and after the fifth year conventional
57 agriculture was resumed, with a sugar beet (*Beta vulgaris vulgaris* var. *altissima*) - winter
58 wheat (*Triticum aestivum*) crop rotation (hereafter referred to as phase 3). From these sites
59 five samples per field were taken at a 0-10cm depth, adding up to a total of 50 samples. In
60 the unmixed field nine samples each were taken at depths from 10-30cm and 60-75cm as
61 described in Uksa et al. (2014), adding up to a total of 18 samples. Chicory (*Cichorium*
62 *intybus*) was grown on this field for the third year.

63 In both studies AM fungal communities were characterized with primers targeting the large
64 ribosomal subunit LSU including the variable D1-D2 region, using similar protocols (see
65 Roy et al. (2017) and Sosa-Hernández et al. (2018) for details). In short, after DNA
66 extraction PCR was carried out using AM fungal specific primer sets described in Krüger et
67 al. (2009). The product of this amplification was used as a template in a follow up PCR
68 using the general fungal primers LR3 and LR2rev (Hofstetter et al., 2002). Amplicons from
69 the two different studies were sequenced independently but with identical protocols on an

70 Illumina MiSeq platform at the Berlin Center for Genomics in Biodiversity Research
71 (BeGenDiv, Berlin, Germany).

72 **2.2 Bioinformatics processing of amplicons sequences**

73 A total of 2,377,171 raw sequences from the chronosequence experiment and 1,876,440
74 raw sequences from the unmixed experiment were processed separately as follows: Paired-
75 end sequences were merged and quality filtered (maximum error rate of 1) using
76 USEARCH v8.1.1861 (Edgar, 2010). Sequences were dereplicated and singletons were
77 removed. Further quality filtering was performed by aligning those sequences to an AM
78 fungal ribosomal DNA reference database (Krüger et al., 2012) using mothur v.1.38.1
79 (Schloss et al., 2009), this process also eliminated the primer sequence. Sequences not
80 overlapping the region were discarded.

81 Quality filtered and dereplicated sequences from the chronosequence experiment (58,686
82 sequences) and from the unmixed experiment (53,595 sequences) were pooled together and
83 clustered into operational taxonomic units (OTUs) at a 97% similarity level using UPARSE
84 (Edgar, 2013), which includes internal chimera removal. OTU centroids were identified and
85 non-dereplicated filtered sequences from both experiments including previously discarded
86 singletons, were mapped to those OTUs centroids at a 97% similarity level. Various format
87 editing steps such as sequence counting were performed with OBITools 1.2.9 (Boyer et al.,
88 2016).

89 Taxonomic assignment of the OTUs was carried out using BLAST+ (Camacho et al., 2009)
90 against Glomeromycotina reference sequences published in Krüger et al. (2012) and against
91 the EMBL nucleotide database (Kanz et al., 2005). Alignments below 70% similarity
92 and/or shorter than 300 bp were discarded. Results from both databases were checked for
93 consistency and matches contained in Krüger et al. (2012) were used to assign the OTUs.
94 We decided to favor matches in Krüger et al. (2012) over EMBL, due to the often imprecise
95 description of the match in the latter (e.g. “soil fungus”, ”uncultured Glomeromycota”).
96 When the taxonomic resolution of the match was sufficient, we followed a similar approach
97 to that used in Martínez-García et al. (2015) for SSU sequences, and assigned OTUs with \geq
98 97% similarity match to a species, $\geq 90\%$ to a genus, $\geq 80\%$ to a family and $\geq 70\%$ to the
99 subphylum. In cases with insufficient resolution in the match description, the OTUs were
100 assigned to the closest available taxonomic level. A species level match refers to how
101 confidently we assign a name to our OTU based on known sequences, and does not imply
102 that these OTUs are to be considered equivalent to those species.

103 **2.3 Statistical analysis**

104 All subsequent analyses were performed with R version 3.3.1 (R Core Team, 2016).
105 Community analyses were performed with the package “vegan” (Oksanen et al., 2016).
106 Before conduction the statistics, five samples belonging to the 45-year old samples in the

107 chronosequence experiment, were excluded from any subsequent analysis due to very low
108 numbers of AM fungal sequences reads. After this removal, the lowest amount of reads in a
109 sample was determined as 559 and all samples were normalized to this number by random
110 subsampling without replacement with the function “`rrarefy`”.

111 Using the sequences retrieved from the unmixed field samples we identified sub and topsoil
112 indicator OTUs using the function `multipatt()` in the package “`indicspecies`” (Cáceres and
113 Legendre, 2009) and traced their fate in the the chronosequence since the coalescent event.

114 Compositional changes between samples were measured with Bray-Curtis (Bray and
115 Curtis, 1957) and Jaccard (Jaccard, 1912) dissimilarities with the function “`vegdist`” and
116 visualized with a non-metric multidimensional scaling (NMDS) using the function
117 “`metaMDS`”. Additionally, we compared these Bray-Curtis and Jaccard distances between
118 unmixed topsoil or unmixed subsoil samples to the samples from the chronosequence to
119 test for changes in multivariate distances over time. Comparisons between dissimilarities in
120 different phases were performed with pairwise Mann–Whitney tests with correction for
121 multiple testing, as implemented with the function “`pairwise.wilcox.test`”.

122

123 **3. Results**

124 After taxonomic assignment and normalization, we identified a total of 136 AM fungal
125 OTUs. Details on the taxonomic assignation of each OTU can be found in **Table S1**. The
126 chronosequence fields yielded a diversity of 123 OTUs and the unmixed fields a diversity
127 of 73 OTUs. Between the two experiments 60 OTUs were shared, representing 44.12% of
128 the total diversity but 93.49% of the reads in “unmixed” fields and 76.53% of the reads in
129 “chronosequence” fields.

130 We identified three subsoil indicator OTUs (**Table 1**), and we detected two of these subsoil
131 indicator OTUs in topsoil from chronosequence fields with time since the mixing event up
132 to five years (**Fig. 1a**). However, we did not detect these OTUs in chronosequence fields
133 older than five years, neither in the rarefied nor in the non-normalized raw OTU tables.
134 Similarly we identified nine topsoil indicator OTUs (**Table 1**). Those topsoil indicators
135 could be detected in all chronosequence fields and they showed a tendency to increase in
136 relative abundance after the first two years since the mixing event (**Fig. 1b**).

137 AM fungal communities in recently restored chronosequence fields (i.e. shortly after the
138 mixing event) are more similar to unmixed subsoil communities, and with increasing time
139 since mixing, chronosequence communities show increasing dissimilarity to unmixed
140 subsoil communities (**Fig. 2**).

141 Bray-Curtis distances from chronosequence fields to the unmixed subsoil samples increase
142 with time, forming two significantly different groups (phase 1 + phase 2, and phase 3; **Fig.**

143 **3A**, for statistics see **Table S2**). Analogous results are obtained when considering Jaccard
144 distances (**Fig. S1A**, for statistics see **Table S2**). Bray-Curtis distances to the unmixed
145 topsoil communities follow a unimodal trend with intermediate values in phase1, minimum
146 values in phase 2, and maximum dissimilarity values in phase 3 (**Fig. 3B**, for statistics see
147 **Table S2**). Similarly, Jaccard distances to unmixed topsoil follow a unimodal trend with
148 minimum values in phase 2, but phases 1 and 3 are not significantly different (**Fig. S1B**, for
149 statistics see **Table S2**).

150 **4. Discussion**

151 We show that i) AM fungal taxa identified as subsoil indicators are present only in young
152 fields (1-3 year since the mixing event), while taxa identified as topsoil indicators are
153 present across the entire chronosequence and ii) early mixed communities from the
154 chronosequence resembled to some extent unmixed subsoil communities and this similarity
155 decreased through time after the mixing event. These results strongly suggest the inability
156 of subsoil-specific AM fungal OTUs to persist in topsoil after a subsoil-topsoil mixing
157 event. Additionally, the detection of topsoil indicators through the entire chronosequence
158 suggests that the observed loss was specific to subsoil phylotypes rather than a generalized
159 diversity loss due to soil treatment during initial deposition or subsequent management.

160 There are essentially two, not mutually exclusive hypotheses to explain this inability to
161 coexist in the topsoil: abiotic filtering and biotic interactions (Vályi et al., 2016). Possible
162 abiotic filters to subsoil AM fungal taxa in topsoil layers include disturbance in the form of
163 tillage (Kabir, 2005) and greater diurnal and seasonal variations in temperature and
164 moisture (Fierer et al., 2003). Alternatively, possible biotic filters are competitive exclusion
165 by topsoil AM fungal taxa, increased grazing pressure or differential partner selection by
166 the plant due to different nutrient availability. Particularly interesting is the notion that
167 plants might demand different services from AM fungal communities at different depths.
168 By allocating carbon selectively to the desired phylotypes (Werner and Kiers, 2015) plants
169 may shape the observed vertical distribution in AM fungal taxa. It is not clear what the
170 relative importance of abiotic filtering and biotic interactions in driving this species loss is,
171 that is to say, whether subsoil is for this AM taxa a fundamental or a realized niche
172 (Devictor et al., 2010). Equally unknown is whether the subsoil phylotypes established in
173 topsoil of the chronosequence fields and disappeared after a period of time or if they never
174 did establish and the sequences we detect represent dormant inoculum or relic DNA (Carini
175 et al., 2016). We believe that patterns in dissimilarity from “chronosequence” fields to the
176 unmixed topsoil and subsoil communities with time can be interpreted as indirect evidence
177 of the fate of these respective communities across the chronosequence. The slow increase in
178 dissimilarities to unmixed subsoil with time may point at an inactivity and/or slow decline
179 of these OTUs in topsoil, regardless of the host plants or the management. Nonetheless, the
180 observed pattern could as well be explained by the presence and slow decay of relic DNA,
181 as mentioned above. In contrast, the dissimilarities to unmixed topsoil are more responsive

182 to the changes in management in the different phases, suggesting that the members of these
183 communities were active and their populations were part of dynamic turnovers.

184 Overall, our results support our hypothesis of an ecological specialization of certain AM
185 fungal taxa to deep soil layers. Identifying the specific mechanisms driving the observed
186 patterns will require experimental approaches such as greenhouse reverse transplant
187 experiments or in-vitro competition trials. Nonetheless, our results provide a first snapshot
188 of the outcome of top- and subsoil community coalescence events. They show that AM
189 fungal taxa found in subsoils are not able to persist in topsoil layers for longer periods of
190 time. Some deep tillage practices, including deep ploughing or deep mixing, can have
191 positive effects on yield under particular scenarios (Schneider et al., 2017); however, our
192 results suggest that any practice inverting the soil profile has the potential for deleterious
193 effects on AM fungal diversity. Therefore, we suggest that such practices should only be
194 considered as extraordinary measures in soils with root-restricting layers that meet the
195 criteria for potential benefits of deep tillage (Schneider et al., 2017). Whenever possible,
196 subsoiling (i.e. deep ripping) should be preferred over any practice that inverts or mixes the
197 soil profile. With growing awareness of the potential role of AM fungi in sustainable
198 agriculture (Thirkell et al., 2017) acquiring fine-tuned knowledge about the response of
199 particular AM fungal phylotypes to tillage and soil mixing events is crucial if we are to
200 exploit the potential of mycorrhizal technology (Rillig et al., 2016). Caution is needed
201 while handling subsoil AM fungal communities if we are to not irrevocably alter them even
202 before unearthing their secrets and functional potential.

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208

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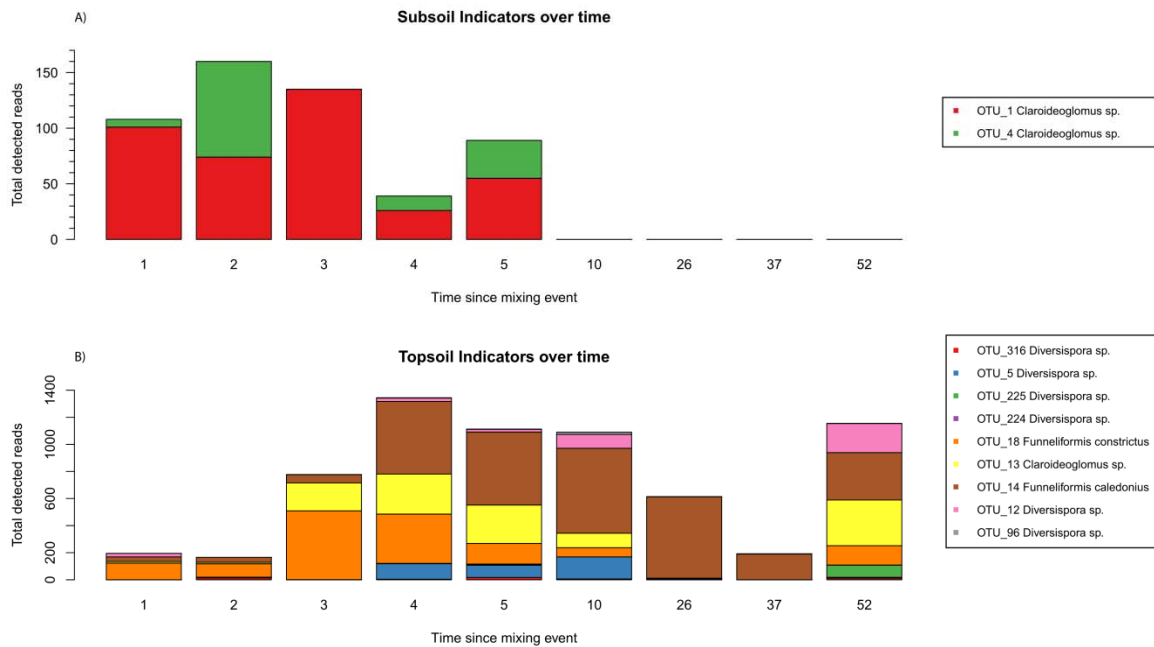
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333

334 Figures

335 Fig. 1 Sub- and Topsoil indicators over time.

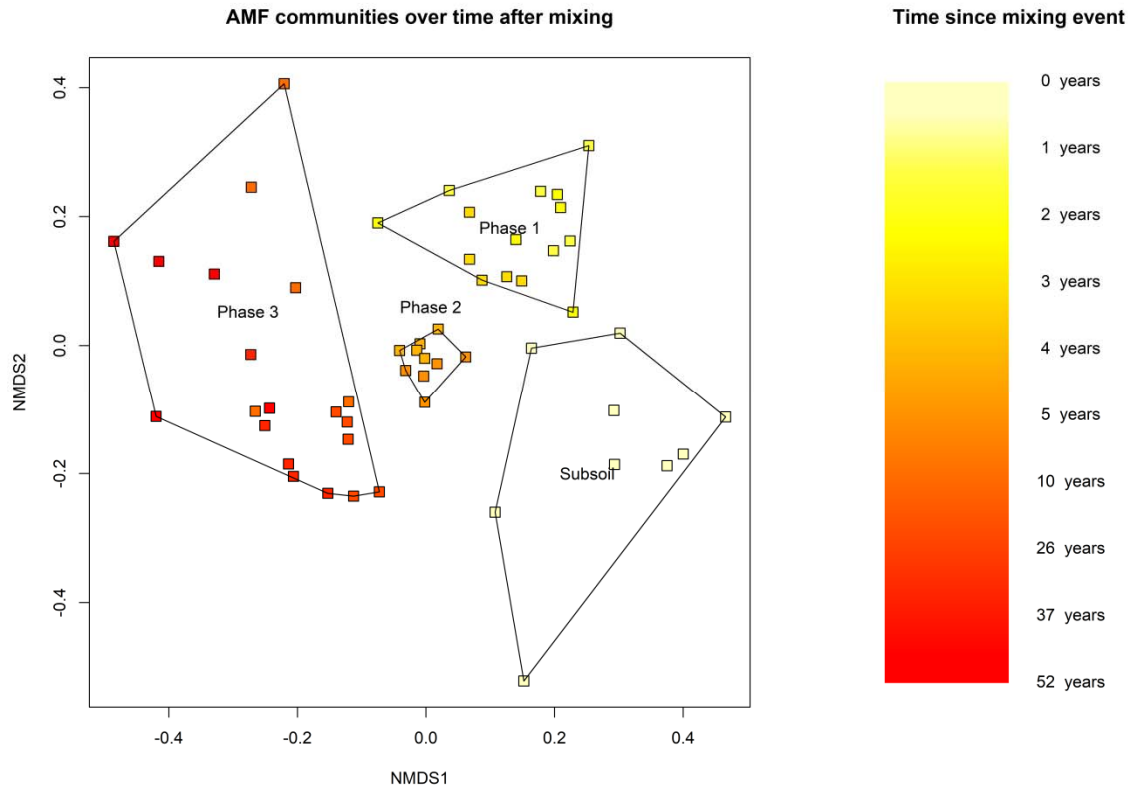
336 Number of reads detected in the chronosequence fields, for each of the subsoil A) and
337 topsoil B) indicators identified in the unmixed field. Horizontal axis represents the time
338 since the recultivation started, in years. Different indicator OTUs are coded by color.



339

340 Fig. 2 Community ordination of AMF over time.

341 Non-metric multidimensional scaling (NMDS) of a Bray-Curtis pairwise dissimilarity of
342 the AMF communities. The OTU table was rarefied to 559 reads, the minimum amount of
343 reads per sample and includes all chronosequence samples and subsoil samples from the
344 unmixed field. Time since start of the recultivation is coded by color. The polygons
345 encompass all samples from that group. Subsoil = 60–75 cm, n=9. Phase 1: 1-3 years,
346 n=15. Phase 2: 4-5 years, n=10. Phase 3: 10-52 years, n=20.



347

348 **Fig. 3 Dissimilarities to Sub- and Topsoil over time.**

349 Bray-Curtis distances (i.e. dissimilarities in both community composition and relative
350 abundances) between chronosequence fields and A) subsoil communities and B) topsoil
351 communities. Dotted lines link the means, bars represent the standard error. Different
352 phases are coded by color, significant differences between phases are represented by
353 different letters. Details on the statistics are presented in **Table S2**. Phase 1: 1-3 years,
354 n=135. Phase 2: 4-5 year, n=90. Phase 3: 10-52 years, n=180.

