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1 Substrate affinities of wood decay fungi are foremost structured by wood properties not  
2 climate

3

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20 **Abstract**

21 Wood decomposing fungi differ in their substrate affinities, but to what extent factors like  
22 wood properties influence host specialization, compared to climate, is largely unknown. In  
23 this study, we analysed British field observations of 61 common wood decay species  
24 associated with 41 tree and shrub genera. While white rot fungi ranged from low- to high-  
25 substrate affinity, brown rot fungi were exclusively mid- to high- affinity. White rot fungi  
26 associated with dead fallen wood demonstrated the least substrate affinity. The composition of  
27 wood decomposer fungi was mostly structured by substrate properties, sorted between  
28 angiosperms and conifers. Any relationships with temporal and regional climate variability  
29 were of far less significance, but did predict community-based and substrate-usage host shifts,  
30 especially for fungi on fallen deadwood. Our results demonstrate that substrate shifts by  
31 wood-decay fungi will depend primarily upon their degree of affinity to, and the distribution  
32 of, related woody genera, followed less at regional levels by climate impacts.

33

34 **Keywords:** affinity, climate, decay, reproductive traits, specialization, substrate usage, wood

35 **Introduction**

36 Assemblage patterns of fungal species are strongly linked to climate across large spatial scales  
37 (Andrew et al. 2018a), impacting distributions and ranges (Davis & Shaw 2001; Kelly and  
38 Goulden 2008, Wollan et al. 2008, Diez et al. 2020). Simultaneously, rapid changes in  
39 atmospheric greenhouse gasses and aerosols have drastically modified the global climate,  
40 perhaps most clearly manifested in mean annual temperatures. A prominent warming can be  
41 observed even in the past half century, in direct correlation with increasing fossil fuel  
42 emissions (Pachauri et al. 2014). Driving ecosystem dynamics, overall climate change can  
43 influence the biology of fungi (e.g., reproduction and phenology; Boddy et al. 2014), with  
44 consequences for species interactions and substrate associations (Allen et al. 2010; Gange et  
45 al. 2011).

46

47 The timing of fruit body production is different to that of half a century ago for many species  
48 (Boddy et al. 2014; Andrew et al. 2018b), and the shifts are direct consequences of climate  
49 (Andrew et al. 2018c). For example, in southern England the average fruiting period of 315  
50 fungal species has more than doubled from  $33.2 \pm 1.6$  d to  $74.8 \pm 7.6$  d in the timespan of  
51 1950 to 2005 (Gange et al. 2007). Kauserud et al. (2008) likewise demonstrated a similar  
52 pattern in Norway, when investigating the phenology of 83 agaricoid species. By utilizing  
53 large fungarium datasets of records from 1940 to 2006, they detected a delay in fungal  
54 fruiting by  $13.3 \pm 1.2$  d across all species. While it is unquestionable that climate change is a  
55 cause of fungal phenological shifts (Andrew et al. 2018c), less clear is how the degree of  
56 affinity of fungi to their substrates might interact with climate-based impacts (Boddy et al.  
57 2014).

58

59 Observations, especially in Great Britain, have suggested that the substrate affinity of wood  
60 decay fungi can change with time and is related to climate (Renvall 1995; Gange et al. 2011;  
61 Boddy et al. 2014; Bien & Damm 2020), but this phenomenon has not yet been linked to the  
62 degree of host specificity. Early criticism highlighted the need for systematic study  
63 (Heilmann-Clausen & Læssøe 2012); when more rigorous analyses were implemented, the  
64 results of wood decay host shifts remained, as did the more ecologically relevant questions  
65 regarding the causes, outside of climate, for host shifts (Gange et al. 2012) . However, active  
66 debate remains among scientists regarding this subject, suggesting further research is needed.  
67 More recent studies have implicated the importance of substrate characteristics to the  
68 diversity and composition of wood decay fungi (Purahong et al. 2018a; 2018b). Taken  
69 together, this emphasises the need to understand fungal dynamics in relation to substrate  
70 affinity(-ies), and alongside the influence of temporal change and climate.

71

72 Wood-decomposing fungi are especially suitable for investigating relationships between  
73 substrate affinity and climate for multiple reasons: wood rot fungi include many species that  
74 form macroscopic fruit bodies (Schmidt 2006), and which are easy to identify, ensuring  
75 taxonomic credibility. Many species also fruit frequently over time, providing multitudes of  
76 observer recordings, by which sampling bias is negated. Other wood decay fruit bodies may  
77 fruit less frequently but are robust and long-lived, another reason they are easier to observe  
78 and, hence, to measure reliably and analyse across time. Very recently, Runnel et al. (2021)  
79 similarly advocated their usage in conservation biology. As wood-decomposing fungi fruit  
80 from the woody substrate they rot, co-recorded substrate metadata are more readily available  
81 for wood decay fungi than others, e.g., soil-borne fruit bodies.

82

83 Ecologically, the brown- and white rot decay systems make them important decomposers and  
84 nutrient-cyclers in forest ecosystems. For example, in Europe alone, at least 393 polypore  
85 species have been recorded, with 99 of them (25%) classified as causing brown rot (Ryvarden  
86 et al. 2014). Recent genome analyses have shown that wood decay mechanisms are more  
87 diverse than previously thought and that a categorical classification of decay types into white  
88 rot and brown rot must be refined (Riley et al. 2014; Floudas et al. 2015). However,  
89 quantitative physiological or genomic information about decay types is currently not  
90 available. Wood decay roles are dual, and thus ecologically important, as they can sometimes  
91 also live as endophytes or pathogens in standing trees (e.g., heart, stem and root rots) as well  
92 as in decaying fallen logs and branches and stumps (Song et al. 2016).

93  
94 Dynamics of wood decay fungi indicate that priority effects (Ottosson et al. 2014; Hiscox et  
95 al. 2015; 2016; Leopold et al. 2017; Norberg et al. 2019) as well as competitive interactions  
96 (van der Wal et al. 2016; Hiscox et al. 2018) shape communities and successional change.  
97 Wood properties are also important, for example wood chemistry (Fukasawa 2021; Lunde et  
98 al. 2022), decay stage (Holec et al. 2020) and stem diameter (Brazee et al. 2012). Less clear at  
99 this point, however, is the degree that fungal affinity, i.e., specialization to specific substrate  
100 taxa, might influence and inter-relate to climate effects.

101

102 In this study, we questioned whether any temporal and climate-related trends may be  
103 responsible for the substrate affinities of British wood decomposer fungi, from individual  
104 species' affinities to overall compositional trends. We utilized multi-source wood decay  
105 fungal fruit body records, for the past four decades, from across the mainland parts of the  
106 United Kingdom (UK; England, Wales and Scotland). By focusing on the UK, we could  
107 follow-up on research gaps in earlier substrate affinity studies (i.e., Gange et al. 2011; 2012,

108 Boddy et al. 2014), and within the same temporal period that climate conditions have changed  
109 in the UK, i.e., largely overlapping with the Gange et al. (2011; 2012) studies. Given that the  
110 UK offered strict differences in climate between the more oceanic western side, in contrast to  
111 the more continental eastern area, in this way we focused on the effect of climate on  
112 decomposer fungi's host associations and traits.

113

114 The research was based on three objectives: (1) To establish the degree of substrate affinity  
115 for the most commonly and consistently recorded wood decay species, in relation to  
116 ecological characteristics of the fungal taxa as well as their substrate genera. Fungal species  
117 were delineated by their reproductive traits and decay characteristics. Substrates were  
118 characterised as wood from either angiosperms or conifers. (2) Investigate the effects of  
119 climate on fungal composition between substrates, over a 40-y temporal scale. In this case, we  
120 were interested in both compositional changes related to temporal shifts, as well as  
121 differences between the east-west climate regions of the UK. (3) Determine the extent,  
122 through modelling, that fungal characteristics influenced their substrate affinities, in relation  
123 to climate and temporal change.

124

## 125 **Materials and Methods**

### 126 *Data filtering and processing*

127 Fungal multi-source data (museum specimens, citizen science and scientific observations)  
128 were extracted from the Fungal Records Database of Britain and Ireland (FRDBI:  
129 [www.fieldmycology.net](http://www.fieldmycology.net), [www.frdbi.info](http://www.frdbi.info)) for all recorded fruiting events in the mainland UK  
130 countries of England, Scotland and Wales. Accompanying annotations on the associated  
131 substrate genera, the exact locations, and the year of observation were required. Records were  
132 taxonomically filtered to saprotrophic and pathogenic taxa found on woody substrates

133 (including taxa in the Polyporales, Hymenochaetales, Russulales, Thelephorales and including  
134 stereoid fungi) – this removed common species such as *Fistulina hepatica*, a well-known  
135 specialist of oak (*Quercus*) and sweet chestnut (*Castanea sativa*) in Great Britain. Fungal  
136 species with less than 100 total observations were removed, as were substrate genera with less  
137 than 10 fruit body records associated with them. Limiting the taxonomic scope was suitable  
138 for the goals of the project, and simultaneously reduced potential sampling bias by retaining  
139 only those taxa that were readily identifiable and with distributional prevalence (Cao and  
140 Larsen 2001). Temporal limitation of 1970 to 2010 captured the latest trends in temperature  
141 increase (Pachauri et al. 2014) while ensuring sufficient record amounts for analyses. The  
142 final dataset contained 53,094 records of 61 fungal species and 41 associated woody substrate  
143 genera.

144

145 A variety of climate variables were investigated for potential trends, entirely and across two  
146 equal time periods (1970-1990 and 1991-2010). Data were obtained from the Met Office in  
147 2016, and overlain with a Watsonian vice-county map (a geographical division of the British  
148 Isles for the purpose of scientific data collection). The geography of the UK contributed to a  
149 longitudinal climate gradient that we used to divide, as appropriate for analyses, the climate  
150 and fungal data into two (multivariate) and eight (regression) zones (Supplemental Figure 1).  
151 Temperature (minimum, maximum and mean), and precipitation (total) were investigated at  
152 both annual and seasonal (meteorological spring, summer, autumn, and winter) aggregations.  
153 Mean annual temperature between the years and regions ranged from 8.07 to 9.65 °C, while  
154 maximum annual temperatures ranged from 11.34 to 13.43 °C, and minimum annual  
155 temperatures from 4.77 to 5.75 °C. Annual rainfall ranged from 746.11 to 1501.95 mm.

156



157 Fungal traits were extracted from Breitenbach (1986) and Ryvarden et al. (2014) for a variety  
158 of reproductive and ecological characteristics: the general rot type (white or brown); the  
159 fruiting frequency (annual or perennial, the latter defined as lasting two or more years); the  
160 substrate stage when fruiting typically occurred (lying deadwood substrate (a combination  
161 dead logs, stumps, and branches), or standing substrate (which could be alive or dead)) and  
162 the average spore volume. While we were aware that specimen-based trait data, capturing  
163 intraspecific variation, would be preferential, unfortunately such data were not available.  
164 Traits were filtered to reduce multi-collinearity, with Pearson correlation used for continuous  
165 variables, eta for continuous and categorical variables, and Cramer's V (phi) for two  
166 categorical variables. Only non-collinear traits were included in the analyses. Substrate genera  
167 were categorised as either angiosperms or conifers.

168

169 Genetic distance data contained the ribosomal DNA (rDNA) large subunit (LSU) 28S  
170 sequences from GenBank (accessed October 26<sup>th</sup> 2015) for the study species. Sequences were  
171 available for 39 of the 61 fungal species. The LSU region was optimal to use in this study due  
172 to its limited mutation rate, making it possible to align the sequences across varying  
173 taxonomic orders in the Basidiomycota. One representative LSU sequence was selected for  
174 each of the 39 taxa. The sequences were aligned and pairwise genetic distances among all  
175 taxa were calculated using MEGA5 (Tamura et al. 2011).

176

### 177 *Substrate specialization and substrate preferences*

178 To visualize the ranges of associations for substrates and fungal species and the potential  
179 substrate affinities of the fungal species, a heat-map (package gplots in R) was generated  
180 using the proportion of observations that were recorded for a given fungal species (61) across  
181 the different woody substrate genera (41). Hierarchical clustering sorted the fungal species

182 from substrate generalist to specific, generating a sequence of substrate affinity, and the  
183 woody substrate genera from commonly to more rarely hosting the fruiting fungi. Trait and  
184 taxonomy attributes were added to the axes, selecting those as found important in the  
185 statistical models (described below).

186

187 The 28S LSU genetic distance data were correlated with a Simpson index of variation in  
188 substrate distributions among the fungal species, calculated as the difference in the Simpson  
189 1-D index values between the species, to understand if genetic similarity related to substrate  
190 usage. The degree that genetic distance and the substrate-diversity distances matched was  
191 estimated with a Pearson correlation across all interspecific comparisons. We investigated  
192 how linear transformations of one matrix matched the other matrix with Procrustes analyses  
193 (package *vegan* in R; Oksanen et al. 2016). A large residual difference of the matrices would  
194 indicate a poor match between the genetic and Simpson index differences. The Procrustes  
195 analyses results were further verified through a permutation test (999 permutations) with a  
196 null-hypothesis that the Simpson index was independently distributed among the fungal  
197 species. The diversity distance matrix was recalculated for each species from permuted values  
198 and compared to the original genetic distances. If the observed correlation and Procrustes  
199 match was high, the permuted absolute correlation was assumed lower, and the residuals of  
200 the Procrustes larger. From these results, the correlation in genetic similarity and substrate  
201 usage was determined.

202

### 203 *Compositional changes by substrate, time, region and climate*

204 Each woody substrate genus was considered to host a fungal assembly, measured by the  
205 association of the fungal species with that genus. From the data we implemented a post hoc  
206 experimental design with replications by region (eastern or western UK), decade (1970s,

207 1980s, 1990s, 2000s), and the associated climate (mean annual temperature and precipitation).  
208 Little difference in fungal species composition on substrate genera between regions or  
209 decades would suggest few spatiotemporal changes in substrate specificities, hence little  
210 substrate specificity based on the prevailing climatic conditions. Greater compositional  
211 differences would, on the other hand, suggest substrate specificity changes for fungi on  
212 certain substrate genera.

213

214 By investigating the composition of decay fungi associated with woody substrate genera, we  
215 were able to examine the potential of climate-related shifts of fungi on their substrates, and in  
216 relation to spatiotemporal variance. We utilized canonical-correlation analysis (CCA)  
217 combined with variance partitioning. The compositional variability was dissected by the  
218 extent the available variables explained it, either individually or when confounded (Borcard et  
219 al. 2011). Forward selection of the 23 original Met Office variables identified the four  
220 variables that were included in the final analyses (mean annual temperature, decadal time  
221 period, region, and substrate genus). Four decadal time periods (1970's, 1980's, 1990's, and  
222 2000's) and two climate-driven regions (eastern and western UK; Supplemental Figure 1)  
223 were selected. To reduce any possible effects of different sampling efforts between regions,  
224 the data were transformed to presence-absence of fungal species per substrate, region and  
225 time period. Two substrate genera (*Euonymus* and *Ribes*) and the one fungal species that  
226 associated most often with them, *Phylloporia ribis*, were outliers and were subsequently  
227 removed from the analyses. The variance attributable to spatiotemporal effects was  
228 constrained in the CCA, so as to focus upon the dominating impacts of climate and substrate  
229 genera.

230

231 *Influence of fungal traits and climate on substrate usage*

232 A fourth-corner and RLQ analysis explained how fungal traits on a compositional level  
233 related to the fungal habitat, i.e., plot-level variables of environment and substrate (Dray et al.  
234 2014). The variables as previously selected in the CCA and variance partitioning analyses  
235 were included. A correspondence analysis was run for the continuous variables. A Hill-Smith  
236 analysis was utilized on matrices with both categorical and continuous variables (Brown et al.  
237 2014; Dray et al. 2014). Randomization procedures implemented 49,000 permutations and the  
238 *P* values were adjusted for multiple testing with the false discovery rate procedure (Dray et al.  
239 2014). From these analyses, directional correlations were generated based on the trait trends  
240 in the fungal species composition by those of the habitat variables.

241

242 For temporal analyses related to trait characteristics between fungal species and the woody  
243 substrate genera, Bayesian inference applying Integrated Nested Laplace Approximation  
244 (INLA) was utilized (Rue et al. 2009). From the total fungal occurrences on substrates, we  
245 could assess whether there was a temporal change induced by the trait of a fungal species. The  
246 statistical specifications followed that outlined for trait-specific multivariate regressions  
247 (Jamil et al. 2012) and the community assembly by traits (Brown et al. 2014), using a negative  
248 binomial distribution. In this case, we investigated if there was an effect of traits on fungal  
249 occurrences between two equal time periods, 1970-1990 and 1991-2010. The shared effects  
250 among species of the same trait were analysed, while allowing individualistic species  
251 responses by time period, with paired substrates across the time periods. The analyses were  
252 conducted in R version 3.2.2 using the following packages: vegan, MASS (Ripley 2011), ade4  
253 (Dray and DuFour 2007), and R-INLA (Rue et al. 2009).

254

## 255 **Results**

### 256 *Substrate specializations*

257 In our UK study sample, the decomposer fungi that demonstrated the greatest substrate  
258 specialisation ( $\geq 95\%$  of occurrences with one substrate and any other genera  $\leq 1\%$ ) were  
259 associated with birch (*Betula*) and oak (Figure 1; Supplemental Table 1). Those fungal  
260 species with greatest affinity for birch were *Fomitopsis betulina* (= *Piptoporus betulinus*) and  
261 *Inonotus obliquus*, although each were recorded on more substrates than birch, with 8 and 5  
262 total tree genera, respectively. Those species mostly specializing with oak were *Piptoporus*  
263 *quercinus*, *Daedalea quercina*, and *Pseudoinonotus dryadeus*, which were recorded with 1, 12  
264 and 9 total substrate genera, respectively.

265

266 Further species that were recorded on few substrate genera, but which demonstrated less  
267 specificity, were (Figure 1; Supplemental Table 1): *Hericium erinaceus* with 4 recorded  
268 substrate genera, but 92% of occurrences with beech (*Fagus*); *Hericium cirrhatum* with 7  
269 recorded substrate genera, but 83% of occurrences with beech; and *Corioloopsis gallica* which  
270 was associated with 5 recorded substrate genera, although 50% of occurrences were with  
271 beech and 39% with ash (*Fraxinus*). The species with overall less substrate specialisation and  
272 also the greatest amount of substrate species were *Bjerkandera adusta* (35 recorded total  
273 substrate genera; mostly with beech (49%), birch (14%) and oak (11%)) and *Trametes*  
274 *versicolor* (34 recorded substrate genera; predominantly associated with beech (28%), birch  
275 (22%) and oak (17%)).

276

277 While fungal species with epithets signifying their substrate affinity could be often associated  
278 with those trees – for example, *Lenzites betulina* (13 total recorded substrate genera) had a  
279 67% affinity to birch, some were more misleading in terms of associations in the UK (but  
280 could be based on other regions, e.g., the prevalence of *Picea abies* in Scandinavia, or due to  
281 tree decline, e.g., the loss of *Ulmus* in the UK) – such as *Trichaptum abietinum* (14 total

282 genera), that had 81% of occurrences with pine and 0.4% with fir (*Abies*), and *Rigidoporus*  
283 *ulmarius* (14 total genera) with only 40% of occurrences with elm (*Ulmus*) and 11% with  
284 maple (*Acer*)).

285

286 The Polyporales dominated the wood decay fungi in this UK study, and were found across all  
287 degrees of substrate affinity. In comparison, fewer wood decay species were found in the  
288 Hymenochaetales and Russulales, which demonstrated mid-level to specialist substrate  
289 affinities. This corresponded with a greater amount of white rot fungi with low substrate  
290 affinity, while brown rot fungi had exclusively mid- to high affinity levels with specific  
291 woody substrate genera (Figure 1; Supplemental Table 1). Fungi of both rot types were  
292 associated with both deciduous angiosperms and conifers, especially the Fagales (beech, oak,  
293 birch and hornbeam (*Carpinus*)) and the Pinaceae (pine (*Pinus*), spruce (*Picea*), larch (*Larix*),  
294 and fir (*Abies*)). Fungal species with greater affinity to pine were often also associated with  
295 spruce and larch at levels equalling and greater than to those of the Fagales taxa, i.e., the  
296 suggestion for conifer preference. The fungal species with lowest substrate affinities were  
297 always annually fruiting white rot fungi in association with dead, downed wood. For the  
298 woody genera, the horticultural and hedgerow shrub-like taxa (e.g., *Ligustrum*) and trees with  
299 peeling to flaking bark (e.g., *Platanus*) were the rarest substrates for the wood decay fungi.

300

301 There was a positive correlation between genetic distances among fungal species and their  
302 host distribution; hence, phylogenetically related species shared more of the same host taxa ( $p$   
303  $\leq 0.008$ ). The significance of the Procrustes analysis ( $p \leq 0.005$ ) suggested a very low  
304 probability of the correlation being the result of random effects.

305

306 *Compositional changes on substrates by time, region and climate*

307 Substrate genera accounted for a major part of the variation in fungal assemblies (31.5%;  
308 Table 1). In the CCA, composition sorted largely by wood type of the substrates along the  
309 first axis (angiosperm versus conifer wood; Figure 2). The fungal composition also aligned  
310 with substrate taxonomy and host growth form, i.e., those communities associated with trees  
311 in the Fagales were more similar to one another, and in terms of the angiosperm taxa, they  
312 arranged most distally from the horticultural and shrub-like taxa. The second CCA axis  
313 gradient differentiated fungal compositions within the angiosperm tree substrates, and less  
314 that of the conifer substrates. Hence, while the second axis primarily captured variation in  
315 fungal assemblies across deciduous substrates, the first axis captured that between the types of  
316 woody genera, as well as within the conifer group (Figure 2). For example, medium- to high  
317 affinity brown rot fungi were mostly associated with the conifer and Fagales- associated  
318 communities.

319

320 Compared to the host genera, climate variability accounted for a very small part of the  
321 variation in fungal assemblies (Table 1). Among the assessed climate variables, mean annual  
322 temperature was the most important, still, it only accounted for 0.4% of the variation in fungal  
323 composition (Table 1). Temperature functioned simultaneously (i.e., non-linearly) between  
324 the two gradients, as illustrated by the isolines in the CCA plot, encompassing the fungal  
325 communities of both wood type groupings in annual means of 8.6 to 9.4 °C, with some  
326 suggestion for climate trends within-groupings of conifer and Fagales associations (Figure 2).

327

328 The compositional variance by decade (1970's to 2000's) and region (eastern or western UK)  
329 constituted 1.7% of the overall variability, which was conditioned from the host and climate-  
330 focused effects on composition. In fact, compositional variance that had been related to the  
331 decade (1970's to 2000's) and region (eastern or western UK) were so minimal that the eight

332 decade-region combinations were as effective to display as averages (Figure 2), but see  
333 Supplemental Figure 2 for the non-averaged version.

334

335 *Influence of fungal traits and climate on substrate usage*

336 The fourth-corner and RLQ analyses were used to test associations between fungal traits and  
337 habitat characteristics based on the compositional trends. There was a positive association  
338 between fungal spore volume and species having angiosperm trees as substrate (Table 2).

339 There were also associations between rot type and substrate, with white rot taxa preferentially  
340 appearing on angiosperm substrates ( $p \leq 0.00$ ) and brown rot taxa on coniferous substrates. A  
341 trend for relatively more occurrences on downed deadwood (as opposed to standing) occurred  
342 across the four decades (adj.  $p \leq 0.05$ ).

343

344 Fungal species' trait-mediated changes in abundance between earlier (1970 – 1990) and later  
345 (1991 – 2010) decades were impacted by the substrate stage (lying deadwood versus standing  
346 wood; Supplemental Table 2 & Supplemental Table 3). For fungi of both substrate types and  
347 after accounting for recorder effort, models indicated an increase in species from the first to  
348 the second time period (Figure 3). Importantly, the trends were parallel for the two substrate  
349 types, suggesting no bias and equal increases in abundances. In contrast, when considered in  
350 terms of relative percent increase, the mean expected substrate value was disproportionately  
351 higher with time, predicting more fungi on fallen deadwood compared to standing substrate.



352 **Discussion**

353 Our three objectives related to quantifying the degree of substrate affinity by wood decay  
354 fungi (including taxa in the Polyporales, Hymenochaetales, Russulales, and Thelephorales).  
355 We questioned how this might be influential to climate- and temporal- related change of fungi  
356 on their substrates. The records of wood decay fungi in this study originated from multiple  
357 sources (mainly citizen science records and research studies) of fruit body records limited to  
358 the UK for the 1970 to 2000's decades.

359

360 We (1) found that specialization occurred for fungi across rot types, while generalization was  
361 restricted to white rot fungi. Fungi were mostly restricted to fallen deadwood if their substrate  
362 affinity was low, while the frequency of fungi on standing trees increased with higher  
363 substrate affinities. Fungal species more often exhibited substrate affinity classifiable by  
364 characteristics of angiosperm or conifers. There were (2) also discernible compositional  
365 patterns, where woody substrate genera arranged primarily along gradients clustering conifers  
366 and angiosperm substrate characteristics, with temperature gradients only in minor part  
367 interacting within these more dominant compositional forces. Among the trait-related trends,  
368 a positive relationship between fungal species composition by rot type and substrate type  
369 (angiosperm or conifer) was detected, again evidencing the importance of substrate  
370 characteristics that relate to wood properties in structuring composition. Across the four  
371 decades, the fungal species composition shifted on downed deadwood due to a positive  
372 association with it, in contrast to the lack of any correlation with species composition on  
373 standing wood. This corroborated our final finding (3) where models indicated more fungi on  
374 downed deadwood than standing substrate across time.

375

376 Our results suggest that substrate shifts by wood decay fungi may be mediated to some extent  
377 by climate change (as defined in this case by broad geographical trends of eastern and western  
378 UK), but are primarily determined by woody hosts, related to their general wood properties,  
379 for example, angiosperms or conifers. Substrate shifts by fungi on downed deadwood will be  
380 the most challenging to discern, as they are primarily generalists in affinity (Supplemental  
381 Table 1), and can also be infrequent and patchily distributed within wood, based on molecular  
382 evidence (Baldrian et al. 2016). That fungi were modelled to have increased (relatively) more  
383 on deadwood than standing substrate in the latter half of the time period added further  
384 challenge in discerning causes for trends. We could only speculate whether management  
385 programmes towards coarse woody debris retainment contributed to this trend.

386

387 We were originally most interested in the potential for substrate shifts by taxa with higher  
388 substrate affinity within the UK, for example as has been found for *Auricularia auricula-*  
389 *judae* (Gange et al. 2011; 2012; Boddy et al. 2014). One interpretation is that our results  
390 demonstrate the potential for considerable plasticity in even the most host-specific taxa  
391 (Figure 1), as no species was singularly observed on one host genus. This could make sense in  
392 terms of the structure and chemical composition of wood primarily differentiating between  
393 angiosperm and conifers (e.g., Miller 1999) than nuances between species. It does also  
394 explain the results we report here, in that species' substrate affinity and compositional  
395 patterns related foremost to wood properties, and only very limited extent to climate (Table 1,  
396 Figure 2). Recently Leonhardt et al. (2018) demonstrated, somewhat similarly, distinction in  
397 wood decay fungi by tree leaf type (deciduous or evergreen), and when combined with further  
398 results, pinpointed the ligninolytic manganese peroxidase enzymatic pathway as influential  
399 for explaining fungal substrate differences. Even more recently Runnel et al. (2021) likewise  
400 dissected differences in polypores related to Estonian forest biodiversity.

401  
402 Potential biases must always be borne in mind when trying to interpret these data. For  
403 example, are rare substrate reports for fungal species clearly specialising on certain tree taxa  
404 genuine or misidentifications? Instances have been found where people tended towards  
405 reporting the more unusual sightings of fungal fruiting, for example, when out of season or on  
406 an unusual host (Halme et al. 2016). However, as in other cases, is there observer bias (e.g.,  
407 Heilmann-Clausen et al. 2019), or even unknown effects of cryptic speciation (e.g., Runnel et  
408 al. 2021) impacting the results? Our analytical approach cannot answer these questions. In the  
409 future, more certainty could be obtained where there is access to vouchers connected to the  
410 records for molecular investigations alongside morphological comparisons for species  
411 assessment (e.g., Andrew et al. 2018d).

412  
413 Heilmann-Clausen et al. (2016) also, through citizen science data, investigated more than  
414 1000 fungal species and 91 woody substrate genera in Denmark. They showed that substrate  
415 tree size, wood pH, and the number of species within each substrate genus to positively  
416 influence fungal wood decay species richness. This concurs with our findings that properties  
417 of the wood, and decay type, are the bases for differences in the substrate affinity of fungi.  
418 We again suggest that experiments on the growth of fungi in different wood types, both alone  
419 and competing with others, and environmental cues for fruiting (Moore et al. 2008), are  
420 needed to elucidate this further. Substrate affinity of fungal species may also vary depending  
421 on environment (Boddy et al. 2014), and our results demonstrated that mean annual  
422 temperature was more influential than annual precipitation in structuring communities.

423  
424 Future studies in different regions, time periods or other spatiotemporal scales, would benefit  
425 from focusing separately on angiosperm or conifer wood, so as to more clearly discern

426 impacts of climate on fungal decay, which are clearly lesser to that of biotic associations  
427 (Figure 2). Our results, and those of Heilmann-Clausen et al. (2016) match well with those  
428 found using molecular methodologies (Baldrian et al. 2016; Leonhardt et al. 2018; Purahong  
429 et al. 2018a; 2018b), indicating the value of both approaches.

430

### 431 **Conclusion**

432 Substrate affinity, in the strictest sense, is less frequent than generalization for wood decay  
433 species. Greater affinity is more likely to occur with fungi associated with standing wood than  
434 for those on downed wood, and the latter are more often generalist white rot fungi. Substrate  
435 shifts are likely to be exacerbated in conditions that change the presence of wood decay taxa,  
436 properties related to wood types (angiosperm or conifer) and substrate location (standing  
437 versus downed dead), and, to a very limited extent, climate when defined by mean annual  
438 temperature. It would be extremely beneficial to continue to characterize fungal affinities  
439 across other regions than those discussed here, to determine the actual mechanisms related to  
440 fungal decay affinity (which would better distinguish degree of affinity), and to assess any  
441 impacts that new biotic associations, resultant from substrate shifts, may have on extant  
442 fungal communities and their dynamics.

443

### 444 **Conflicts of interest**

445 All authors affirm that no competing interests exist with respect to this manuscript.

446

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453

454 **Authors' contributions (listed alphabetically by last name)**

455 CJA, LB, ACG, EH, KH, HK, FR conceived of and participated in the design of the study.

456 CJA, EH, HK, FR conducted the statistical analyses.

457 CJA, LB, ACG, EH, KH, HK, FR drafted the manuscript.

458 All authors gave final approval for publication.

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665 **Figure and Table captions**

666

667 Figure 1: The proportional occurrences ( $\log_{10}$ ) of fungal species on woody substrate genera  
668 are color-coded from none (lightest beige) to approaching complete specificity (darkest  
669 maroon). Fungal species are ordered from the most general to the most specific substrate  
670 affinities to the woody genera listed. The substrate genera are ordered from high (left) to low  
671 (right) numbers of fungal species. The dendrogram for the fungal species is accompanied by  
672 shading of general rot type (brown versus white), and for the substrate genera by their wood  
673 types (broad-leaved angiosperm versus needle-leaved conifer).

674

675 Figure 2: The compositional similarity between the fungal species recorded in association with  
676 woody substrate genera. Each abbreviation represents the fungal composition associated with  
677 a substrate genus (the three letters), averaged across the four decades and two regions for  
678 clarity. The woody substrate genera are shaded by their wood type, broad-leaved angiosperm  
679 (lighter green) or needle-leaved conifer (darker green), which differentiate the fungal  
680 compositions. The greyscale numbers represent the fungal species' scores relative to the  
681 substrate communities, i.e., species influencing neighbouring communities. The fungal  
682 species are ordered (1 to 61) in increasing substrate affinity. The grey shadings reflect the  
683 species' general rot-type (brown versus white). Mean annual temperature, which explains less  
684 compositional variability than does the substrate genus, is non-linearly associated with  
685 compositional variance and is represented by the orange isolines. See Figure 1 for the full  
686 substrate genera names as well as the fungal species names in order of substrate affinities. See  
687 Table 1 for the extent of variance explained by habitat variables. Figure 2 is the matching, the  
688 more detailed plot version that includes fungal communities by decade and region.

689

690 Table 1: Variance partitioning of climate (mean annual temperature) and substrate (woody  
691 genera) from the CCA analysis. The model is spatiotemporally conditioned by four decades  
692 (1970's to 2000's) and region (eastern or western UK), which contributed <2% variance.

693

694 Table 2: The statistically significant correlations in the fungal traits related to the  
695 compositional variability are shown with respect to the associated habitat characteristics  
696 (woody substrate and environmental properties). The directions of the relationships were  
697 always positive, as noted by the plus signs, which designate the degree of statistical  
698 significance. The adjusted p-value is also reported in the parentheses (adj.  $p \leq$  value).

699

700 Figure 3: Modelled absolute (A) and relative change (B) in fungal observations by time (1970  
701 – 1990 versus 1991 – 2010) depends upon the substrate type, i.e. downed deadwood versus  
702 standing wood. For both substrate types, the total number of fungal observations are predicted  
703 to increase with time, but the relationship between the use of host stages remains relatively  
704 stable, i.e., parallel responses. In contrast, the relative percent change between the substrates  
705 is proportionally different, demonstrating that downed deadwood is favoured with time.