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1 **Title**

2 Change in heathland dominant plants strongly increases C mineralization potential despite marginally affecting  
3 microbial community structure

4

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21

22 **Abstract**

23 Purpose

24 In many ecosystems, the identity of the dominant plant is changing because of global change. If the new dominant  
25 species has different litter and root traits than the one it replaces, it is likely to have an influence on soil microbial  
26 communities and the functions they perform. We used a grass-encroached heathland, where dwarf shrubs are  
27 replaced by grasses with different ecological traits, as a case study to explore this question. We hypothesized that  
28 grass colonization of heathland would improve litter quality, which would favor soil copiotroph microbes and  
29 increase C mineralization rate.

30 Methods

31 We established a 13-plot field observatory spanning across a 0-100% gradient of grass cover percentage. In each  
32 plot, we characterized plant, fungal and bacterial communities, using a combination of ARISA (taxonomic  
33 diversity), metabarcoding plus hierarchical modelling of species communities (community structure), FDA assay  
34 (metabolic activity) and Biolog ecoplates (functional diversity and rate of C mineralization).

35 Results

36 Our results show that microbial taxonomic and functional diversities are not affected by grass colonization.  
37 Microbial communities were also similar at high phylogenetic level, including for ericoid mycorrhizas and typical  
38 oligo- and copiotrophic species. At a finer phylogenetic level, some abundant extremophilic OTUs (*e.g.*  
39 *Acidothermus* bacteria) were progressively replaced by fungal black yeasts. Functional response of microbial  
40 communities was more obvious. The C mineralization potential significantly increased across the grass gradient.

41 Conclusion

42 Change in dominant plant traits may induce drastic functional changes in microbial communities despite having  
43 only a very minor effect on their diversity or structure.

44

45 **Keywords**

46 Heathland, grass encroachment, microbial communities, functional diversity, taxonomic diversity, C  
47 mineralization

48

49 **Introduction**

50 Global change is causing a shift in the identity of the dominant plant species. For example, southern species are  
51 moving northwards or up the altitudinal gradient and challenging local plant species because of climate change  
52 (Kelly and Goulden 2008). As another example, nitrogen deposition leads to a replacement of dominant grassland  
53 species by other species (Isbell et al. 2013). Such a change in the dominant plant in the ecosystem may lead to  
54 significant alterations on soil processes, if the new dominant plant differs in its traits from the one it replaces  
55 (Brown et al. 2001). Indeed, these new traits may affect litter quality and quantity, or rhizodeposition, both of  
56 which are recognized to influence the structural and functional properties of microbial communities  
57 (Blagodatskaya and Kuzyakov 2008; Fierer et al. 2009), which are in turn driving many soil processes, such as C  
58 and N mineralization.

59 Grass-encroached heathland in North-Western Europe are a good illustration of this phenomenon. The initially  
60 dominant ericaceous shrubs, most often belonging to the species *Calluna vulgaris*, and the invading grasses,  
61 belonging to the species *Molinia caerulea* or *Deschampsia flexuosa*, differ in terms of their spatial biomass

62 distribution, their litter quality, and the type of mycorrhizal association they are involved in. These grasses indeed  
63 have deeper and denser root systems, while heather shrubs concentrate most of their roots as a mat in the topsoil  
64 (Aerts and Heil 1993). Thus root-derived litter input in C, N and P is about one order of magnitude higher for  
65 grasses (Aerts et al. 1992). Heather litter is much more lignified than that of grasses, resulting in lower litter quality  
66 and twice higher C/N ratios under *C. vulgaris* than under *M. caerulea* (Certini et al. 2015). Finally, *M. caerulea*  
67 and *D. flexuosa* form arbuscular mycorrhizal associations, and *C. vulgaris* ericoid ones (Wang & Qiu, 2006). Both  
68 mycorrhizal types significantly differ in their functional profiles, with for example ericoid mycorrhizae having  
69 higher potential for secretion of organic-matter degrading enzymes (Genney et al. 2000; Read and Perez-Moreno  
70 2003).

71 These major changes in plant traits are expected to influence both community structure and activity of microbes,  
72 especially in the rhizosphere, where plant litter, root necromass and exudates have most impact. For example, in  
73 grasslands, switching of the dominant species to one of different productivity increased soil microbial biomass but  
74 reduced diversity (Bardgett et al. 1999). In fact, even a switch from dominance by *C. vulgaris* to another ericoid  
75 shrub, *V. myrtillus*, led to fungal assemblages in the rhizosphere, with higher frequency of Basidiomycetes under  
76 *V. myrtillus* (Bougoure et al. 2007). Even domination by a different ecotype of the same species led to significant  
77 changes in soil microbial respiration in a salt marsh ecosystem, probably because of its higher sugar percentage in  
78 rhizomes (Seliskar et al. 2002). As the microbial community is affected, so are the functions it performs: change  
79 in plant communities have been associated with alterations of soil microbial enzyme activities (Kardol et al. 2010),  
80 N mineralization, nitrification, and basal respiration (Massaccesi et al. 2015), even though these results occur due  
81 to changes in plant community and not only dominant species.

82 In the case of heathland, a change in litter input and quality and higher rhizodeposition should improve C  
83 availability, which favours microbial copiotrophs (Fierer et al. 2007). These species have a fast growth rate and  
84 therefore accelerate mineralization. Hence, grass colonization in heathlands induces significant shifts in ecological  
85 traits of the dominant plant, which has high chances to cascade down to a change in microbial community structure  
86 and functioning, and eventually in soil processes. This has however never been tested, at least to our knowledge.

87 The goal of this study is to assess how the change in a dominant plant from shrub to grass affects microbial  
88 community structure and functioning. We focused on C mineralization as a crucial microbe-driven soil function.  
89 Our hypothesis is that higher litter quality and higher rhizodeposition of the grass improves soil quality, and  
90 especially C availability, and that this affects the microbial community structure by favouring copiotrophic species,  
91 which increases C mineralisation because of their fast growth rate. To test this, we used a field observatory of 13  
92 plots differing only in terms of grass dominance. There, we assessed microbial community structure using  
93 metabarcoding, taxonomic diversity using ARISA, metabolic activity using FDA assay and mineralization of a  
94 range of C substrates using Biolog plates.

95

## 96 **Material & Methods**

97

### 98 *Field site*

99 The study was carried out in the National Park Hoge Kempen (Limburg, Belgium), at the Mechelse Heide site  
100 (50°59'07.0"N, 5°38'01.7"E), at an altitude of 104 m above sea level. The mean annual temperature is 10.3°C and  
101 the average annual precipitation is 839 mm. This area is dominated mainly by the dwarf shrub *Calluna vulgaris*,

102 with local encroachment by the subdominant species *Molinia caerulea*. The site measures 287 500 m<sup>2</sup> and consists  
103 of a mosaic of 500-1000m<sup>2</sup> heathland patches that are managed by mowing, burning or sod-cutting in order to  
104 maximize spatial heterogeneity, and is characterized by various degrees of grass encroachment. The dominant soil  
105 types within this area are albic podzols and brunic-dystric arenosols. We chose 13 of these plots, that were similar  
106 in terms of vegetation age (5-10 years old, in the early building-up phase), density, management history (burning),  
107 and slope (flat), but varied only in the *Molinia caerulea* / *Calluna vulgaris* ratio, to set up a field observatory of  
108 grass encroachment (see Figure S1 for the spatial arrangement of the plots in a map of the site), arranged in a  
109 gradient design (Figure 1).

110

#### 111 *Soil sampling*

112 The soil samples were collected on the 13th of August 2018, after a few days of rain in a summer characterised by  
113 intense droughts. In each of the 13 plots, we took soil samples to assess features of microbial communities, with  
114 two constraints in mind: i) they had to be taken at a random position in the plot and ii) we wanted to keep track of  
115 the vegetation cover at the exact spot where the soil sample was taken. Indeed, the plant cover was heterogeneous  
116 at the sub-meter scale, especially in the plots in the middle of the gradient, so there was a significant probability  
117 to sample under a heather plant even in a plot with 50% grass cover.

118 Therefore, we randomly set out three 50 cm quadrats in each plot. In each quadrat, we measured plant cover (*C.*  
119 *vulgaris*, *M. caerulea* in the herbaceous stratum and bare soil and mosses in the ground stratum; we did not observe  
120 other vascular plant species in the quadrats). In the middle of each quadrat, we took a 10 cm deep and 10 cm  
121 diameter soil core, using an auger, that was sterilized with 70% ethanol between each core. The soil sample was  
122 immediately stored at 4°C. Once in the lab, the samples (including the roots) were sieved using a 2 mm mesh sieve  
123 and aliquots of the sieved soil were labelled and stored at -20°C. The samples issued from the sieving step therefore  
124 included both bulk and rhizospheric soil.

125

#### 126 *Measurement of microbial diversity by bacterial- and fungal-ARISA*

127 Soil DNA was extracted using the Dneasy® PowerSoil® Kit (QIAGEN, Venlo, The Netherlands). The 16S-23S  
128 and 18S-28S intergenic spacer region of bacteria and fungi, respectively, were then amplified by PCR. The  
129 bacterial and fungal Automated Ribosomal Intergenic Spacer Analysis (ARISA) PCR mastermix (50 µl) contained  
130 5 µl 10x buffer, 1 µl of the primers (10 µM), 5 µl 10x dNTP's, 1µl Advantage polymerase mix, 36 µl PCR grade  
131 water (Takara Bio, Kusatsu, Japan) and 1 µl of the DNA-samples. We used the primers S-D-Bact-1522-b-S-20  
132 (5'-TGCGGCTGGATCCCCTCCTT-3') and L-D-Bact-132-a-A-18 (5'-CCGGGTTTCCCCATTCGG-3') for  
133 bacteria and 2234C (5'-GTTTCCGTAGGTGAACCTGC-3') and 3126T (5'-ATATGCTTAAGTTCAGCGGGT-  
134 3') for fungi. Both these primer sets were used and recommended for ARISA by (Ranjard et al. 2001). The PCR  
135 was executed in a PCR thermocycler Biorad T100 or Biorad C1000 (Biorad, Temse, Belgium). First, a hotstart of  
136 3 minutes was performed at 94°C, followed by 25 cycles at 94°C for 1 minute, 55° for 30 seconds and 72°C for 1  
137 minute. Finally, a terminal elongation step was performed at 72°C for 5 minutes (37).

138 The resulting PCR-amplicons were used for ARISA to generate an electropherogram. This is used to determine  
139 the microbial fingerprint of the soil sample (Ranjard et al. 2001). ARISA on each sample was done using an Agilent  
140 2100 Bioanalyzer (Agilent Technologies, Santa Clara, California, US). An Agilent DNA 1000 Kit was used with  
141 the Agilent DNA-chips. Amplicons were separated by length through capillary electrophoresis on the microchip.

142 ROX gel-dye was used as an internal standard. After adjusting the gel-dye mix to room temperature for 30 minutes,  
143 9 µl of gel-dye mix was added to the 3 designated wells on the microchip. One of the wells was firmly put under  
144 pressure with the chip priming station. Next, 5 µl marker was added to the wells designated to the ladder and  
145 samples. After running the chip in the Bioanalyzer, the results were presented in the form of gels and  
146 electropherograms in the Agilent Expert 2100 software. DNA fragments are presented as peaks in the  
147 electropherogram.

148

#### 149 *Characterization of microbial community structure*

150 The same DNA samples were used as for ARISA. DNA samples were subjected to bacterial 16S rRNA gene and  
151 fungal ITS2 region sequencing. For the bacteria, in the first round of 16S rRNA gene PCR, an amplicon of 290 bp  
152 was generated, using primers 515F and 806R (Caporaso et al. 2010). For the first round of Fungi-specific PCR,  
153 the primers gITS86f and ITS4R generate an amplicon with an average length of 450 bp (Op De Beeck et al. 2014).  
154 Using the Q5 High-Fidelity DNA Polymerase system (M0491, New England Biolabs, Ipswich, Massachusetts,  
155 US), a reaction volume of 25 µl per sample was prepared containing 1 µl of extracted DNA (final DNA-  
156 concentration per reaction 1-10 ng), 1x Q5 Reaction Buffer with 2 mM MgCl<sub>2</sub>, 200 µM dNTP mix, 1x Q5 High  
157 GC Enhancer (for the soil and fungi samples), 0.2 µM forward and reverse primer, and 1.2 U Q5 High-Fidelity  
158 DNA polymerase. The PCR program started with an initial denaturation for 3 min at 98 °C, followed by a 30 s  
159 denaturation at 98 °C, a 30 s annealing at 53 °C for V4 (58 °C for ITS) and a 1 min extension at 72 °C, all three  
160 steps were repeated for a total of 35 cycles. The reaction was ended by a final 7 min extension at 72 °C. The  
161 amplified DNA was purified using the AMPure XP beads (Beckman Coulter, Brea, California, US) and the  
162 MagMax magnetic particle processor (Thermo Fisher Scientific, Waltham, Massachusetts, US). Subsequently, 5  
163 µl of the cleaned PCR product was used for the second PCR, attaching the Nextera indices (Nextera XT Index Kit  
164 v2 Set A (FC-131-2001), and D (FC-131-2004), Illumina, San Diego, California, US). For these PCR reactions, 5  
165 µl of the purified PCR product was used in a 25 µl reaction volume, and prepared following the 16S Metagenomic  
166 Sequencing Library Preparation Guide. PCR conditions were the same as described above, but the number of  
167 cycles reduced to 20, and 55 °C annealing temperature. PCR products were cleaned with the AMPure XP kit, and  
168 then quantified using the Qubit dsDNA HS assay kit (Thermo Fisher Scientific) and the Qubit 2.0 Fluorometer  
169 (Thermo Fisher Scientific). Once the molarity of the sample was determined, the samples were diluted down to 4  
170 nM using 10 mM Tris pH 8.5 prior to sequencing on an Illumina MiSeq. Samples were sequenced using the MiSeq  
171 Reagent Kit v3 (600 cycle) (MS-102-3003) and 15% PhiX Control v3 (FC-110-3001). For quality control, a DNA-  
172 extraction blank and PCR blank were included throughout the process, and also the ZymoBIOMICS Microbial  
173 Mock Community Standard (D6300) to test efficiency of DNA extraction (Zymo Research, Irvine California, US).  
174 Obtained sequences were clustered into operational taxonomic units (OTUs) and annotated using Qiime (Caporaso  
175 et al. 2010) within the DADA2 package with standard settings, with the SILVA database for bacteria and the Unite  
176 database for Fungi.

177

#### 178 *Measurement of microbial metabolic activity*

179 The Fluorescein Diacetate (FDA) hydrolysis assay was performed following the protocol of (Schnurer and  
180 Rosswall 1982) and (Adam and Duncan 2001). Aliquot soil samples (1g) were placed in a 125 ml Erlenmeyer  
181 flask with 50 mL of 60 mM sodium phosphate buffer (pH = 7.6) and 0.5 mL of an FDA stock solution (2 mg

182 fluorescein diacetate per mL of reagent-grade acetone), and incubated for 3 h at 37°C to allow hydrolysis of  
183 fluorescein diacetate to fluorescein by a variety of different microbial enzymes (e.g. proteases, lipases, and  
184 esterases) present in the soil samples. Negative control flasks without soil samples were included to check for non-  
185 specific fluorescein release. After incubation, 2 mL of acetone was added to the suspension and swirled to mix the  
186 contents and terminate the reaction. Thereafter, 30 mL of the soil suspension was transferred to a 50 mL centrifuge  
187 tube and centrifuged at 3500 g for 10 min. The resulting supernatant was filtered through a Whatman No. 2 filter  
188 paper into a new 50 mL centrifuge tube. Finally, the filtrate (200 µL) was transferred to a 96 well plate and the  
189 absorbance was measured at 490 nm on a plate reader (FLUOstar® Omega plate reader, BMG LABTECH GmbH,  
190 Ortenberg, Germany).

191

#### 192 *Measurement of microbial C mineralization functions*

193 The C mineralization potential by microbial communities was measured using The functions of C mineralization  
194 by soil microbes were investigated using community-level physiological profiling (Biolog PM1 MicroPlate  
195 Carbon Sources; BioLog Inc., Hayward, California, US), where an inoculum prepared from a given soil sample  
196 was exposed to 32 carbon sources, and the mineralization rate measured spectrophotometrically as the reduction  
197 of a tetrazolium dye by microbial cell respiration, integrated over time using trapezoidal estimation. Using  
198 sterilized equipment, aliquot soil samples (1g) were dissolved in 10 mL sterile 10 mM PBS buffer (130 mM NaCl,  
199 7 mM Na<sub>2</sub>HPO<sub>4</sub>, 3 mM NaH<sub>2</sub>PO<sub>4</sub>, pH 7.4) and shaken for 20 minutes at room temperature. After shaking, soil  
200 particles were allowed to settle for 30 minutes at 4°C. Subsequently, 130 µL of the supernatant was dispensed into  
201 each well of the Biolog MicroPlate. Inoculated plates were placed in self-sealing plastic bags containing a water-  
202 soaked paper towel to minimize evaporation from the wells, and incubated at 28-30 °C. Absorbance was measured  
203 at 595 nm with a plate reader immediately after inoculation (0 h) and at 3, 6, 18, 24, 48, 72, and 144 h. Raw  
204 absorbance values were recorded at each time point and individually standardized by subtracting the corresponding  
205 absorbance value measured immediately after inoculation (0 h) (reaction-independent absorbance). Furthermore,  
206 to semi-quantitatively evaluate the kinetic Biolog MicroPlate dataset, the net area under the absorbance versus  
207 time curve was calculated according to the trapezoidal approximation (Guckert et al. 1996). The resulting value  
208 calculated via the trapezoidal approximation summarizes different aspects of the measured respirometric reaction,  
209 including differences in lag phases, increase rates, or maximum optical densities (Preston-Mafham et al. 2002).  
210 The Shannon-Weaver index (H) was calculated per sample as an indicator of functional diversity, or *range* of C  
211 sources used by the community of microbes. The *rate* of C source use by the microbial community was calculated  
212 as the average of the area under the curve for all C sources (referred to as AWCD: average well colour  
213 development). The C sources could be categorized into 6 different types, based on (Rutgers et al. 2016):  
214 carbohydrates (β-methyl-D-glucoside, n-acetyl-D-glucosamine, D-cellobiose, glucose-1-phosphate, α-D-lactose,  
215 D-1-α-glycerol-phosphate, D-galactonicacid-1-lactone, D-xylose, 1-erythritol, D-mannitol), amino acids (L-  
216 arginine, L-asparagine, L-serine, L-phenylalanine, L-threonine), carboxylic acids (glycyl-L-glutamic acid, D-  
217 galactonic acid, γ-hydroxybutyric acid, D-glucosamic acid, itaconic acid, alpha-ketobutyric acid, pyruvic acid  
218 methyl ester, D-malic acid), polymers (tween40, tween80, α-cyclodextrin, glycogen), phenolics (2-hydroxy-  
219 benzoic acid, 4-hydroxy-benzoic acid), and amines (phenylethylamine, putrescine).

220

#### 221 *Statistics*

222 The strength of the relationship between grass colonization and both microbial diversity and activity was assessed  
223 using correlation tests. Correlation between grass cover and both bacterial and fungal microbial diversity was  
224 tested in three ways: with OTU richness (number of peaks in the ARISA electropherogram) using a Kendall rank  
225 correlation coefficient test, after normality and homoscedasticity assumptions were tested and not met; and with  
226 Shannon and Simpson indices (diversity indices based on OTU richness and abundance, assessed by  
227 metabarcoding, see above) using a Pearson's correlation test. The correlation between grass cover and microbial  
228 activity (measured as the absorbance at 690 nm in the FDA assay, and where no distinction between bacteria and  
229 fungi can be made) was tested using a Pearson's correlation test.

230 To evaluate the effect of grass colonization on the microbial community structure, we used Hierarchical Modelling  
231 of Species Communities (HMSC) (Tikhonov et al. 2020), which is a type of joint species distribution modelling  
232 (Wilkinson et al. 2021). The advantage of these models over PERMANOVA, ANOSIM or a Mantel Test is that  
233 HMSC explicitly accounts for the effect of species-species interactions on their distribution over an environmental  
234 variable (Tikhonov et al. 2020), to reduce the effect of species interactions as a confounding variable. We built  
235 three separate mixed models, all with quadrat's grass cover as independent variable and quadrat identity as random  
236 variable; only the dependant variable (OTU abundance as counts) differed between the models. Model 1 grouped  
237 all OTUs present in at least 40% of the samples (16 of 39), model 2 in 25% (10 of 39), and model 3 in 10% (4 of  
238 39). The rationale behind testing these three models was that we wanted to find a compromise between including  
239 as many OTUs in the analysis as possible, on the one hand, and keeping the highest explanatory power, on the  
240 other hand. We then fitted all models using Bayesian statistics. For that, we sampled the posterior distribution  
241 using two Monte Carlo Markov Chains (MCMC), each of which being run for 5000 iterations, out of which the  
242 first 4900 were removed as burn-in and the remaining ones thinned by either 1, 10, or 100. We then evaluated and  
243 compared model's explanatory power using the Root Mean Square Error (RMSE), coefficient of determination  
244 ( $SR^2$ ), area under the receiver operating characteristic curve (AUC) and coefficient of discrimination ( $Tjur^2$ ) (see  
245 results in Figure S2); and verified that convergence of the Markov chains was achieved by comparing  $ess.\beta$ ,  $ess.V$ ,  
246  $PSRF.\beta$  and  $PSRF.V$  parameters (for more details see (Tikhonov et al. 2020)). We chose model 1, which was the  
247 best compromise between number of OTUs and explanatory power (Table S1), with a thinning of 100 (Table S2).  
248 The post estimates of the model are displayed in Table S2: we selected the ones having a support higher than 0.95.  
249 Statistical analyses were done in R (R core team 2019), using the Hmsc package for the joint species distribution  
250 modelling part.

251

## 252 **Results**

253

### 254 *Microbial taxonomic diversity*

255

256 There was no significant influence of grass encroachment on both bacterial and fungal diversity as assessed by  
257 number of peaks detected by ARISA in soil samples ( $p>0.05$ , Figure 2). In both cases, however, only a small  
258 number of peaks were detected (0 to 6 for bacteria and 0 to 7 for fungi), which suggests that either many species  
259 did not differ in their amplicon lengths, or only the very most abundant species were detected. This non-significant  
260 trend was however confirmed by deeper investigation on diversity, this time using metabarcoding data, where  
261 diversity was calculated as Shannon or Simpson indices: there was no significant relationship between grass cover



262 in a given quadrat and both bacterial and fungal diversity measured from soil samples taken in the middle of this  
263 quadrat (Figure S3).

264

#### 265 *Microbial community structure*

266

267 Bacterial and fungal communities were very similar across the grass cover gradient, at least at high phylogenetic  
268 level. Bacterial communities were dominated by three phyla, representing 92% of all OTUs: *Proteobacteria*  
269 (46%), *Actinobacteria* (31%) and *Acidobacteria* (16%), of which the relative proportion remained unaffected  
270 throughout the gradient of grass encroachment (Figure S4, top). Fungal communities were dominated by the  
271 *Ascomycetes* phylum (86%), from which *Leotiomycetes* (45%) and *Eurotiomycetes* (23%) were the most abundant  
272 classes, and were also unaffected by the degree of grass encroachment (Figure S4, bottom), though a large number  
273 of OTUs could not be identified at the phylum level (22% of the total reads).

274 This broad analysis may however miss many taxonomical responses to grass colonization at a finer phylogenetic  
275 level. We therefore investigated the relationship between grass dominance and community structure at the OTU  
276 level, by means of a hierarchical modelling of species communities (HMSC), where metabarcoding read counts  
277 were the independent variables, the grass cover in the quadrat from which the soil sample community was  
278 characterized the dependant variable, and sample the random variable. Results showed that grass cover was  
279 significantly positively correlated ( $p < 0.05$ ) with the abundance of 5 OTUs, all belonging to the fungal kingdom (1  
280 *Basidiomycete*, 4 *Ascomycetes*), and negatively with 3 OTUs, all belonging to the bacterial kingdom (Figure 3).  
281 The 5 fungal OTUs were attributed to the *Tremellaceae* (*Saitozyma podzolica*), *Dermataceae*, *Coniochaetaceae*,  
282 *Herpotrichiellaceae*, and *Teratosphaeriaceae* (*Devriesia* sp.) families; the 3 bacterial OTUs all belonged to the  
283 genus *Acidothermus*. These 8 OTUs were all abundant to very abundant, accounting altogether for 2% of all reads.  
284 Additionally, all ranked among the 15% most abundant OTUs, 4 of them being in the top 5% (*Devriesia* sp.,  
285 *Herpotrichiellaceae*, and two *Acidothermus* species). Additionally, these OTUs were almost always the most  
286 abundant among the ones that shared the same phylogenetic assignment: there were in total 5 OTUs attributed to  
287 *Devriesia* sp., 14 to *Coniochaetaceae*, 41 to *Herpotrichiellaceae*, 2 to *Saitozyma podzolica*, 12 to *Dermataceae*  
288 and 112 to *Acidothermus*.

289 Some OTUs were attributed to the *Glomales*, but were present in only one sample. The OTUs belonging to  
290 *Helotiales* were by far the most abundant of all microbes. The ones assigned to *Rhizoscyphus* or *Pezoloma ericae*  
291 were not correlated with the grass cover.

292

#### 293 *Microbial metabolic activity*

294

295 Microbial activity, assessed using FDA assay, was not significantly correlated with grass dominance ( $p > 0.05$ ,  
296 Figure S5). The absorbance of the FDA solution at 690 nm oscillated between 0.3 and 1.2, with a high variability  
297 between samples.

298

#### 299 *Microbial C mineralization functions*

300

301 We found that the diversity of C substrates used was not significantly correlated with grass cover ( $p > 0.05$ , Figure  
302 S6). However, there was a significant, positive correlation ( $p < 0.05$ ) between grass dominance and the rate of C  
303 mineralization, with an estimate of 22, which corresponds to a 2.6 factor increase from pure heather to pure grass  
304 (Figure S7); moreover, this correlation was significant across all 6 C source types (Figure 4). The estimates of the  
305 linear model were the highest for amino acids (Pearson's  $R^2 = 0.85$ , an increase of a factor 2.7 throughout the  
306 gradient) and carbohydrates (0.75, 2.3), and to a lower extent, carboxylic acids (0.45, 1.8) and polymers (0.36,  
307 1.6). Mineralization rate of phenolics and amines were not significantly affected by the grass cover (Figure 4).

308

## 309 Discussion

310 We investigated how replacement of the dominant plant affected microbial community structure and function,  
311 using heathland ecosystem as a case study. There, the dominant ericoid plant (*C. vulgaris*) is replaced by grasses  
312 (*M. caerulea*). Our hypothesis was that, as grass becomes dominant, i) litter quality improves, which should favour  
313 fast-growing species thriving on easily decomposable organic matter, and ii) ericoid mycorrhizal fungi should  
314 become less dominant. To test this hypothesis, we set up a field observatory of 13 heathland plots, varying in the  
315 levels of grass dominance, and organized them in a gradient. In each plot, we set up 3 quadrats, in which we  
316 measured plant cover (in order to determine local plant dominance) and took a soil sample from which we  
317 measured microbial metabolic activity, microbial taxonomic diversity, microbial community structure, as well as  
318 functional diversity and rates of C mineralization.

319

### 320 *Microbial community structure was affected only at the level of some abundant OTUs*

321 We found that a shift in vegetation had only limited impact on microbial community structure: microbial  
322 taxonomic and functional diversity were not significantly different, nor were the structure of the communities at  
323 high phylogenetic level, and the overall metabolic activity of microbes. This was surprising. We expected a change  
324 in litter quality to lead to a change in microbial community structure, since C availability is among the most  
325 important structuring factors for microbial communities (Fierer 2017). Increase in C availability has been for  
326 example shown to promote abundance of *β-Proteobacteria* and *Bacteroidetes* at the expense of *Acidobacteria* in  
327 a grassland ecosystem (Fierer et al. 2007), while the proportion of these phyla remained the same across the grass  
328 dominance gradient in our study. However, OTUs belonging to the genus *Acidotherrmus* were at the same time  
329 abundant and negatively correlated with grass dominance. This genus has been investigated for its potential to  
330 produce cellulases with high thermal stability, as well as for its thermophilic and acidophilic growth optima  
331 (Mohagheghi et al. 1986). This suggests that this genus is particularly sensitive to changes in soil microhabitat  
332 induced by replacement of shrubs by grasses.

333 The response of fungal OTUs to grass colonization was characterized by an increased abundance of black yeasts  
334 (*Hepotrichiellaceae*, genus *Devriesia*, and *Saitozyma podzolica*). More precisely, members of  
335 *Hepotrichiellaceae* are black yeast anamorphs known for being extremotolerant (Untereiner et al. 1995). The  
336 members of the genus *Devriesia*, often found as a plant endophytes (Crous and Groenewald 2011), are soil-borne  
337 heat-resistant black yeasts related to the genus *Cladosporium* (Seifert et al. 2004). Finally, the species *Saitozyma*  
338 *podzolica* is a frequent oleaginous basidiomycete black yeast commonly found in soils (Moreira et al. 2020),  
339 especially in heathlands (Op De Beeck et al. 2015). It has been several times associated with vegetation change,  
340 for example after mining activity, but also in human disturbed forests, where the abundance of this species was

341 negatively correlated with the disturbance (Moreira et al. 2020). In that case the presence of grass was used as a  
342 proxy for disturbance. (Yurkov 2018) noted that this species is usually a good marker of acid, well-drained soils.  
343 Measurements in soils in nearby plots (data not shown) does however not support the evidence that acidity and  
344 soil drainage increased with grass colonization, so the reasons for increase in relative abundance of *S. podzolica*  
345 probably lie somewhere else. The two other OTUs that were significantly (and positively) affected by grass  
346 colonization belong to the *Dermataceae*, which are most often associated with roots of Ericaceous plants (Obase  
347 et al. 2009), and to the *Coniochaetaceae*, which are often plant endophytes or pathogens, and more numerous in  
348 extreme habitats (Chen et al. 2021).

349 Most of microbial OTUs, however, remained unaffected by grass colonization. Many phylogenetic groups known  
350 to be associated with easily decomposable C (such as genera *Penicillium* or *Trichoderma*) had the same relative  
351 abundance throughout the gradient, as well as oligotrophs such as *Mycena*, *Actinobacteria*, or *Deltaproteobacteria*,  
352 which confirms that there was no shift in the oligotrophic / copiotrophic ratio, contrary to what we hypothesized.  
353 This result implies that litter quality either did not change significantly across the gradient, or that it had no impact  
354 on the abundance of these species. Moreover, we expected at least a shift towards a replacement of ericoid  
355 mycorrhizal genera (*Rhizoscyphus*, associated with roots of ericoid plants) by arbuscular ones (*Glomus*, associated  
356 with roots of most grasses) (Read et al. 2004). The primer pair we used picked up only 60 reads for Glomales,  
357 which may be due to a poor performance at amplifying this phylogenetic group, and therefore we cannot draw  
358 conclusions on arbuscular mycorrhizal fungal (AMF) abundance. However, we definitely did not observe a drop  
359 in the relative abundance of *Rhizoscyphus* and *Pezoloma* as grasses became more dominant. On a side note, we  
360 identified no *Archaeorhizomyces* sequences in our metabarcoding results, contrary to what was found in similar  
361 heathlands (Radujković et al. 2020), or even in samples taken less than a hundred meters away from two of the  
362 plots of this study (data not shown). Again, the primer pair used here could be the explanation, as ITS4 mismatches  
363 with *Archaeorhizomyces* sequences (Ihrmark et al. 2012).

364 It seems difficult to synthesize all of these contradictory data in a coherent picture. What is clear is that there is no  
365 evidence that grass colonization increased abundance of copiotrophic species, nor of species commonly associated  
366 with higher litter or soil quality. Instead, we see a community shift, where some abundant strains of extremotolerant  
367 bacterial species (*Acidothermus*) are progressively replaced by others fungal ones (black yeasts), and ericoid  
368 mycorrhizal fungi seem unaffected.

369

#### 370 *Possible influence of the 2018 drought*

371 Heathland soil macrocosms taken from the same area were following ambient conditions measured at an ICOS  
372 ecosystem tower located on the same site (Rineau et al. 2019). The response of these 12 macrocosms to 2018  
373 drought was very close to what we observed in the field. We have therefore precise measures of precipitation and  
374 of the top 10 cm soil moisture on that date. They show that the soil moisture was low (6% +/- 1 SE, n=36) despite  
375 re-wetting events before sampling (3 and 10mm on the 07/08 and 09/08, respectively), because it was preceded by  
376 a long 2-month period where there was no rain event (01/06-06/08). The community at the date of sampling is  
377 therefore likely affected by this low water availability. Drought is usually reported to select for Gram + bacteria  
378 (Firmicutes and Actinomycetes in particular) against Gram – ones (Proteobacteria, Verrucomicrobia and  
379 Bacteroidetes), that are less tolerant to stress (Fierer et al. 2003; Naylor and Coleman-Derr 2018). And indeed, we  
380 see that the Gram + / Gram - ratio based on the abovementioned bacterial phyla is twice higher in our samples than

381 the ones reported by (Seaton et al. 2021) on a very similar upland heathland. This suggests that the proportion of  
382 Gram + bacteria in particular, and of extremophiles in general is higher in this study than what heathland soils may  
383 harbour outside of extreme events; but also, that community shifts may be larger in non-extreme weather  
384 conditions, as plant properties would proportionally contribute more to shaping soil microbe communities when  
385 selection by water availability is relieved.

386

387 *Was our main hypothesis rejected because of wrong assumptions?*

388 As our hypothesis was rejected, we may need to take a step back and evaluate the accuracy of our assumptions.  
389 The basis of our reasoning is that increase in grass dominance improves C availability. Literature supports the  
390 assumption that *M. caerulea* litter inputs brings higher amounts of more available C to soil than *C. vulgaris*,  
391 through both above (Van Vuuren and Berendse 1993; Certini et al. 2015) and belowground inputs (Aerts et al.  
392 1992). However, we do not know exactly how this biomass interacts with organic matter that has been shaped by  
393 decades of dominance by ericoid shrubs. Indeed, litter mixes have different biochemical properties that influence  
394 each other's decomposition rate: their dynamics of decomposition is therefore non-additive (Gartner and Cardon  
395 2004). Heathland ecosystem is no exception: *M. caerulea* litter degrades significantly slower under *C. vulgaris*  
396 than under itself (Certini et al. 2015). We therefore cannot rule out that the increase in available C coming from  
397 the grass litter is inhibited by some compounds originating from long-term accumulated heather litter in the organic  
398 topsoil, such as polyphenols (Kraus et al. 2003). The soil properties would then remain the same, and so would  
399 most of the microbial community. Moreover, the literature that we use as comparison to understand the effects of  
400 soil properties on microbial communities may represent a much more extreme case than grass colonization in our  
401 field sites. For example, (Fierer et al. 2007) amended their soil samples with large amounts (up to 800 g of  
402 equivalent C/m<sup>2</sup>/y) of a C source of high quality (sucrose), and (Zhou et al. 2021) with 120 µg/g of soil of glucose  
403 carbon. While litter input from a grass species differ from shrubs mostly by the relative abundance of structural  
404 compounds, with more crystalline cellulose, and less aromatic compounds (Certini et al. 2015). This improves  
405 litter quality, but to a much lower extent than glucose or sucrose, as polysaccharides need to be processed by  
406 hydrolytic and/or oxidative enzymes before being incorporated for microbial metabolism, which is energy  
407 demanding. Hence it appears logical to expect less drastic shifts in microbial communities in our study than in  
408 (Fierer et al. 2007) and (Zhou et al. 2021). Finally, the absence of response of the ericoid genus *Rhizoscyphus* to  
409 grass dominance could be caused by a slower necromass or propagule decomposition than other species (Lenaers  
410 et al. 2018), which would artificially maintain its relative abundance figures in metabarcoding surveys (since it is  
411 based on DNA which can be found in necromass and spores). Alternatively, that could be explained by the ability  
412 of these species to persist in soil as saprotrophs after their host disappeared (Read et al. 2004).

413

414 *Functional responses to grass colonization were more clear-cut than structural ones*

415 Contrary to microbial community structure, C mineralization functions responded significantly to the increase in  
416 grass cover. Even though the overall metabolic activity of the microbial communities remained the same, the  
417 mineralization rate of tested C sources increased by an estimated factor of 2.7 across the full range of grass cover.  
418 In other words, similar microbial assemblages showed almost three times more C mineralization potential under  
419 pure grass than under pure heather, and this was not because microbes were just more metabolically active, as  
420 shown by the results of the FDA assay. We cannot rule out, however, that the fungal OTUs favoured by grass

421 colonization (belonging to black yeasts, *Dermataceae*, *Coniochaetaceae*) contributed to this higher activity. In  
422 fact, the number of read counts of *S. podzolica* was significantly correlated with the mineralization of D-cellobiose,  
423  $\alpha$ -ketobutyric acid, 2-hydroxybenzoic acid, and  $\beta$ -methyl-D-glucoside; and the number of read counts of the OTU  
424 belonging to *Devriesia sp.* was positively correlated with the mineralization rate of  $\beta$ -methyl-D-glucoside (data  
425 not shown). The mineralization rates may nevertheless result from more complex inhibitory or synergistic  
426 functional interactions.

427 Hence, the communities present in this heathland soil were adapted to these soil conditions, and responded to  
428 change in litter input by raising their ability to mineralize C. In particular, grass dominance stimulated  
429 mineralization rates of carbohydrates, amino acids, carboxylic acids, and polymers. The former three are  
430 characteristic of root exudates (Griffiths et al. 1998), and the latter of plant biomass. We therefore speculate that  
431 the increase in grass dominance stimulates C mineralization of root exudates, and to a lower extent, litter biomass;  
432 but that litter input is in too low quantity relative to the soil organic matter pool to significantly affect its  
433 composition, and therefore microbial community structure, before a decade or more. It also implies that the large-  
434 scale grass colonization we see in many North-Western European heathlands potentially results in significant  
435 increases in C mineralization rates, which has negative consequences for this ecosystem's C sequestration.  
436 However, this case study has been conducted in only one site, albeit a large one (287 500 m<sup>2</sup>), and different  
437 responses may be observed in other heathlands because of site-to-site variability.

438

#### 439 *Methodological considerations*

440 For practical reasons, samples could not be analyzed fresh, and had to be stored long-term at -20C. Such storage  
441 conditions usually do not alter microbial community structure (Wallenius et al. 2010). They are however known  
442 to affect soil enzyme activities (Peoples and Koide 2012), decreasing them by about 20-30% (Wallenius et al.  
443 2010), likely through a decrease in viable microbial cells (Shishido and Chanway 1998). Nevertheless, since all  
444 samples have been stored the same way, the relative differences between samples may be less biased. The method  
445 used to measure the mineralization rates is also not without drawbacks: just like most in vitro soil enzyme assays,  
446 it measures only potential activities (because they are measured in optimal conditions of moisture, temperature  
447 and substrate availability); it is also influenced by soil water content and inoculum density (Preston-Mafham et al.  
448 2002). The C mineralization and FDA results have therefore to be interpreted with caution. The responses may  
449 especially reflect those of bacteria, and may not fully represent field microbial communities. Still, they are  
450 apparently large and consistent enough to be detectable despite noise associated with methodology.

451

#### 452 *Conclusion*

453 Our results show that microbial community structure is only moderately affected by grass colonization, with the  
454 exception of a limited number of abundant OTUs. However, this is not the case for community functioning: C  
455 mineralization potential significantly increases, almost by a factor 3 overall. This was especially clear for the  
456 substrates that were chemically related to root exudates, and to a lower extent the ones related to plant litter. This  
457 led us to speculate that the change in dominant plant increased soil C mineralization by microbes through a change  
458 in availability and nature of root exudates, and to a lesser extent litter input. Our study also points to some species  
459 of both bacteria and fungi who were responsive to such a change in plant traits, thereby extending our knowledge  
460 on the otherwise still mysterious ecology of *Acidotherrmus* bacteria and black yeast fungi. Taken altogether, these

461 results show that a change in dominant plant traits may induce drastic functional changes in microbial communities  
 462 despite having only a very minor effects on their diversity or community structure. Moreover, our study implies  
 463 that grass colonization observed in many heathlands in Northwestern Europe leads to higher C mineralization  
 464 rates, which has potentially negative impact on the C sequestration by this ecosystem.

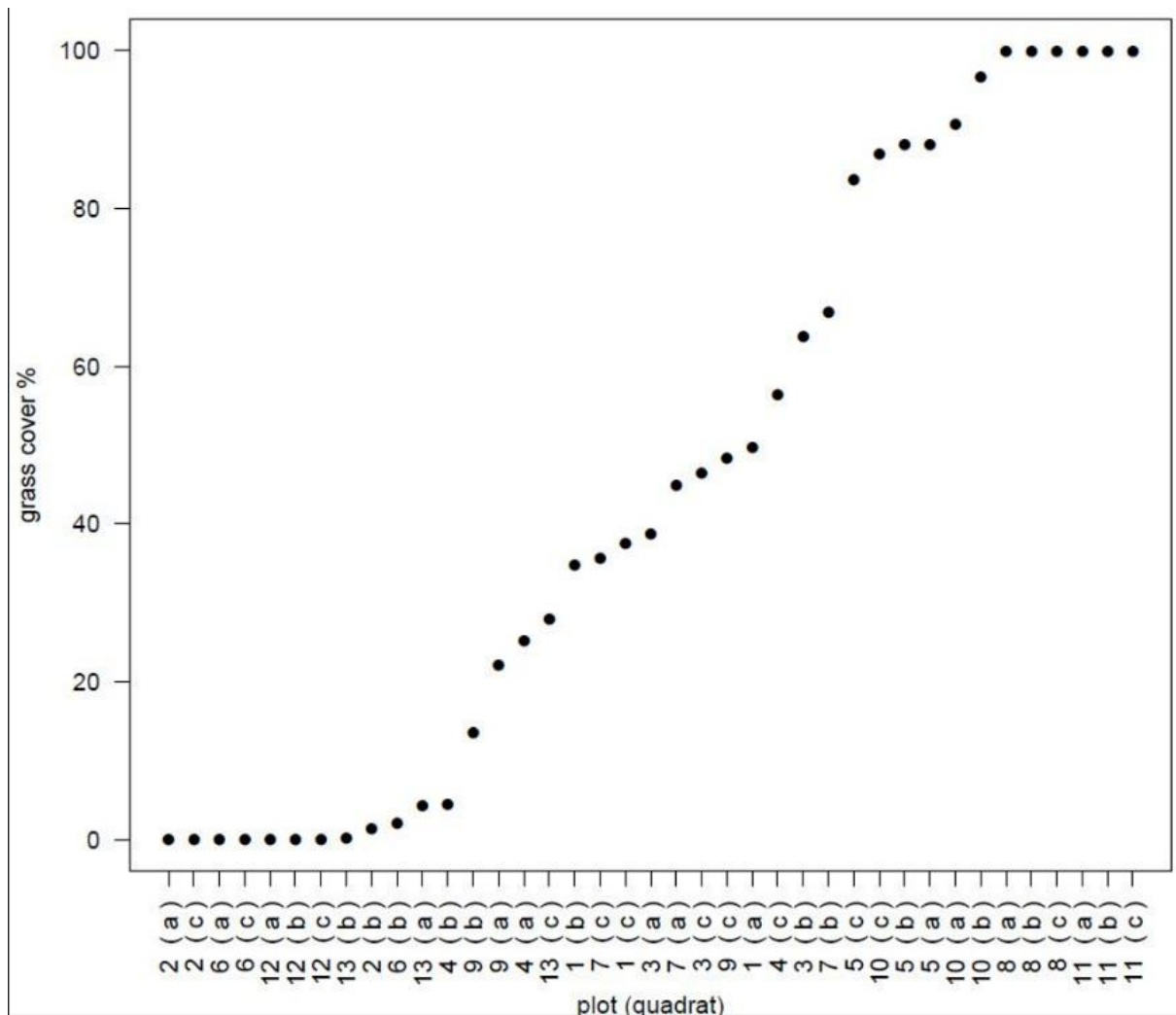
465

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 468 National Park Hoge Kempen for its collaboration, support, and providing access to the sampling area.

469

470 **Figure captions**

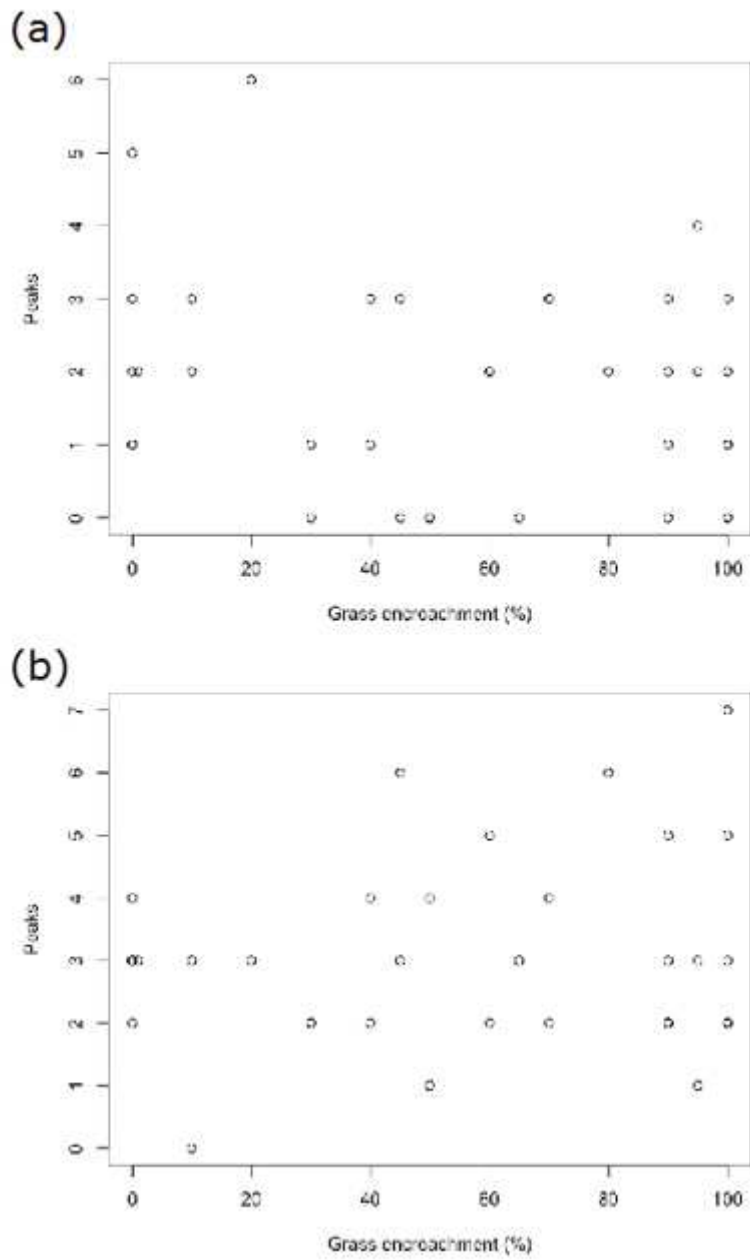


471

472 Figure 1. Grass cover (%) in each quadrat ((a), (b), (c)) of the 13 plots (1-13) of the gradient, sorted by increasing  
 473 order. The heather cover is not represented here for clarity purposes but is the exact opposite of grass cover.

474

475



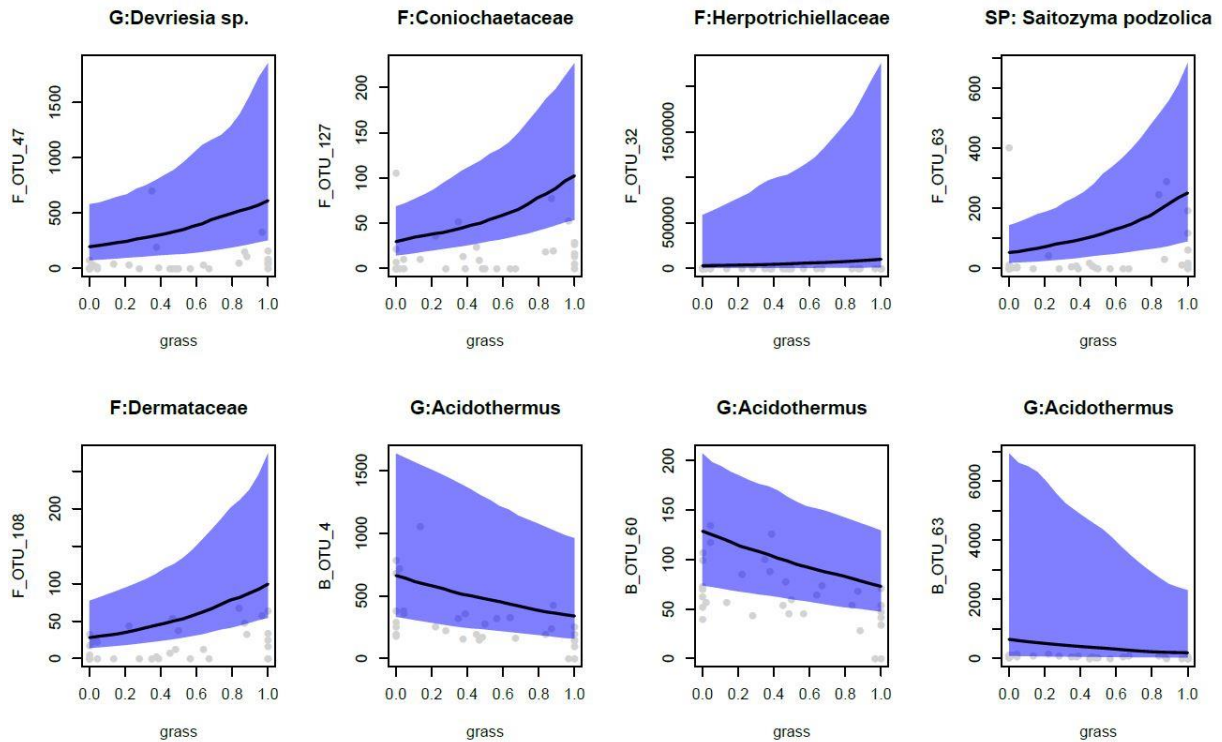
477

478 Figure 2: Effect of grass colonization on the bacterial (a) and fungal (b) diversity in function of grass encroachment.

479 Diversity is expressed as the number of OTU-peaks in bacterial and fungal ARISA fingerprints.

480

481

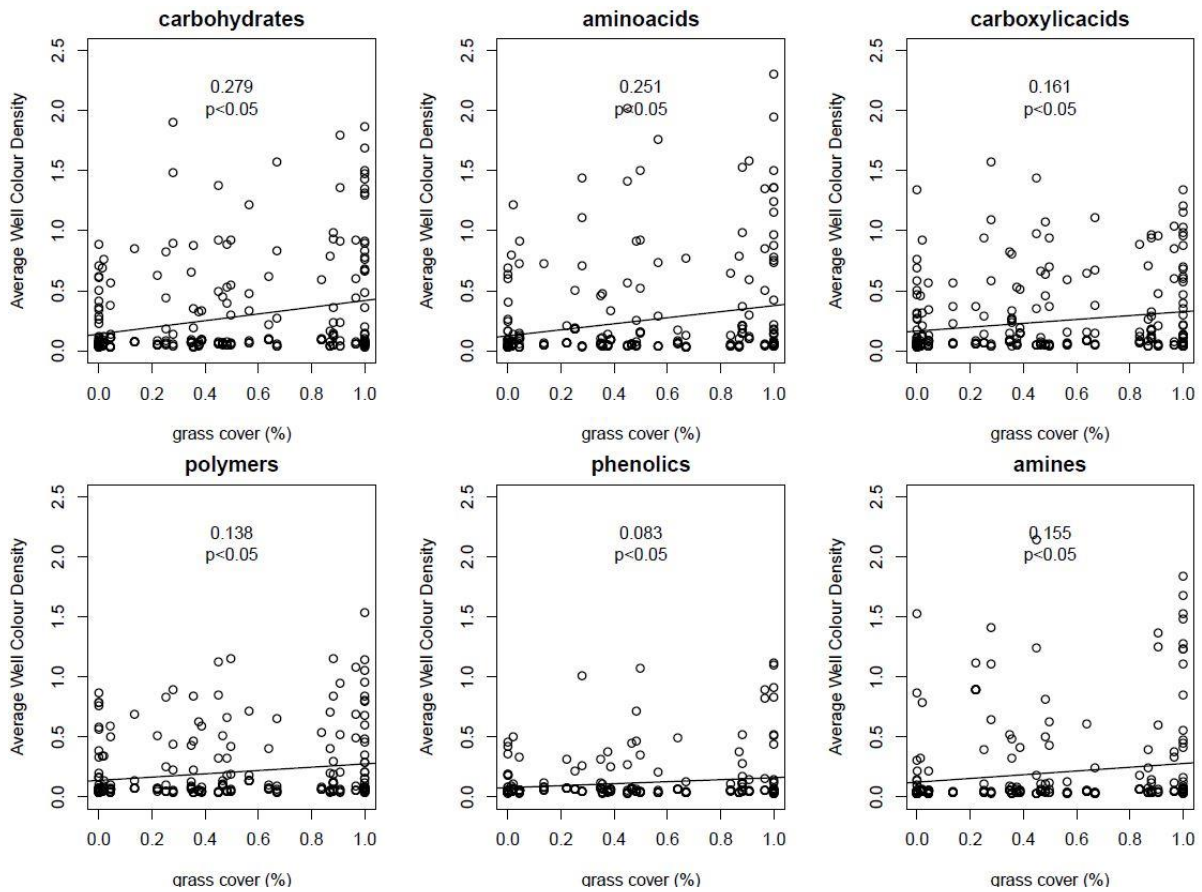


482

483 Figure 3: Effect of grass colonization on abundance of microbial species as predicted by hierarchical modelling of  
 484 species communities. Only the OTUs for which estimates were significant in the model are shown ( $p < 0.05$ ). X-  
 485 axis: grass cover proportion in the quadrat (1=100%). Y-axis: read counts. Black line: model predictions. Blue  
 486 area: 95% confidence interval of the model. Grey dots: actual data. The title of the graph corresponds to the name  
 487 of the OTU's best hit according to the BROCC software pipeline (Dollive et al. 2012) at the highest phylogenetic  
 488 resolution possible (F: Family, G: Genus, SP: Species). The OTU ID appears as the title of the y axis of each graph  
 489 (F: fungus, B: bacteria; the smaller the OTUs ID number, the more reads).

490





491

492

Figure 4. Relationship between grass cover (0=no grass, 1= 100% grass) and mineralization rate of six types of

493 carbon substrates. The mineralization rate is measured spectrophotometrically as the reduction of a tetrazolium

494 dye by the respiration of microbial cells having a single C substrate in a microplate well. The microbial cells have

495 been inoculated from a soil sample. We used a linear model and Bonferroni correction to test the significance of

496 the relationship between the two variables. Significant relationships are labelled with “ $p < 0.05$ ”. The slope of the

497 relationship is given above the significance level.

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638 Data availability: The datasets generated during and/or analysed during the current study are available from the  
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