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Change in heathland dominant plants strongly increases C mineralization potential despite marginally affecting microbial community structure Peer-reviewed author version

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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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#### 22 Abstract

23 Purpose

In many ecosystems, the identity of the dominant plant is changing because of global change. If the new dominant species has different litter and root traits than the one it replaces, it is likely to have an influence on soil microbial

- 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

We established a 13-plot field observatory spanning across a 0-100% gradient of grass cover percentage. In each
plot, we characterized plant, fungal and bacterial communities, using a combination of ARISA (taxonomic
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, C47 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
- 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

grasses (Aerts et al. 1992). Heather litter is much more lignified than that of grasses, resulting in lower litter quality

- and twice higher C/N ratios under C. vulgaris than under M. caerulea (Certini et al. 2015). Finally, M. caerulea
- and *D. flexuosa* form arbuscular mycorrhizal associations, and *C. vulgaris* ericoid ones (Wang & Qiu, 2006). Both
   mycorrhizal types significantly differ in their functional profiles, with for example ericoid mycorrhizae having
- mycorrhizal types significantly differ in their functional profiles, with for example ericoid mycorrhizae havinghigher 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  $(50^{\circ}59'07.0"N, 5^{\circ}38'01.7"E)$ , at an altitude of 104 m above sea level. The mean annual temperature is  $10.3^{\circ}C$  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  $500-1000m^2$  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

types within this area are albic podzols and brunic-dystric arenosols. We chose 13 of these plots, that were similar

- in terms of vegetation age (5-10 years old, in the early building-up phase), density, management history (burning),
- 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

The soil samples were collected on the 13th of august 2018, after a few days of rain in a summer characterised by intense droughts. In each of the 13 plots, we took soil samples to assess features of microbial communities, with 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 the vegetation cover at the exact spot where the soil sample was taken. Indeed, the plant cover was heterogeneous at the sub-meter scale, especially in the plots in the middle of the gradient, so there was a significant probability

- 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
- other vascular plant species in the quadrats). In the middle of each quadrat, we took a 10 cm deep and 10 cmdiameter soil core, using an auger, that was sterilized with 70% ethanol between each core. The soil sample was
- 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
  - Soil DNA was extracted using the Dneasy® PowerSoil® Kit (QIAGEN, Venlo, The Netherlands). The 16S-23S
    and 18S-28S intergenic spacer region of bacteria and fungi, respectively, were then amplified by PCR. The
  - 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. The bacterial primers that were used, were the
  - 132 S-D-Bact-1522-b-S-20 (5'-TGCGGCTGGATCCCCTCCTT-3') and L-D-Bact-132-a-A-18 (5'-
  - 133 CCGGGTTTCCCCATTCGG-3') primers. The fungal primers that were used, were the 2234C (5'-
  - 134 GTTTCCGTAGGTGAACCTGC-3') and 3126T (5'-ATATGCTTAAGTTCAGCGGGT-3') primers. Both these
  - primer sets were used and recommended for ARISA by (Ranjard et al. 2001). The PCR was executed in a PCR
  - thermocycler Biorad T100 or Biorad C1000 (Biorad, Temse, Belgium). First, a hotstart of 3 minutes was performed
  - 137 at 94°C, followed by 25 cycles at 94°C for 1 minute, 55° for 30 seconds and 72°C for 1 minute. Finally, a terminal
  - elongation step was performed at 72°C for 5 minutes (37).
  - 139 The resulting PCR-amplicons were used for ARISA to generate an electropherogram. This is used to determine
  - 140 the microbial fingerprint of the soil sample (Ranjard et al. 2001). ARISA on each sample was done using an Agilent
  - 141 2100 Bioanalyzer (Agilent Technologies, Santa Clara, California, US). An Agilent DNA 1000 Kit was used with

the Agilent DNA-chips. Amplicons were separated by length through capillary electrophoresis on the microchip.

143 ROX gel-dye was used as an internal standard. After adjusting the gel-dye mix to room temperature for 30 minutes,

144  $9 \mu l$  of gel-dye mix was added to the 3 designated wells on the microchip. One of the wells was firmly put under

145 pressure with the chip priming station. Next, 5 µl marker was added to the wells designated to the ladder and

146 samples. After running the chip in the Bioanalyzer, the results were presented in the form of gels and

147 electropherograms in the Agilent Expert 2100 software. DNA fragments are presented as peaks in the

- electrophoregram.
- 149

# **150** *Characterization of microbial community structure*

151 The same DNA samples were used as for ARISA. DNA samples were subjected to bacterial 16S rRNA gene and 152 fungal ITS2 region sequencing. For the bacteria, in the first round of 16S rRNA gene PCR, an amplicon of 290 bp 153 was generated, using primers 515F and 806R (Caporaso et al. 2010). For the first round of Fungi-specific PCR, 154 the primers gITS87f and ITS4R generate an amplicon with an average length of 450 bp. Using the Q5 High-Fidelity DNA Polymerase system (M0491, New England Biolabs, Ipswich, Massachusetts, US), a reaction volume 155 156 of 25 µl per sample was prepared containing 1 µl of extracted DNA (final DNA-concentration per reaction 1-10 ng), 1x Q5 Reaction Buffer with 2 mM MgCl2, 200 µM dNTP mix, 1x Q5 High GC Enhancer (for the soil and 157 158 fungi samples), 0.2 µM forward and reverse primer, and 1.2 U Q5 High-Fidelity DNA polymerase. The PCR 159 program started with an initial denaturation for 3 min at 98 °C, followed by a 30 s denaturation at 98 °C, a 30 s 160 annealing at 53 °C for V4 (58 °C for ITS) and a 1 min extension at 72 °C, all three steps were repeated for a total 161 of 35 cycles. The reaction was ended by a final 7 min extension at 72 °C. The amplified DNA was purified using 162 the AMPure XP beads (Beckman Coulter, Brea, California, US) and the MagMax magnetic particle processor 163 (Thermo Fisher Scientific, Waltham, Massachusetts, US). Subsequently, 5 µl of the cleaned PCR product was used 164 for the second PCR, attaching the Nextera indices (Nextera XT Index Kit v2 Set A (FC-131-2001), and D (FC-131-2004), Illumina, San Diego, California, US). For these PCR reactions, 5 µl of the purified PCR product was 165 166 used in a 25 µl reaction volume, and prepared following the 16S Metagenomic Sequencing Library Preparation 167 Guide. PCR conditions were the same as described above, but the number of cycles reduced to 20, and 55 °C 168 annealing temperature. PCR products were cleaned with the AMPure XP kit, and then quantified using the Qubit dsDNA HS assay kit (Thermo Fisher Scientific) and the Qubit 2.0 Fluorometer (Thermo Fisher Scientific). Once 169 170 the molarity of the sample was determined, the samples were diluted down to 4 nM using 10 mM Tris pH 8.5 prior 171 to sequencing on an Illumina MiSeq. Samples were sequenced using the MiSeq Reagent Kit v3 (600 cycle) (MS-172 102-3003) and 15% PhiX Control v3 (FC-110-3001). For quality control, a DNA-extraction blank and PCR blank 173 were included throughout the process, and also the ZymoBIOMICS Microbial Mock Community Standard 174 (D6300) to test efficiency of DNA extraction (Zymo Research, Irvine California, US). Obtained sequences were 175 clustered into operational taxonomic units (OTUs) and annotated using Qiime (Caporaso et al. 2010) within the 176 DADA2 package with standard settings, with the SILVA database for bacteria and the Unite 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
- 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  $\mu$ L) was transferred to a 96 well plate and the
- 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 Biolog PM1 MicroPlate Carbon 194 Sources (BioLog Inc., Hayward, California, US) test. Using sterilized equipment, aliquot soil samples (1g) were dissolved in 10 mL sterile 10 mM PBS buffer (130 mM NaCl, 7 mM Na<sub>2</sub>HPO<sub>4</sub>, 3 mM NaH<sub>2</sub>PO<sub>4</sub>, pH 7.4) and 195 196 shaken for 20 minutes at room temperature. After shaking, soil particles were allowed to settle for 30 minutes at 197 4°C. Subsequently, 130 µL of the supernatant was dispensed into each well of the Biolog MicroPlate. Inoculated 198 plates were placed in self-sealing plastic bags containing a water-soaked paper towel to minimize evaporation from 199 the wells, and incubated at 28-30 °C. Absorbance was measured at 595 nm with a plate reader immediately after 200 inoculation (0 h) and at 3, 6, 18, 24, 48, 72, and 144 h. Raw absorbance values were recorded at each time point 201 and individually standardized by subtracting the corresponding absorbance value measured immediately after 202 inoculation (0 h) (reaction-independent absorbance). Furthermore, to semi-quantitatively evaluate the kinetic 203 Biolog MicroPlate dataset, the net area under the absorbance versus time curve was calculated according to the 204 trapezoidal approximation (Guckert et al. 1996). The resulting value calculated via the trapezoidal approximation 205 summarizes different aspects of the measured respirometric reaction, including differences in lag phases, increase 206 rates, or maximum optical densities. The Shannon-Weaver index (H) was calculated per sample as an indicator of 207 functional diversity, or range of C sources used by the community of microbes. The rate of C source use by the 208 microbial community was calculated as the average of the area under the curve for all C sources (referred to as 209 AWCD: average well colour development). The C sources could be categorized into 6 different types, based on 210 (Rutgers et al. 2016): carbohydrates (β-methyl-D-glucoside, n-acetyl-D-glucosamine, D-cellobiose, glucose-1-211 phosphate, a-D-lactose, D-1-a-glycerol-phosphate, D-galactonicacid-1-lactone, D-xylose, 1-erythritol, Dmannitol), amino acids (L-arginine, L-asparagine, L-serine, L-phenylalanine, L-threonine), carboxylic acids 212 213 (glycyl-L-glutamic acid, D-galactonic acid,  $\gamma$ -hydroxybutyric acid, D-glucosamic acid, itaconic acid, alpha-214 ketobutyric acid, pyruvic acid methyl ester, D-malic acid), polymers (tween40, tween80, α-cyclodextrin, 215 glycogen), phenolics (2-hydroxy-benzoic acid, 4-hydroxy-benzoic acid), and amines (phenylethylamine, putrescine). 216

217

218 Statistics

219 The strength of the relationship between grass colonization and both microbial diversity and activity was assessed

220 using correlation tests. Correlation between grass cover and both bacterial and fungal microbial diversity was

tested in three ways: with OTU richness (number of peaks in the ARISA electropherogram) using a Kendall rank

correlation coefficient test, after normality and homoscedasticity assumptions were tested and not met; and with Shannon and Simpson indices (diversity indices based on OTU richness and abundance, assessed by metabarcoding, see above) using a Pearson's correlation test. The correlation between grass cover and microbial activity (measured as the absorbance at 690 nm in the FDA assay, and where no distinction between bacteria and fungi can be made) was tested using a Pearson's correlation test.

227 To evaluate the effect of grass colonization on the microbial community structure, we used Hierarchical Modelling 228 of Species Communities (HMSC) (Tikhonov et al. 2020), which is a type of joint species distribution modelling 229 (Wilkinson et al. 2021). The advantage of these models over PERMANOVA, ANOSIM or a Mantel Test is that 230 HMSC explicitly accounts for the effect of species-species interactions on their distribution over an environmental 231 variable (Tikhonov et al. 2019a), to reduce the effect of species interactions as a confounding variable. We built 232 three separate mixed models, all with quadrat's grass cover as independent variable and quadrat identity as random 233 variable; only the dependant variable (OTU abundance as counts) differed between the models. Model 1 grouped 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 234 235 39). The rationale behind testing these three models was that we wanted to find a compromise between including 236 as many OTUs in the analysis as possible, on the one hand, and keeping the highest explanatory power, on the 237 other hand. We then fitted all models using Bayesian statistics. For that, we sampled the posterior distribution 238 using two Monte Carlo Markov Chains (MCMC), each of which being run for 5000 iterations, out of which the 239 first 4900 were removed as burn-in and the remaining ones thinned by either 1, 10, or 100. We then evaluated and 240 compared model's explanatory power using the Root Mean Square Error (RMSE), coefficient of determination 241 (SR<sup>2</sup>), area under the receiver operating characteristic curve (AUC) and coefficient of discrimination (Tjur<sup>2</sup>) (see 242 results in Figure S2); and verified that convergence of the Markov chains was achieved by comparing  $ess.\beta$ , ess.V, 243 PSRF. $\beta$  and PSRF.V parameters (for more details see (Tikhonov et al. 2019b)). We chose model 1, which was the 244 best compromise between number of OTUs and explanatory power (Table S1), with a thinning of 100 (Table S2). 245 The post estimates of the model are displayed in Table S2: we selected the ones having a support higher than 0.95. 246 Statistical analyses were done in R (R core team 2019), using the Hmsc package for the joint species distribution 247 modelling part.

248

249 Results

250

# 251 Microbial taxonomic diversity

252

253 There was no significant influence of grass encroachment on both bacterial and fungal diversity as assessed by 254 number of peaks detected by ARISA in soil samples (p>0.05, Figure 2). In both cases, however, only a small 255 number of peaks were detected (0 to 6 for bacteria and 0 to 7 for fungi), which suggests that either many species 256 did not differ in their amplicon lengths, or only the very most abundant species were detected. This non-significant 257 trend was however confirmed by deeper investigation on diversity, this time using metabarcoding data, where 258 diversity was calculated as Shannon or Simpson indices: there was no significant relationship between grass cover 259 in a given quadrat and both bacterial and fungal diversity measured from soil samples taken in the middle of this 260 quadrat (Figure S3).

261

262 *Microbial community structure* 

263

264 Bacterial and fungal communities were very similar across the grass cover gradient, at least at high phylogenetic 265 level. Bacterial communities were dominated by three phyla, representing 92% of all OTUs: Proteobacteria 266 (46%), Actinobacteria (31%) and Acidobacteria (16%), of which the relative proportion remained unaffected 267 throughout the gradient of grass encroachment (Figure S4, top). Fungal communities were dominated by the 268 Ascomycetes phylum (86%), from which Leotiomycetes (45%) and Eurotiomycetes (23%) were the most abundant 269 classes, and were also unaffected by the degree of grass encroachment (Figure S4, bottom), though a large number 270 of OTUs could not be identified at the phylum level (22% of the total reads). 271 This broad analysis may however miss many taxonomical responses to grass colonization at a finer phylogenetic 272 level. We therefore investigated the relationship between grass dominance and community structure at the OTU 273 level, by means of a hierarchical modelling of species communities (HMSC), where metabarcoding read counts 274 were the independent variables, the grass cover in the quadrat from which the soil sample community was 275 characterized the dependant variable, and sample the random variable. Results showed that grass cover was 276 significantly positively correlated (p<0.05) with the abundance of 5 OTUs, all belonging to the fungal kingdom (1 277 Basidiomycete, 4 Ascomycetes), and negatively with 3 OTUs, all belonging to the bacterial kingdom (Figure 3). 278 The 5 fungal OTUs were attributed to the Tremellaceae (Saitozyma podzolica), Dermataceae, Coniochaetaceae, 279 Herpotrichiellaceae, and Teratosphaeriaceae (Devriesia sp.) families; the 3 bacterial OTUs all belonged to the 280 genus Acidothermus. These 8 OTUs were all abundant to very abundant, accounting altogether for 2% of all reads. 281 Additionally, all ranked among the 15% most abundant OTUs, 4 of them being in the top 5% (Devriesia sp., 282 Herpotrichiellacaeae, and two Acidothermus species). Additionally, these OTUs were almost always the most 283 abundant among the ones that shared the same phylogenetic assignment: there were in total 5 OTUs attributed to 284 Devriesia sp., 14 to Coniochaetaceae, 41 to Herpotrichiellaceae, 2 to Saitozyma podzolica, 12 to Dermataceae 285 and 112 to Acidothermus. 286 Some OTUs were attributed to the *Glomales*, but were present in only one sample. The OTUs belonging to 287 Helotiales were by far the most abundant of all microbes. The ones assigned to Rhizoscyphus or Pezoloma ericae 288 were not correlated with the grass cover.

- 289
- 290 Microbial metabolic activity
- 291

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

- 295
- 296 Microbial C mineralization functions

297

The functions of C mineralization by soil microbes were investigated using community-level physiological profiling (BiologEcoplates), where an inoculum prepared from a given soil sample was exposed to 32 carbon sources, and the mineralization rate measured spectrophotometrically as the reduction of a tetrazolium dye by microbial cell respiration, integrated over time using trapezoidal estimation. We found that the diversity of C

- 302 substrates used was not significantly correlated with grass cover (p>0.05, Figure S6). However, there was a
- 303 significant, positive correlation (p<0.05) between grass dominance and the rate of C mineralization, with an
- estimate of 22, which corresponds to a 2.6 factor increase from pure heather to pure grass (Figure S7); moreover,

this correlation was significant across 4 out of the 6 C source types (Figure 4). The estimates of the linear model

- 306 were the highest for amino acids (Pearson's  $R^2=0.85$ , an increase of a factor 2.7 throughout the gradient) and
- 307 carbohydrates (0.75, 2.3), and to a lower extent, carboxylic acids (0.45, 1.8) and polymers (0.36, 1.6).
- 308 Mineralization rate of phenolics and amines were not significantly affected by the grass cover (Figure 4).
- 309

### 310 Discussion

- 311 We investigated how replacement of the dominant plant affected microbial community structure and function, 312 using heathland ecosystem as a case study. There, the dominant ericoid plant (C. vulgaris) is replaced by grasses 313 (*M. caerulea*). Our hypothesis was that, as grass becomes dominant, i) litter quality improves, which should favour 314 fast-growing species thriving on easily decomposable organic matter, and ii) ericoid mycorrhizal fungi should 315 become less dominant. To test this hypothesis, we set up a field observatory of 13 heathland plots, varying in the 316 levels of grass dominance, and organized them in a gradient. In each plot, we set up 3 quadrats, in which we 317 measured plant cover (in order to determine local plant dominance) and took a soil sample from which we 318 measured microbial metabolic activity, microbial taxonomic diversity, microbial community structure, as well as 319 functional diversity and rates of C mineralization.
- 320

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

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

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

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

extreme habitats (Chen et al. 2021).

- 350 Most of microbial OTUs, however, remained unaffected by grass colonization. Many phylogenetic groups known 351 to be associated with easily decomposable C (such as genera *Penicillium* or *Trichoderma*) had the same relative 352 abundance throughout the gradient, as well as oligotrophs such as Mycena, Actinobacteria, or Deltaproteobacteria, 353 which confirms that there was no shift in the oligotrophic / copiotrophic ratio, contrary to what we hypothesized. 354 This result implies that litter quality either did not change significantly across the gradient, or that it had no impact 355 on the abundance of these species. Moreover, we expected at least a shift towards a replacement of ericoid 356 mycorrhizal genera (Rhizoscyphus, associated with roots of ericoid plants) by arbuscular ones (Glomus, associated 357 with roots of most grasses) (Read et al. 2004). The primer pair we used picked up only 60 reads for Glomales, 358 which may be due to a poor performance at amplifying this phylogenetic group, and therefore we cannot draw 359 conclusions on arbuscular mycorrhizal fungal (AMF) abundance. However, we definitely did not observe a drop 360 in the relative abundance of Rhizoscyphus and Pezoloma as grasses became more dominant. On a side note, we 361 identified no Archaeorhizomycetes sequences in our metabarcoding results, contrary to what was found in similar 362 heathlands (Radujković et al. 2020), or even in samples taken less than a hundred meters away from two of the 363 plots of this study (data not shown). Again, the primer pair used here could be the explanation, as ITS4 mismatches 364 with Archaeorhizomyces sequences (Ihrmark et al. 2012). 365 It seems difficult to synthesize all of these contradictory data in a coherent picture. What is clear is that there is no
- **366** evidence that grass colonization increased abundance of copiotrophic species, nor of species commonly associated
- 367 with higher litter or soil quality. Instead, we see a community shift, where some abundant strains of extremotolerant
- bacterial species (Acidothermus) are progressively replaced by others fungal ones (black yeasts), and ericoid
- 369 mycorrhizal fungi seem unaffected.
- 370

#### 371 Was our main hypothesis rejected because of wrong assumptions?

- 372 As our hypothesis was rejected, we may need to take a step back and evaluate the accuracy of our assumptions. 373 The basis of our reasoning is that increase in grass dominance improves C availability. Literature supports the 374 assumption that M. caerulea litter inputs brings higher amounts of more available C to soil than C. vulgaris, 375 through both above (Van Vuuren and Berendse 1993; Certini et al. 2015) and belowground inputs (Aerts et al. 376 1992). However, we do not know exactly how this biomass interacts with soil organic matter, which has been 377 shaped by decades of dominance by ericoid shrubs; especially since the interaction between organic matter of 378 different biochemical properties often makes dynamics of decomposition of litter mixes non-additive (Gartner and 379 Cardon 2016). Heathland ecosystem is no exception: M. caerulea litter degrades significantly slower under C. 380 vulgaris than under itself (Certini et al. 2015). We therefore cannot rule out that the increase in available C coming
- 381 from the grass litter is inhibited by some compounds originating from long-term accumulated heather litter in the

382 organic topsoil, such as polyphenols (Kraus et al. 2003). The soil properties would then remain the same, and so 383 would most of the microbial community. Moreover, the literature that we use as comparison to understand the 384 effects of soil properties on microbial communities may represent a much more extreme case than grass 385 colonization in our field sites. For example, (Fierer et al. 2007) amended their soil samples with large amounts (up 386 to 800 g of equivalent C/m2/y) of a C source of high quality (sucrose), and (Zhou et al. 2021) with 120  $\mu$ g/g of 387 soil of glucose carbon. While litter input from a grass species differ from shrubs mostly by the relative abundance 388 of structural compounds, with more crystalline cellulose, and less aromatic compounds (Certini et al. 2015). This 389 improves litter quality, but to a much lower extent than glucose or sucrose, as polysaccharides need to be processed 390 by hydrolytic and/or oxidative enzymes before being incorporated for microbial metabolism, which is energy 391 demanding. Hence it appears logical to expect less drastic shifts in microbial communities in our study than in 392 (Fierer et al. 2007) and (Zhou et al. 2021). Finally, the absence of response of the ericoid genus Rhizoscyphus to 393 grass dominance could be caused by a slower necromass or propagule decomposition than other species (Lenaers 394 et al. 2018), which would artificially maintain its relative abundance figures in metabarcoding surveys (since it is 395 based on DNA which can be found in necromass and spores). Alternatively, that could be explained by the ability 396 of these species to persist in soil as saprotrophs after their host disappeared (Read et al. 2004).

397

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

399 Contrary to microbial community structure, C mineralization functions responded significantly to the increase in 400 grass cover. Even though the overall metabolic activity of the microbial communities remained the same, the 401 mineralization rate of tested C sources increased by an estimated factor of 2.7 across the full range of grass cover. 402 In other words, similar microbial assemblages showed almost three times more C mineralization potential under 403 pure grass than under pure heather, and this was not because microbes were just more metabolically active, as 404 shown by the results of the FDA assay. We cannot rule out, however, that the fungal OTUs favoured by grass 405 colonization (belonging to black yeasts, Dermataceae, Coniochaetaceae) contributed to this higher activity. In 406 fact, the number of read counts of S. podzolica was significantly correlated with the mineralization of D-cellobiose, 407  $\alpha$ -ketobutyric acid, 2-hydroxybenzoic acid, and  $\beta$ -methyl-D-glucoside; and the number of read counts of the OTU 408 belonging to *Devriesia sp.* was positively correlated with the mineralization rate of  $\beta$ -methyl-D-glucoside (data 409 not shown). The mineralization rates may nevertheless result from more complex inhibitory or synergistic

410 functional interactions.

411 Hence, the communities present in this heathland soil were adapted to these soil conditions, and responded to 412 change in litter input by raising their ability to mineralize C. In particular, grass dominance stimulated 413 mineralization rates of carbohydrates, amino acids, carboxylic acids, and polymers. The former three are 414 characteristic of root exudates (Griffiths et al. 1998), and the latter of plant biomass. We therefore speculate that 415 the increase in grass dominance stimulates C mineralization of root exudates, and to a lower extent, litter biomass; 416 but that litter input is in too low quantity relative to the soil organic matter pool to significantly affect its 417 composition, and therefore microbial community structure, before a decade or more. It also implies that the large-418 scale grass colonization we see in many North-Western European heathlands potentially results in significant 419 increases in C mineralization rates, which has negative consequences for this ecosystem's C sequestration. 420 However, this case study has been conducted in only one site, albeit a large one (287 500 m<sup>2</sup>), and different

421 responses may be observed in other heathlands because of site-to-site variability.

422	
423	Conclusion
424	Our results show that microbial community structure is only moderately affected by grass colonization, with the
425	exception of a limited number of abundant OTUs. However, this is not the case for community functioning: C
426	mineralization potential significantly increases, almost by a factor 3 overall. This was especially clear for the
427	substrates that were chemically related to root exudates, and to a lower extent the ones related to plant litter. This
428	led us to speculate that the change in dominant plant increased soil C mineralization by microbes through a change
429	in availability and nature of root exudates, and to a lesser extent litter input. Our study also points to some species
430	of both bacteria and fungi who were responsive to such a change in plant traits, thereby extending our knowledge
431	on the otherwise still mysterious ecology of Acidothermus bacteria and black yeast fungi. Taken altogether, these
432	results show that a change in dominant plant traits may induce drastic functional changes in microbial communities
433	despite having only a very minor effects on their diversity or community structure. Moreover, our study implies
434	that grass colonization observed in many heathlands in Northwestern Europe leads to higher C mineralization
435	rates, which has potentially negative impact on the C sequestration by this ecosystem.
436	
437	Acknowledgements
438	The authors also thank Regional Landscape Kempen and Maasland, Agentschap voor Natuur en Bos, and the
439	National Park Hoge Kempen for its collaboration, support, and providing access to the sampling area.
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441	Figure captions
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460 been inoculated from a soil sample. We used a linear model and Bonferroni correction to test the significance of

- the relationship between the two variables. Significant relationships are labelled with "p<0.05". The slope of the
- 462 relationship is given above the significance level.

463	References
464	
465	Adam G, Duncan H (2001) Development of a sensitive and rapid method for the measurement of total microbial
466	activity using fluorescein diacetate (FDA) in a range of soils. Soil Biol Biochem 33:943-951.
467	https://doi.org/10.1016/S0038-0717(00)00244-3
468	Aerts R, Bakker C, De Caluwe H (1992) Root turnover as determinant of the cycling of C, N, and P in a dry
469	heathland ecosystem. Biogeochemistry 15:175-190. https://doi.org/10.1007/BF00002935
470	Aerts R, Heil GW (1993) Heathlands. Patterns and processes in a changing environment. Geobotany 20
471	Bardgett RD, Mawdsley JL, Edwards S, et al (1999) Plant species and nitrogen effects on soil biological
472	properties of temperate upland grasslands. Funct Ecol 13:650-660. https://doi.org/10.1046/j.1365-
473	2435.1999.00362.x
474	Blagodatskaya E, Kuzyakov Y (2008) Mechanisms of real and apparent priming effects and their dependence on
475	soil microbial biomass and community structure: Critical review. Biol Fertil Soils 45:115-131.
476	https://doi.org/10.1007/s00374-008-0334-y
477	Bougoure DS, Parkin PI, Cairney JWG, et al (2007) Diversity of fungi in hair roots of Ericaceae varies along a
478	vegetation gradient. Mol Ecol 16:4624-4636. https://doi.org/10.1111/j.1365-294X.2007.03540.x
479	Brown JH, Whitham TG, Morgan Ernest SK, Gehring CA (2001) Complex species interactions and the
480	dynamics of ecological systems: Long-term experiments. Science (80-) 293:643-650.
481	https://doi.org/10.1126/science.293.5530.643
482	Caporaso JG, Kuczynski J, Stombaugh J, et al (2010) correspondence QIIME allows analysis of high-
483	throughput community sequencing data Intensity normalization improves color calling in SOLiD
484	sequencing. Nat Publ Gr 7:335-336. https://doi.org/10.1038/nmeth0510-335
485	Certini G, Vestgarden LS, Forte C, Tau Strand L (2015) Litter decomposition rate and soil organic matter quality
486	in a patchwork heathland of southern Norway. Soil 1:207-216. https://doi.org/10.5194/soil-1-207-2015
487	Chen S, Zhuang QQ, Chu XL, et al (2021) Transcriptomics of different tissues of blueberry and diversity
488	analysis of rhizosphere fungi under cadmium stress. BMC Plant Biol 21:1-19.
489	https://doi.org/10.1186/s12870-021-03125-z
490	Crous PW, Groenewald JZ (2011) Why everlastings don't last. Persoonia Mol Phylogeny Evol Fungi 26:70-84.
491	https://doi.org/10.3767/003158511X574532
492	Dollive S, Peterfreund GL, Sherrill-Mix S, et al (2012) A tool kit for quantifying eukaryotic rRNA gene
493	sequences from human microbiome samples. Genome Biol 13:R60. https://doi.org/10.1186/gb-2012-13-7-
494	r60
495	Fierer N (2017) Embracing the unknown: Disentangling the complexities of the soil microbiome. Nat Rev
496	Microbiol 15:579-590. https://doi.org/10.1038/nrmicro.2017.87
497	Fierer N, Bradford MA, Jackson RB (2007) Toward an ecological classification of soil bacteria. Ecology
498	88:1354–1364
499	Fierer N, Strickland MS, Liptzin D, et al (2009) Global patterns in belowground communities. Ecol Lett
500	12:1238-1249. https://doi.org/10.1111/j.1461-0248.2009.01360.x
501	Gartner TB, Cardon ZG (2016) Decomposition Dynamics in Mixed-Species Leaf Litter. Oikos 104:230-246
502	Genney DR, Alexander IJ, Hartley SE (2000) Exclusion of grass roots from soil organic layers by Calluna: the

- 503 role of ericoid mycorrhizas. J Exp Bot 51:1117–1125. https://doi.org/10.1093/jexbot/51.347.1117
- Griffiths BS, Ritz K, Ebblewhite N, Dobson G (1998) Soil microbial community structure: Effects of substrate
   loading rates. Soil Biol Biochem 31:145–153. https://doi.org/10.1016/S0038-0717(98)00117-5
- 506 Guckert JB, Carr GJ, Johnson TD, et al (1996) Community analysis by Biolog: curve integration for statistical
   507 analysis of activated sludge microbial habitats. J Microbiol Methods 27:183–197
- 508 Ihrmark K, Bödeker ITM, Cruz-Martinez K, et al (2012) New primers to amplify the fungal ITS2 region 509 evaluation by 454-sequencing of artificial and natural communities. FEMS Microbiol Ecol 82:666–677.
- 510 https://doi.org/10.1111/j.1574-6941.2012.01437.x
- Isbell F, Reich PB, Tilman D, et al (2013) Nutrient enrichment, biodiversity loss, and consequent declines in
  ecosystem productivity. Proc Natl Acad Sci U S A 110:11911–11916.
- 513 https://doi.org/10.1073/pnas.1310880110
- 514 Kardol P, Cregger MA, Campany CE, Classen AT (2010) Soil ecosystem functioning under climate change:....
  515 Ecology 91:767–781
- Kelly AE, Goulden ML (2008) Rapid shifts in plant distribution with recent climate change. Proc Natl Acad Sci
   U S A 105:11823–11826. https://doi.org/10.1073/pnas.0802891105
- 518 Kraus TEC, Dahlgren RA, Zasoski RJ (2003) Tannins in nutrient dynamics of forest ecosystems A review.
  519 Plant Soil 256:41–66. https://doi.org/10.1023/A:1026206511084
- Lenaers M, Reyns W, Czech J, et al (2018) Links between heathland fungal biomass mineralization ,
  melanization and hydrophobicity. Microb Ecol. https://doi.org/DOI: 10.1007/s00248-018-1167-3
- Massaccesi L, Bardgett RD, Agnelli A, et al (2015) Impact of plant species evenness, dominant species identity
   and spatial arrangement on the structure and functioning of soil microbial communities in a model
- 524 grassland. Oecologia 747–759. https://doi.org/10.1007/s00442-014-3135-z
- Mohagheghi A, Grohmann K, Himmel M (1986) Isolation and characterization of Acidothermus cellulolyticus
   gen. nov., sp. nov., a new genus of thermophilic, acidophilic, cellulolytic bacteria. Int J Syst Bacteriol

527 36:435–443. https://doi.org/10.1099/00207713-36-3-435

- Moreira GAM, Pires ECC, Barreto CC, Do Vale HMM (2020) Total fungi and yeast distribution in soils over
   native and modified vegetation in central brazil. Rev Bras Cienc do Solo 44:1–19.
- 530 https://doi.org/10.36783/18069657rbcs20200097
- Obase K, Cha JY, Lee JK (2009) Ectomycorrhizal fungal communities associated with Pinus thunbergii in the
   eastern coastal pine forests of Korea. 39–49. https://doi.org/10.1007/s00572-009-0262-1
- 533 Op De Beeck M, Ruytinx J, Smits MM, et al (2015) Belowground fungal communities in pioneer Scots pine
   534 stands growing on heavy metal polluted and non-polluted soils. Soil Biol Biochem 86:58–66
- 535 R core team (2019) R: A language and environment for statistical computing
- 536 Radujković D, van Diggelen R, Bobbink R, et al (2020) Initial soil community drives heathland fungal
- community trajectory over multiple years through altered plant-soil interactions. New Phytol 225:2140–
  2151. https://doi.org/10.1111/nph.16226
- 539 Ranjard L, Poly F, Lata JC, et al (2001) Characterization of Bacterial and Fungal Soil Communities by
- 540 Automated Ribosomal Intergenic Spacer Analysis Fingerprints: Biological and Methodological
- 541 Variability. Appl Environ Microbiol 67:4479–4487. https://doi.org/10.1128/AEM.67.10.4479-4487.2001
- 542 Read DJ, Leake JR, Perez-Moreno J, et al (2004) Mycorrhizal fungi as drivers of ecosystem processes in

- heathland and boreal forest biomes 1. Can J Bot 82:1243–1263. https://doi.org/10.1139/B04-123
- Read DJ, Perez-Moreno J (2003) Mycorrhizas and Nutrient Cycling in Ecosystems: A Journey towards
  Relevance? New Phytol 157:475–492
- Rutgers M, Wouterse M, Drost SM, et al (2016) Monitoring soil bacteria with community-level physiological
  profiles using Biolog TM ECO-plates in the Netherlands and Europe. Appl Soil Ecol 97:23–35.
  https://doi.org/10.1016/j.apsoil.2015.06.007
- Schnurer J, Rosswall T (1982) Fluorescein diacetate hydrolysis as a measure of total microbial activity in soil
   and litter. Appl Environ Microbiol 43:1256–1261. https://doi.org/0099-2240/82/061256-06\$02.00/0
- Seifert K, Nickerson N, Corlett M, et al (2004) Devriesia, a new hyphomycete genus to accommodate heat resistant, cladosporium-like fungi. Can J Bot 82:914–926. https://doi.org/10.1139/b04-070
- Seliskar DM, Gallagher JL, Burdick DM, Mutz LA (2002) The regulation of ecosystem functions by ecotypic
   variation in the dominant plant : a Spartina alterniflora salt-marsh case study. 1–11
- Tikhonov G, Opedal ØH, Abrego N, et al (2020) Joint species distribution modelling with the r-package Hmsc.
  Methods Ecol Evol 11:442–447. https://doi.org/10.1111/2041-210X.13345
- Tikhonov G, Opedal ØH, Abrego N, et al (2019a) Joint species distribution modelling with the R-package Hmsc
   Appendix S3. Comparing the performance of the Hmsc block Gibbs sampler with Hamiltonian Monte
   Carlo. Methods Ecol Evol
- Tikhonov G, Opedal ØH, Abrego N, et al (2019b) Joint species distribution modelling with the R-package Hmsc
   Appendix S2 . Details on how Hmsc was applied to the bird case study. Methods Ecol Evol 1–19
- 562 Untereiner WA, Straus NA, Malloch D (1995) A molecular-morphotaxonomic approach to the systematics of the
   563 Herpotrichiellaceae and allied black yeasts. Mycol Res 99:897–913. https://doi.org/10.1016/S0953 564 7562(09)80748-X
- Van Vuuren MMI, Berendse F (1993) Changes in soil organic matter and net nitrogen mineralization in
  heathland soils, after removal, addition or replacement of litter from Erica tetralix or Molinia caerulea.
  Biol Fertil Soils 15:268–274. https://doi.org/10.1007/BF00337211
- Wilkinson DP, Golding N, Guillera-Arroita G, et al (2021) Defining and evaluating predictions of joint species
   distribution models. Methods Ecol Evol 12:394–404. https://doi.org/10.1111/2041-210X.13518
- 570 Yurkov AM (2018) Yeasts of the soil obscure but precious. Yeast 35:369–378.

571 https://doi.org/10.1002/yea.3310

- Zhou J, Wen Y, Shi L, et al (2021) Strong priming of soil organic matter induced by frequent input of labile
  carbon. Soil Biol Biochem 152:. https://doi.org/10.1016/j.soilbio.2020.108069
- 574 575

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