

1 Revealing hidden insect-fungus interactions; moderately specialized, modular  
2 and anti-nested detritivore networks

3 *Running head: Insect-fungus interaction networks*

4 Rannveig M. Jacobsen\*<sup>a,b</sup>, Anne Sverdrup-Thygeson<sup>a</sup>, Håvard Kauserud<sup>c</sup>, Tone Birkemoe<sup>a</sup>

5 <sup>a</sup> Faculty of Environmental Sciences and Natural Resource Management, Norwegian

6 University of Life Sciences, Høgskoleveien 12, 1433 Ås, Norway

7 <sup>b</sup> The Norwegian Institute for Nature Research (NINA), Gaustadalléen 21, 0349 Oslo,

8 Norway

9 <sup>c</sup> Section for Genetics and Evolutionary Biology (EVOGENE), University of Oslo,

10 Blindernveien 31, 0316 Oslo, Norway

11 \* Corresponding author: [rannveig.jacobsen@nina.no](mailto:rannveig.jacobsen@nina.no)

12

13

## 14 Abstract

15 Ecological networks are composed of interacting communities that influence ecosystem  
16 structure and function. Fungi are the driving force for ecosystem processes such as  
17 decomposition and carbon sequestration in terrestrial habitats, and are strongly influenced by  
18 interactions with invertebrates. Yet, interactions in detritivore communities have rarely been  
19 considered from a network perspective. In the present study, we analyse the interaction  
20 networks between three functional guilds of fungi and insects sampled from dead wood.  
21 Using DNA metabarcoding to identify fungi, we reveal a diversity of interactions differing in  
22 specificity in the detritivore networks, involving three guilds of fungi. Plant pathogenic fungi  
23 were relatively unspecialized in their interactions with insects inhabiting dead wood, while  
24 interactions between the insects and wood-decay fungi exhibited the highest degree of  
25 specialization, which was similar to estimates for animal-mediated seed dispersal networks in  
26 previous studies. The low degree of specialization for insect symbiont fungi was unexpected.  
27 In general, the pooled insect-fungus networks were significantly more specialized, more  
28 modular and less nested than randomized networks. Thus, the detritivore networks had an  
29 unusual anti-nested structure. Future studies might corroborate whether this is a common  
30 aspect of networks based on interactions with fungi, possibly due to their often intense  
31 competition for substrate.

## 32 Introduction

33 Interactions between species shape ecological communities and networks, and drive  
34 evolution. Ecosystems therefore consist of complex networks that vary in structure depending  
35 on the specificity and frequency of the interacting species. Highly specific species interactions  
36 often result in very specialized networks with low robustness to species loss[1], where  
37 extinction of one species also leads to the loss of connected species from the network. As

38 species are currently going extinct at an alarmingly high rate[2], knowledge of ecological  
39 networks and interactions is becoming increasingly important in order to understand and  
40 hopefully prevent extinction cascades.

41 Several studies have underlined the importance of pollination and other well-known  
42 interactions such as predation, herbivory and animal-mediated seed dispersal for ecosystem  
43 structure and function (e.g.[3-5]). However, our knowledge of biotic interactions is highly  
44 skewed towards macroscopic organisms[6], and network studies have largely focused on  
45 well-known interactions such as pollination[7, 8]. There are few studies of interactions  
46 between bacteria, fungi or invertebrates at the community level, despite their overwhelming  
47 abundance and species diversity[9-12]. Bacteria and fungi are integral to terrestrial and  
48 freshwater ecosystems through their roles as pathogens, symbionts and decomposers[13-17].  
49 Up to 90% of terrestrial plant production enters the detrital food chain[18], where the  
50 microbiota of bacteria, fungi and invertebrates determine rate of decomposition and carbon  
51 sequestration[16, 17].

52 Invertebrates can have a significant influence on ecosystem processes through interactions  
53 with bacteria or fungi, as demonstrated for rate of decomposition, nutrient cycling and  
54 mycorrhizal symbiosis in lab experiments[19-21]. However, the role of invertebrates in the  
55 detritivore community is rarely considered from a network perspective, in contrast with the  
56 intensively studied functional roles of invertebrates as pollinators or herbivores[7, 8]. In the  
57 present study, we show that network analysis of understudied species groups such as insects  
58 and fungi can reveal hidden interactions and elucidate the structure of detritivore  
59 communities.

60 Ecological networks are shaped by the frequency of interactions between species, which in  
61 turn is partly determined by abundance of the species and their interaction specialization. The  
62 tendency of species in a network to exhibit specialized interactions can be summed up at the

63 network level as degree of specialization[22, 23]. For instance, as pollinators are generally  
64 more specialized in their resource use than seed-feeding animals, pollination networks in  
65 general have a higher degree of specialization than networks based on animal-mediated seed  
66 dispersal[22].

67 If specialist species mainly interact with a proper subset of the interaction partners of  
68 generalist species, this results in a nested network structure. Nested networks are generally  
69 robust against random species loss[24], while networks with a high degree of specialization  
70 are more vulnerable[25]. Networks can also be organized into compartments called modules,  
71 in which species interact frequently within the modules and infrequently between modules. If  
72 within-module interactions are dominant in number, the network is said to have high  
73 modularity[26]. Modules might be the product of spatial or temporal variability in  
74 interactions, for instance if interaction frequency depends on overlap in phenology, or they  
75 might consist of closely related species or species with similar trait syndromes due to  
76 convergent evolution[27, 28]. Thus, the structure of an interaction network can reveal  
77 selective pressures shaping the interactions and the robustness of networks to species loss.

78 In the present study, we analyse specialization, nestedness and modularity of insect-fungus  
79 networks sampled from dead wood in boreal forests. These networks are vital for the  
80 functioning of forest ecosystems, as they are the driving force for decomposition and nutrient  
81 cycling in these habitats[29-31]. Understanding how these networks are structured is therefore  
82 integral to understanding the basis for ecosystem processes in forests. We used DNA  
83 metabarcoding to identify fungi extracted from individual insects, which enabled us to include  
84 interactions involving microscopic fungal structures such as spores, hyphae or yeast. We  
85 compiled quantitative (i.e. weighted) networks for interactions between insects inhabiting  
86 dead wood and three functional groups of fungi; insect symbiont fungi, wood-decay fungi and  
87 plant pathogenic fungi.

88 As we do not have replicates of each network, this study is not a test of differences between  
89 these groups, but rather an exploratory first step into largely uncharted territory for network  
90 analysis in terms of both methodology[6] (i.e. the combination of DNA metabarcoding and  
91 quantitative networks) and study organisms (i.e. detritivorous insects and fungi). In line with  
92 the few comparable previous studies[32, 33], we demonstrate that such novel network  
93 analysis might reveal network structures differing from those of more well-known  
94 interactions, underlining the necessity of expanding the scope of network studies.

## 95 Methods

96 This study is based on data from Jacobsen et al.[34], where a more detailed description of  
97 insect sampling, DNA-analysis and bioinformatics can be found.

98 We sampled beetles from recently cut logs of aspen (*Populus tremula* L.) that had been placed  
99 at eight sites in two production forests in south-eastern Norway; Losby forest holdings (Lat.  
100 55.98, Long.10.68, 150–300 m.a.s.l.) and Løvenskiold-Vækerø (LV) forest holdings (Lat.  
101 54.49, Long. 21.24, 200–500 m.a.s.l.). Both forest landscapes lie within the southern boreal  
102 vegetation zone[35] and consist mainly of spruce (*Picea abies* (L.) H.Karst.), with pine (*Pinus*  
103 *sylvestris* L.), birch (*Betula pubescens* Ehrh.) and aspen as subdominants.

104 Beetles were sampled individually with tweezers directly from the logs or from sticky traps  
105 on the logs, on eleven occasions during May to August in 2014 and 2015. The sticky traps  
106 were exposed for one or two days prior to insect sampling. The tweezers were sterilized with  
107 ethanol and fire between handling of each insect. The insects were killed by freezing at –  
108 80°C and identified to species or genus in a sterile environment using sterilized equipment.  
109 Insects that could not be confidently identified at least to genus by the first author (RMJ) were  
110 not analysed further (13 of 654 individuals). We selected 343 wood-inhabiting beetle

111 individuals, i.e. species or genera with larval development either in dead wood or in fungal  
112 fruit bodies on dead wood[36, 37], for analysis of fungal DNA.

113 Fungal DNA was extracted from the beetles following a modified CTAB protocol[38] and  
114 amplified by polymerase chain reaction (PCR) on an Eppendorf Thermal Cycler (VWR,  
115 Radnor, USA) using primers ITS4[39] and fITS7[40]. The PCR products were cleaned using  
116 Wizard® SV Gel and PCR Clean-Up System (Promega, Madison, USA) and pooled  
117 according to strength of the bands in gel electrophoresis. Pooled samples were cleaned with  
118 the ChargeSwitch® kit (Invitrogen, California, USA), DNA-concentration was measured with  
119 the Qubit® BR DNA kit (Invitrogen, California, USA), and the sample quality was confirmed  
120 by Nanodrop™ (Thermo Fisher Scientific, Madison, USA). The samples were submitted to  
121 GATC Biotech for adaptor-ligation and Illumina HiSeq Rapid Run 300bp paired-end  
122 sequencing. Quality control and clustering of the resulting sequences was conducted with the  
123 SCATA pipeline (<https://scata.mykopat.slu.se/>, accessed 5<sup>th</sup> of July 2016). The sequences  
124 were subsampled to 10 000 per beetle sample prior to clustering. Taxonomy was assigned to  
125 the representative sequences of each OTU taking the top hit of a Basic Local Alignment  
126 Search Tool (BLASTn[41]) search against the NCBI (National Centre for Biotechnology  
127 Information) and UNITE[42] databases. OTUs with e-values < e-10 and bit-scores > 100 were  
128 annotated to species level if ITS homology was 100 - 98%, genus for 97.9 - 95%, family or  
129 order for 94.9 – 80%, phylum for 79.9 – 70% and “Fungus” for lower homology or e-values >  
130 e-10 and bit-scores < 100. Taxonomy was updated according to the taxonomic database  
131 Dyntaxa (<https://www.dyntaxa.se/>, accessed 24th of February 2017) and MycoBank  
132 (<http://www.mycobank.org/>, export date 26th of October 2017). For further statistical analysis  
133 only OTUs represented by at least 20 reads were included, since we wanted to focus on  
134 widespread fungi more likely to be important in interactions.

### 135 **Classification of fungal functional groups**

136 Fungal OTUs annotated to species or genus level were analysed further, in networks including  
137 all OTUs (Supplementary Table S1) and in networks including OTUs classified into  
138 functional groups based on the FUNGuild database[43] and various literature (see  
139 Supplementary Table S2, S3 and S4). Groups were non-overlapping. We analysed networks  
140 with the three most abundant (in terms of number of sequences) functional groups:

141 1. Insect symbionts (Supplementary Table S5); this group included known insect symbionts  
142 such as *Ophiostoma* spp. or *Phialophoropsis* spp., and yeast species isolated from insect guts  
143 in previous studies such as *Candida* spp. and *Cryptococcus* spp., that were assumed to be  
144 endosymbionts. Fungal parasites or pathogens of insects were not included (only eight OTUs  
145 matched these functional groups according to FUNGuild).

146 2. Wood-decayers (Supplementary Table S6); this group included fungi in the class  
147 Agaricomycetes known to inhabit dead wood, in which the majority of species produce large  
148 fruit bodies and large quantities of spores that attract spore-feeding insects during sporulation  
149 (e.g.[44, 45]).

150 3. Plant pathogens (Supplementary Table S7); this group included pathogens of living plants.  
151 Plant pathogenic fungi known to be insect symbionts such as *Ophiostoma* spp. were excluded,  
152 since these functional groups were meant to mainly reflect the relationship between the  
153 insects and the fungi.

## 154 **Statistics**

155 All analyses were conducted in R version 3.3.2[46].

156 The number of beetle individuals in which each fungal OTU occurred was used as a basis for  
157 quantitative/weighted networks. Excluding insect species represented by single individuals  
158 did not change the results and these species were therefore included in the network analysis.

159 Network specialization was estimated by the standardized two-dimensional Shannon entropy

160  $H_2'$ [47] using the package bipartite v. 2.07[48]. This index defines the degree of  
161 specialization in a network as the deviation from the expected probability distribution of  
162 interactions, which assumes that a species interacts with another species in proportion to its  
163 total frequency of occurrence in the network (i.e. terminal row or column sums). We  
164 estimated the species-level specialization by the standardized Kullback-Leibler distance  
165  $d'$ [47]. The species-level specialization index is defined as a species' deviation of the  
166 utilization of potential partners that is expected based on their terminal row or column sums,  
167 i.e. link numbers in the fungus x insect interaction matrix. Both  $H_2'$  and  $d'$  range from 0 for  
168 most generalized to 1 for most specialized.

169 Modularity of the networks was estimated with the QuanBiMo algorithm developed by  
170 Dormann and Strauss[26] and implemented as function "computeModules" in the bipartite  
171 package. Modularity  $Q$  ranges from 0, meaning that there are no more links between species  
172 in a module than expected by chance, to 1 which signifies maximum modularity for the  
173 network. As the QuanBiMo algorithm is based on a stochastic process, we estimated  
174 modularity ten times for each network and reported the mean value. To estimate nestedness of  
175 the network, we used the weighted version of the nestedness metric based on overlap and  
176 decreasing fill, abbreviated WNODF[49]. This metric ranges from 0 for networks without  
177 nested structure, to 100 for perfectly nested networks.

178 We tested the statistical significance of the metrics for each network by simulating null  
179 models ( $n=1000$ ). Null model P followed Patefield's algorithm[50] as implemented in the  
180 function "r2dtable" in R, which randomises network interactions with the restriction of fixed  
181 marginal sums (i.e. the sum of interactions for each species was kept constant). We also tested  
182 null model V, which in addition to fixed marginal sums also keeps connectance (i.e.  
183 proportion of realised links in the interaction matrix) of the network constant as proposed by  
184 Vazquez et al.[51] and implemented in function "quasiswap\_count" in the vegan package v.

185 2.4-2. We performed two-sided tests of the network metric value against the distribution of  
186 the null model metric values. Finally, we repeated all network analyses for subsets of the  
187 insect-fungus networks with species numbers standardised to those of the smallest network.

## 188 Results

189 Fungal DNA was obtained from 187 saproxylic beetle individuals of 17 species or genera  
190 (Supplementary Table S8). The DNA metabarcoding analyses resulted in 1069 fungal  
191 operational taxonomic units (OTUs) represented by more than 20 sequences and distributed  
192 on a total of 1 714 063 sequences. Of these OTUs, 449 were annotated to species or genus and  
193 analysed further in networks with the insects, either including all fungi or separated into  
194 functional groups; 35 species or genera of fungi (356 279 sequences) were classified as insect  
195 symbionts, 22 (48 196 sequences) were classified as wood-decayers in the class  
196 Agaricomycetes and 61 (158 133 sequences) were classified as plant pathogens (Fig. 1).

197 All insect-fungus networks were significantly more specialized, more modular and less nested  
198 than the null model with randomized interactions (Fig. 2). The results were relatively similar  
199 when compared with the null model which also had constant connectance (Supplementary  
200 Fig. S1), and for the subsampled networks with standardised species numbers (Supplementary  
201 Fig. S2). The network with wood-decay fungi had the highest degree of specialization and  
202 modularity ( $H_2' = 0.21$ ,  $Q = 0.28$ , Fig. 2). Correspondingly, it also had the lowest nestedness  
203 (WNODF = 16.14, Fig. 2). However, when comparing standardized values of nestedness (real  
204 value – mean value of randomization / SE of randomizations), the network with plant  
205 pathogenic fungi had the lowest values (-5.68 standardized WNODF compared with -3.14 for  
206 wood-decay fungi and -3.29 for insect symbiont fungi).

207 We re-calculated the network metrics with OTUs annotated as *Chondrostereum purpureum*  
208 (Pers. : Fr.) Pouzar excluded from the network of wood-decayers, since this species was

209 visibly fruiting on the logs during insect sampling and could have occurred in all samples  
210 indiscriminately. Indeed, DNA from *C. purpureum* was isolated from 43% of the insect  
211 samples, including 12 of 17 taxa. Excluding *C. purpureum* from the wood-decayer network  
212 resulted in even higher specialization ( $H_2' = 0.29$ , null model P 95% CI = 0.13 – 0.23), higher  
213 modularity ( $Q = 0.40$ , null model P 95% CI = 0.28 – 0.36) and lower nestedness (WNODF =  
214 9.38, null model P 95% CI = 12.00 – 23.23, albeit higher standardized WNODF = -2.9 due to  
215 lower SE). Without *C. purpureum*, the network between wood-inhabiting beetles and wood-  
216 decay fungi was organised in six modules (Fig. 3).

217 We estimated specialization at the species level for all networks (Supplementary Tables S9-  
218 S14), but focus here on interactions with the more well-known wood-decay fungi. In  
219 interactions with wood-decay fungi, the insect species *Endomychus coccineus* (Linnaeus,  
220 1758) was significantly (P-value = 0.005) more specialized and *Glischrochilus hortensis*  
221 (Geoffroy, 1785) was nearly significantly (P-value = 0.053) more specialized than expected  
222 from the null model (Supplementary Table S11), with index values ( $d'$ ) of 0.25 and 0.18,  
223 respectively. Among the wood-decay fungi, OTUs annotated as *Trametes versicolor* (L. : Fr.)  
224 Pilát., *Fomes fomentarius* (L. : Fr.) Fr. and *Sistotrema brinkmannii* (Bres.) J. Erikss. were  
225 significantly specialized with index values of 0.45, 0.38 and 0.24 (P-values < 0.05),  
226 respectively (Supplementary Table S12).

## 227 Discussion

228 This study shows that species of two very diverse eukaryotic kingdoms, insects and fungi,  
229 interact in structured networks. The networks had an anti-nested structure (i.e. they were less  
230 nested than randomized networks), they were specialized, though not to a high degree, and  
231 interacting species were compartmentalized in modules. The lack of a nested network  
232 structure might indicate a relatively low species redundancy, which could mean that the

233 insect-fungus networks are vulnerable to species loss[7], although the relatively low degree of  
234 specialization ( $H_2' = 0.21$  or less) might increase robustness[1] and species within modules  
235 might fulfil similar interaction functions.

236 Although *non-nested* structures have been demonstrated more often for quantitative, weighted  
237 networks than for qualitative, binary networks, *anti-nested* structures do not seem to be  
238 common for either network type[52]. However, previous studies using molecular methods to  
239 identify mycorrhizal fungi interacting with plants have also documented anti-nested  
240 networks[32, 33, 53]. Toju et al.[33] found that this anti-nested structure seemed to be  
241 explained by reduced fungal host range overlap, causing a checkerboard pattern of  
242 interactions. They suggested that this pattern might be caused by competitive exclusion by the  
243 fungi, preventing other species of fungi from interacting with their plant host. Although the  
244 insects in the present study are not presumed to function as a substrate and thus a site of  
245 competition for the fungi (possibly with the exception of the symbiont fungi), their  
246 interactions with the fungi might reflect competitive exclusion structuring fungal communities  
247 at shared habitats such as dead wood, where competition for substrate can be fierce[54].  
248 Future studies might confirm whether anti-nestedness is a common aspect of interaction  
249 networks involving fungi.

250 Both degree of specialization ( $H_2' = 0.15$ ) and modularity ( $Q = 0.15$ ) were relatively low for  
251 the network between plant-pathogenic fungi and insects. Although there are examples of  
252 plant-pathogenic fungi being dispersed by insects in species-specific interactions[55], the  
253 insects analysed in the present study only included species inhabiting dead wood. Thus, it is  
254 not unexpected that their interactions with pathogens of *living* plants were relatively  
255 unspecific, perhaps only based on shared forest habitats. Furthermore, plant pathogenic fungi  
256 known to be symbionts of insects, such as *Ophiostoma* spp., were classified as insect  
257 symbionts rather than plant pathogens, as we considered this to be the aspect of their ecology

258 most likely to affect their interaction networks with insects. The versatile ecology of fungi,  
259 where trophic mode might vary depending on context, complicates classification into  
260 functional groups[43]. Fungi have been documented to shift between an endophytic and a  
261 plant pathogenic lifestyle, or between a mycorrhizal and a saprotrophic lifestyle, to mention a  
262 few of the examples summarized by Selosse et al.[56]. Thus, the classification of fungi into  
263 functional groups in the present study is likely to be highly simplified and relatively uncertain  
264 for some taxa, especially the insect symbionts. Nevertheless, this tentative classification  
265 allows us to explore differences in the structure of networks involving different groups of  
266 fungi and build hypotheses for further studies.

267 The network with fungi annotated as insect symbionts had a surprisingly low degree of  
268 specialization ( $H_2' = 0.11$ ). This group included fungi that might live in mutualistic or  
269 commensalistic symbiosis with insects, as insect parasites and pathogens were not included.  
270 Most of these species were classified as insect symbionts based on previous isolation from  
271 beetle guts (references in Supplementary Table S2). In comparison, in a study by Shukla et  
272 al.[57] bacterial endosymbionts had a relatively high degree of specialization ( $H_2' = 0.35$ )  
273 even in an intraspecific network with males, females and larvae of one dung beetle species.  
274 Modularity was also relatively low ( $Q = 0.15$ ), considering that intimate interactions tend to  
275 result in more modular networks[58]. Our results indicate that many of the fungal species  
276 found in insect guts might be unspecific symbionts, or simply contaminants from food or  
277 habitat that do not function as symbionts. Certainly, yeast fungi like *Candida* spp. and  
278 *Cryptococcus* spp. can occur in several different environments such as soil or dead wood[59-  
279 62], where insects are also abundant. Some of the fungi isolated from beetle guts do seem to  
280 be more closely associated with the habitat than with the beetle species[63]. However,  
281 endosymbionts can be relatively unspecific with regard to insect host species, especially if  
282 they are transmitted horizontally[15]. Further in-depth studies, including microscopy and

283 experimentation, are required to clarify to what extent fungi such as *Candida mesenterica* and  
284 the other taxa tentatively classified as insect symbionts in the present study spend part of their  
285 life living as symbionts on or in insects, and whether this affects the insects.

286 The network between wood-inhabiting beetles and wood-decay fungi had the highest degree  
287 of specialization in this study ( $H_2' = 0.21$ ). However, this is still much lower than the  
288 specialization of pollinator-plant networks ( $H_2' = 0.60$ [22]), ant-myrmecophyte networks ( $H_2'$   
289  $= 0.80$ [22, 64]) or legume-rhizobium bacteria networks ( $H_2' = 0.85$ [65]). Instead, it was closer  
290 to that of networks based on more opportunistic interactions, such as ants harvesting  
291 honeydew from true bugs ( $H_2' = 0.43$ [23]) or nectar from plants ( $H_2' = 0.25$ [22]), or animal-  
292 mediated seed dispersal ( $H_2' = 0.18 - 0.47$ [22, 66, 67]). This indicates that the network  
293 between wood-inhabiting beetles and wood-decay fungi was based upon similarly  
294 opportunistic yet reciprocal interactions that would result in a moderate degree of  
295 specialization. Spore feeding and subsequent spore dispersal by the beetles could represent  
296 such an interaction[34]. In line with this hypothesis, the nitidulid beetle *G. hortensis* has  
297 frequently been registered on sporulating fruit bodies of wood-decay fungi such as the  
298 polypore *F. fomentarius*[44, 45], although its habitat is fresh dead wood[36]. In the present  
299 study, this beetle species was found to be significantly more specialized on wood-decay fungi  
300 than expected by chance, and *F. fomentarius* was isolated from eleven individuals of *G.*  
301 *hortensis*. This beetle species might therefore function as a moderately specific propagule  
302 vector for *F. fomentarius*, providing targeted dispersal to fresh dead wood[34]. Although the  
303 network between wood-living beetles and wood-decay fungi might be a food web without  
304 dispersal benefits to the fungi, the beetles were sampled from dead wood that had recently  
305 been cut and placed in these forests, without any other visible fungal fruit bodies than those of  
306 *C. purpureum*.

307 If the network between wood-inhabiting beetles and wood-decay fungi was based on spore  
308 feeding and dispersal, its degree of specialization might be constrained by the same factors  
309 that limit specialization of animal-mediated seed dispersal networks[68]. Optimal dispersal of  
310 both spores and seeds requires the propagule vector to move away from the source and deliver  
311 the propagule not to a conspecific, but to a suitable habitat. The propagule source has no  
312 means to direct the vector, its only chance is to attract vectors that share its habitat. Fungal  
313 odour has been shown to attract several different species of beetles inhabiting dead wood[69-  
314 71], and odour release increases during sporulation[72]. *F. fomentarius* and certain other  
315 polypore species also aggregate spores on top of their fruit bodies, which are visited by  
316 several wood-inhabiting insects[44]. Aggregation of spores and increased odour emission  
317 during sporulation thus seem to function as attractants to wood-inhabiting insects, in much the  
318 same way as brightly coloured fruits attract seed dispersing animals. As such, there is a basis  
319 for selection favouring a certain degree of reciprocity and specialization between wood-decay  
320 fungi and insects. However, spore dispersal effectiveness would be low if the insects were  
321 highly specialized spore-feeders that only moved between sporulating fruit bodies, without  
322 dispersing the spores to unoccupied substrates. For seed dispersal, it has been shown that  
323 generalist frugivores can be very effective seed dispersers[73, 74] and that species in highly  
324 diverse frugivore assemblages fulfil complementary roles[75, 76]. These mechanisms  
325 promote diversified interactions and generalized dispersal systems[77], restraining the degree  
326 of specialization in seed dispersal networks[22, 66, 67] and possibly in the potential spore  
327 dispersal network in the present study.

328 It should be noted that certain aspects of network structure can be subject to strong spatial and  
329 temporal variability[67, 78, 79]. Our networks were based on pooled datasets of beetles  
330 sampled over two seasons in two different landscapes, but the necessity of sampling beetles  
331 individually resulted in a sample size that was too low to explore spatial and temporal

332 variability in network structure. However, the distribution of sampled individuals was  
333 relatively even between landscapes, and the majority of individuals were sampled during the  
334 second year. Furthermore, network level measures tend to exhibit a lower temporal and spatial  
335 variability than species level measures[79]. In any case, our study is but an exploratory first  
336 step into novel methodology and understudied interactions, which can hopefully provide  
337 future research with a foundation for important working hypotheses regarding detritivore  
338 networks and the use of DNA metabarcoding for discerning microscopic interactions.

339 In conclusion, our results demonstrate that there is a diversity of hidden interactions in  
340 detritivore networks. These interactions could have significant influence on fungal  
341 communities in dead wood[62, 80], and thereby affect important ecosystem functions such as  
342 carbon sequestration and decomposition[31]. We encourage the use of molecular methods to  
343 include microscopic organisms in future network studies [6], as the unusual network  
344 structures demonstrated in this and previous studies[32, 33, 53] underline the importance of  
345 expanding the scope of network analysis to understudied and functionally important  
346 organisms such as fungi.

#### 347 **Data Availability**

348 Raw data (fastq-files), barcode and primer mapping file, OTU table and representative  
349 sequence files have been accessioned in Dryad with <http://dx.doi.org/10.5061/dryad.3t2d4>.

#### 350 **References**

- 351 [1] Pocock, M.J., Evans, D.M. & Memmott, J. 2012 The robustness and restoration of a  
352 network of ecological networks. *Science* **335**, 973-977.  
353 [2] Barnosky, A.D., Matzke, N., Tomiya, S., Wogan, G.O.U., Swartz, B., Quental, T.B.,  
354 Marshall, C., McGuire, J.L., Lindsey, E.L. & Maguire, K.C. 2011 Has the Earth's sixth mass  
355 extinction already arrived? *Nature* **471**, 51-57.  
356 [3] Biesmeijer, J.C., Roberts, S.P., Reemer, M., Ohlemüller, R., Edwards, M., Peeters, T.,  
357 Schaffers, A., Potts, S.G., Kleukers, R. & Thomas, C. 2006 Parallel declines in pollinators  
358 and insect-pollinated plants in Britain and the Netherlands. *Science* **313**, 351-354.

- 359 [4] Peres, C.A., Emilio, T., Schietti, J., Desmoulière, S.J. & Levi, T. 2016 Dispersal limitation  
360 induces long-term biomass collapse in overhunted Amazonian forests. *Proceedings of the*  
361 *National Academy of Sciences* **113**, 892-897.
- 362 [5] Ripple, W.J. & Beschta, R.L. 2012 Trophic cascades in Yellowstone: The first 15 years  
363 after wolf reintroduction. *Biol. Conserv.* **145**, 205-213.
- 364 [6] Toju, H. 2015 High-throughput DNA barcoding for ecological network studies. *Popul.*  
365 *Ecol.* **57**, 37-51.
- 366 [7] Bascompte, J. & Jordano, P. 2007 Plant-animal mutualistic networks: the architecture of  
367 biodiversity. *Annu. Rev. Ecol. Evol. Syst.* **38**, 567-593.  
368 (doi:10.1146/annurev.ecolsys.38.091206.095818).
- 369 [8] Ings, T.C., Montoya, J.M., Bascompte, J., Blüthgen, N., Brown, L., Dormann, C.F.,  
370 Edwards, F., Figueroa, D., Jacob, U. & Jones, J.I. 2009 Ecological networks – beyond food  
371 webs. *J. Anim. Ecol.* **78**, 253-269.
- 372 [9] Hamilton, A.J., Basset, Y., Benke, K.K., Grimbacher, P.S., Miller, S.E., Novotny, V.,  
373 Samuelson, G.A., Stork, N.E., Weiblen, G.D. & Yen, J.D.L. 2010 Quantifying uncertainty in  
374 estimation of tropical arthropod species richness. *The American Naturalist* **176**, 90-95.  
375 (doi:10.1086/652998).
- 376 [10] Hamilton, A.J., Basset, Y., Benke, K.K., Grimbacher, P.S., Miller, S.E., Novotný, V.,  
377 Samuelson, G.A., Stork, N.E., Weiblen, G.D. & Yen, J.D.L. 2011 Correction. *The American*  
378 *Naturalist* **177**, 544-545. (doi:10.1086/659643).
- 379 [11] Hawksworth, D. 2012 Global species numbers of fungi: are tropical studies and  
380 molecular approaches contributing to a more robust estimate? *Biodivers. Conserv.* **21**, 2425-  
381 2433.
- 382 [12] Locey, K.J. & Lennon, J.T. 2016 Scaling laws predict global microbial diversity.  
383 *Proceedings of the National Academy of Sciences* **113**, 5970–5975.  
384 (doi:10.1073/pnas.1521291113).
- 385 [13] Benítez, M.-S., Hersh, M.H., Vilgalys, R. & Clark, J.S. 2013 Pathogen regulation of  
386 plant diversity via effective specialization. *Trends Ecol. Evol.* **28**, 705-711.
- 387 [14] Clemmensen, K., Bahr, A., Ovaskainen, O., Dahlberg, A., Ekblad, A., Wallander, H.,  
388 Stenlid, J., Finlay, R., Wardle, D. & Lindahl, B. 2013 Roots and associated fungi drive long-  
389 term carbon sequestration in boreal forest. *Science* **339**, 1615-1618.
- 390 [15] Engel, P. & Moran, N.A. 2013 The gut microbiota of insects—diversity in structure and  
391 function. *FEMS Microbiol. Rev.* **37**, 699-735.
- 392 [16] Gessner, M.O., Swan, C.M., Dang, C.K., McKie, B.G., Bardgett, R.D., Wall, D.H. &  
393 Hättenschwiler, S. 2010 Diversity meets decomposition. *Trends Ecol. Evol.* **25**, 372-380.
- 394 [17] Nielsen, U.N., Ayres, E., Wall, D.H. & Bardgett, R.D. 2011 Soil biodiversity and carbon  
395 cycling: a review and synthesis of studies examining diversity–function relationships. *Eur. J.*  
396 *Soil Sci.* **62**, 105-116.
- 397 [18] Cebrian, J. 1999 Patterns in the fate of production in plant communities. *The American*  
398 *Naturalist* **154**, 449-468.
- 399 [19] A'Bear, A.D., Jones, T.H. & Boddy, L. 2014 Size matters: What have we learnt from  
400 microcosm studies of decomposer fungus–invertebrate interactions? *Soil Biol. Biochem.* **78**,  
401 274-283.
- 402 [20] De Meester, N., Gingold, R., Rigaux, A., Derycke, S. & Moens, T. 2016 Cryptic  
403 diversity and ecosystem functioning: a complex tale of differential effects on decomposition.  
404 *Oecologia* **182**, 559-571.
- 405 [21] Gange, A.C., Bower, E. & Brown, V.K. 2002 Differential effects of insect herbivory on  
406 arbuscular mycorrhizal colonization. *Oecologia* **131**, 103-112.
- 407 [22] Blüthgen, N., Menzel, F., Hovestadt, T., Fiala, B. & Blüthgen, N. 2007 Specialization,  
408 constraints, and conflicting interests in mutualistic networks. *Curr. Biol.* **17**, 341-346.

- 409 [23] Ivens, A.B., von Beeren, C., Blüthgen, N. & Kronauer, D.J. 2016 Studying the complex  
410 communities of ants and their symbionts using ecological network analysis. *Annu. Rev.*  
411 *Entomol.* **61**, 353-371.
- 412 [24] Thébault, E. & Fontaine, C. 2010 Stability of ecological communities and the  
413 architecture of mutualistic and trophic networks. *Science* **329**, 853-856.
- 414 [25] Kaiser-Bunbury, C.N., Mougial, J., Whittington, A.E., Valentin, T., Gabriel, R., Olesen,  
415 J.M. & Blüthgen, N. 2017 Ecosystem restoration strengthens pollination network resilience  
416 and function. *Nature* **542**, 223-227.
- 417 [26] Dormann, C.F. & Strauss, R. 2014 A method for detecting modules in quantitative  
418 bipartite networks. *Methods in Ecology and Evolution* **5**, 90-98.
- 419 [27] Lewinsohn, T.M., Inácio Prado, P., Jordano, P., Bascompte, J. & M Olesen, J. 2006  
420 Structure in plant–animal interaction assemblages. *Oikos* **113**, 174-184.
- 421 [28] Olesen, J.M., Bascompte, J., Dupont, Y.L. & Jordano, P. 2007 The modularity of  
422 pollination networks. *Proceedings of the National Academy of Sciences* **104**, 19891-19896.
- 423 [29] Fekete, I., Kotroczó, Z., Varga, C., Nagy, P.T., Várбірó, G., Bowden, R.D., Tóth, J.A. &  
424 Lajtha, K. 2014 Alterations in forest detritus inputs influence soil carbon concentration and  
425 soil respiration in a Central-European deciduous forest. *Soil Biol. Biochem.* **74**, 106-114.
- 426 [30] Ulyshen, M.D. 2016 Wood decomposition as influenced by invertebrates. *Biological*  
427 *Reviews* **91**, 70-85. (doi:10.1111/brv.12158).
- 428 [31] van der Wal, A., Geydan, T.D., Kuyper, T.W. & de Boer, W. 2013 A thready affair:  
429 linking fungal diversity and community dynamics to terrestrial decomposition processes.  
430 *FEMS Microbiol. Rev.* **37**, 477-494.
- 431 [32] Toju, H., Guimarães, P.R., Olesen, J.M. & Thompson, J.N. 2014 Assembly of complex  
432 plant–fungus networks. *Nature Communications* **5**, 5273. (doi:10.1038/ncomms6273).
- 433 [33] Toju, H., Guimarães, P.R., Olesen, J.M. & Thompson, J.N. 2015 Below-ground plant–  
434 fungus network topology is not congruent with above-ground plant–animal network topology.  
435 *Science Advances* **1**, e1500291. (doi:10.1126/sciadv.1500291).
- 436 [34] Jacobsen, R.M., Kauserud, H., Sverdrup-Thygeson, A., Bjorbækmo, M.M. & Birkemoe,  
437 T. 2017 Wood-inhabiting insects can function as targeted vectors for decomposer fungi.  
438 *Fungal Ecology* **29**, 76-84.
- 439 [35] Moen, A. 1998 Nasjonalatlas for Norge: Vegetasjon (Norwegian National Atlas:  
440 Vegetation). *Norwegian Mapping Authority, Hønefoss*.
- 441 [36] Dahlberg, A. & Stokland, J.N. 2004 Vedlevande arters krav på substrat. *Skogsstyrelsen,*  
442 *Rapport 7*, 1-74.
- 443 [37] Wheeler, Q. & Blackwell, M. 1984 *Fungus-insect relationships: perspectives in ecology*  
444 *and evolution*, Columbia University Press.
- 445 [38] Murray, M. & Thompson, W.F. 1980 Rapid isolation of high molecular weight plant  
446 DNA. *Nucleic Acids Res.* **8**, 4321-4326.
- 447 [39] White, T.J., Bruns, T., Lee, S. & Taylor, J. 1990 Amplification and direct sequencing of  
448 fungal ribosomal RNA genes for phylogenetics. In *PCR Protocols: A Guide to Methods and*  
449 *Applications* (eds. M. Innis, D. Gelfand, J. Sninsky & T. White), pp. 315-322. San Diego,  
450 CA, Academic Press.
- 451 [40] Ihrmark, K., Bodeker, I.T.M., Cruz-Martinez, K., Friberg, H., Kubartova, A., Schenck,  
452 J., Strid, Y., Stenlid, J., Brandstrom-Durling, M., Clemmensen, K.E., et al. 2012 New primers  
453 to amplify the fungal ITS2 region - evaluation by 454-sequencing of artificial and natural  
454 communities. *FEMS Microbiol. Ecol.* **82**, 666-677. (doi:10.1111/j.1574-6941.2012.01437.x).
- 455 [41] Altschul, S.F., Gish, W., Miller, W., Myers, E.W. & Lipman, D.J. 1990 Basic Local  
456 Alignment Search Tool. *J. Mol. Biol.* **215**, 403-410. (doi:10.1006/jmbi.1990.9999).
- 457 [42] Abarenkov, K., Henrik Nilsson, R., Larsson, K.H., Alexander, I.J., Eberhardt, U., Erland,  
458 S., Høiland, K., Kjøller, R., Larsson, E. & Pennanen, T. 2010 The UNITE database for

459 molecular identification of fungi—recent updates and future perspectives. *New Phytol.* **186**,  
460 281-285.

461 [43] Nguyen, N.H., Song, Z., Bates, S.T., Branco, S., Tedersoo, L., Menke, J., Schilling, J.S.  
462 & Kennedy, P.G. 2016 FUNGuild: an open annotation tool for parsing fungal community  
463 datasets by ecological guild. *Fungal Ecology* **20**, 241-248.

464 [44] Hågvar, S. 1999 Saproxylic beetles visiting living sporocarps of *Fomitopsis pinicola* and  
465 *Fomes fomentarius*. *Nor. J. Entomol.* **46**, 25-32.

466 [45] Schigel, D.S. 2011 Polypore-beetle associations in Finland. *Ann. Zool. Fenn.* **48**, 319-  
467 348.

468 [46] R Core Team. 2016 R: A language and environment for statistical computing. (Vienna,  
469 Austria, R Foundation for Statistical Computing.

470 [47] Blüthgen, N., Menzel, F. & Blüthgen, N. 2006 Measuring specialization in species  
471 interaction networks. *BMC Ecol.* **6**, 9.

472 [48] Dormann, C.F., Gruber, B. & Fründ, J. 2008 Introducing the bipartite package: analysing  
473 ecological networks. *R News* **8**, 8-11.

474 [49] Almeida-Neto, M. & Ulrich, W. 2011 A straightforward computational approach for  
475 measuring nestedness using quantitative matrices. *Environmental Modelling & Software* **26**,  
476 173-178.

477 [50] Patefield, W. 1981 Algorithm AS 159: an efficient method of generating random  $R \times C$   
478 tables with given row and column totals. *Journal of the Royal Statistical Society. Series C*  
479 (*Applied Statistics*) **30**, 91-97.

480 [51] Vázquez, D.P., Melián, C.J., Williams, N.M., Blüthgen, N., Krasnov, B.R. & Poulin, R.  
481 2007 Species abundance and asymmetric interaction strength in ecological networks. *Oikos*  
482 **116**, 1120-1127.

483 [52] Staniczenko, P.P., Kopp, J.C. & Allesina, S. 2013 The ghost of nestedness in ecological  
484 networks. *Nature Communications* **4**, 1391.

485 [53] Bahram, M., Harend, H. & Tedersoo, L. 2014 Network perspectives of ectomycorrhizal  
486 associations. *Fungal Ecology* **7**, 70-77.

487 [54] Boddy, L. 2000 Interspecific combative interactions between wood-decaying  
488 basidiomycetes. *FEMS Microbiol. Ecol.* **31**, 185-194.

489 [55] Piepenbring, M., Hagedorn, G. & Oberwinkler, F. 1998 Spore liberation and dispersal in  
490 smut fungi. *Botanica Acta* **111**, 444-460.

491 [56] Selosse, M.A., Schneider-Maunoury, L. & Martos, F. 2018 Time to re-think fungal  
492 ecology? Fungal ecological niches are often prejudged. *New Phytol.* **217**, 968-972.

493 [57] Shukla, S.P., Sanders, J.G., Byrne, M.J. & Pierce, N.E. 2016 Gut microbiota of dung  
494 beetles correspond to dietary specializations of adults and larvae. *Mol. Ecol.* **25**, 6092–6106.  
495 (doi:10.1111/mec.13901).

496 [58] Fontaine, C., Guimarães, P.R., Kéfi, S., Loeuille, N., Memmott, J., van Der Putten, W.H.,  
497 van Veen, F.J. & Thébault, E. 2011 The ecological and evolutionary implications of merging  
498 different types of networks. *Ecol. Lett.* **14**, 1170-1181.

499 [59] Baldrian, P., Kolařík, M., Štursová, M., Kopecký, J., Valášková, V., Větrovský, T.,  
500 Žifčáková, L., Šnajdr, J., Rídl, J. & Vlček, Č. 2012 Active and total microbial communities in  
501 forest soil are largely different and highly stratified during decomposition. *The ISME journal*  
502 **6**, 248-258.

503 [60] O'Brien, H.E., Parrent, J.L., Jackson, J.A., Moncalvo, J.-M. & Vilgalys, R. 2005 Fungal  
504 community analysis by large-scale sequencing of environmental samples. *Appl. Environ.*  
505 *Microbiol.* **71**, 5544-5550.

506 [61] Ottosson, E., Kubartová, A., Edman, M., Jönsson, M., Lindhe, A., Stenlid, J. &  
507 Dahlberg, A. 2015 Diverse ecological roles within fungal communities in decomposing logs  
508 of *Picea abies*. *FEMS Microbiol. Ecol.* **91**, fiv012.

- 509 [62] Strid, Y., Schroeder, M., Lindahl, B., Ihrmark, K. & Stenlid, J. 2014 Bark beetles have a  
510 decisive impact on fungal communities in Norway spruce stem sections. *Fungal Ecology* **7**,  
511 47-58.
- 512 [63] Suh, S.-O. & Blackwell, M. 2005 Four new yeasts in the *Candida mesenterica* clade  
513 associated with basidiocarp-feeding beetles. *Mycologia* **97**, 167-177.
- 514 [64] Barriga, P.A., Dormann, C.F., Gbur, E.E. & Sagers, C.L. 2015 Community structure and  
515 ecological specialization in plant–ant interactions. *J. Trop. Ecol.* **31**, 325-334.
- 516 [65] Le Roux, J.J., Mavengere, N.R. & Ellis, A.G. 2016 The structure of legume–rhizobium  
517 interaction networks and their response to tree invasions. *AoB Plants* **8**, plw038.  
518 (doi:10.1093/aobpla/plw038).
- 519 [66] Correa, S.B., Arujo, J.K., Penha, J., da Cunha, C.N., Bobier, K.E. & Anderson, J.T. 2016  
520 Stability and generalization in seed dispersal networks: A case study of frugivorous fish in  
521 Neotropical wetlands. *Proceedings of the Royal Society of London Series B-Biological*  
522 *Sciences* **283**, 20161267.
- 523 [67] Schleuning, M., Blüthgen, N., Flörchinger, M., Braun, J., Schaefer, H.M. & Böhning-  
524 Gaese, K. 2011 Specialization and interaction strength in a tropical plant–frugivore network  
525 differ among forest strata. *Ecology* **92**, 26-36.
- 526 [68] Wheelwright, N.T. & Orians, G.H. 1982 Seed dispersal by animals: contrasts with pollen  
527 dispersal, problems of terminology, and constraints on coevolution. *The American Naturalist*  
528 **119**, 402-413.
- 529 [69] Johansson, T., Olsson, J., Hjältén, J., Jonsson, B.G. & Ericson, L. 2006 Beetle attraction  
530 to sporocarps and wood infected with mycelia of decay fungi in old-growth spruce forests of  
531 northern Sweden. *For. Ecol. Manag.* **237**, 335-341.
- 532 [70] Jonsell, M. & Nordlander, G. 1995 Field attraction of Coleoptera to odours of the wood-  
533 decaying polypores *Fomitopsis pinicola* and *Fomes fomentarius*. *Ann. Zool. Fenn.* **32**, 391-  
534 402.
- 535 [71] Leather, S.R., Baumgart, E.A., Evans, H.F. & Quicke, D.L. 2013 Seeing the trees for the  
536 wood – beech (*Fagus sylvatica*) decay fungal volatiles influence the structure of saproxylic  
537 beetle communities. *Insect Conservation and Diversity* **7**, 314-326.
- 538 [72] Fäldt, J., Jonsell, M., Nordlander, G. & Borg-Karlson, A.-K. 1999 Volatiles of bracket  
539 fungi *Fomitopsis pinicola* and *Fomes fomentarius* and their functions as insect attractants. *J.*  
540 *Chem. Ecol.* **25**, 567-590.
- 541 [73] Carlo, T.A. & Morales, J.M. 2016 Generalist birds promote tropical forest regeneration  
542 and increase plant diversity via rare-biased seed dispersal. *Ecology* **97**, 1819-1831.
- 543 [74] Wehncke, E., Hubbell, S., Foster, R. & Dalling, J. 2003 Seed dispersal patterns produced  
544 by white-faced monkeys: implications for the dispersal limitation of neotropical tree species.  
545 *J. Ecol.* **91**, 677-685.
- 546 [75] Escribano-Avila, G., Calviño-Cancela, M., Pías, B., Virgos, E., Valladares, F. &  
547 Escudero, A. 2014 Diverse guilds provide complementary dispersal services in a woodland  
548 expansion process after land abandonment. *J. Appl. Ecol.* **51**, 1701-1711.
- 549 [76] McConkey, K.R. & Brockelman, W.Y. 2011 Nonredundancy in the dispersal network of  
550 a generalist tropical forest tree. *Ecology* **92**, 1492-1502.
- 551 [77] Schupp, E.W., Jordano, P. & Gómez, J.M. 2010 Seed dispersal effectiveness revisited: a  
552 conceptual review. *New Phytol.* **188**, 333-353.
- 553 [78] Morris, R.J., Sinclair, F.H. & Burwell, C.J. 2015 Food web structure changes with  
554 elevation but not rainforest stratum. *Ecography* **38**, 792-802.
- 555 [79] Trøjelsgaard, K. & Olesen, J.M. 2016 Ecological networks in motion: micro-and  
556 macroscopic variability across scales. *Funct. Ecol.* **30**, 1926-1935.

557 [80] Jacobsen, R.M., Birkemoe, T. & Sverdrup-Thygeson, A. 2015 Priority effects of early  
558 successional insects influence late successional fungi in dead wood. *Ecology and Evolution* **5**,  
559 4896-4905. (doi:10.1002/ece3.1751).

560

## 561 Acknowledgements

562 We would like to thank Sindre Ligaard for advice during insect identification, Synnøve  
563 Botnen for tutoring in DNA analysis, Sundy Maurice and Janina Fuss for advice and help  
564 during lab work, Marit M. Bjorbækmo for indispensable help with bioinformatics, Sebastian  
565 Seibold for sharing data on fungal guilds, Marie Davey and Elisabet Ottosson for advice and  
566 information about fungal ecology, and Markus A.K. Sydenham for inspiration.

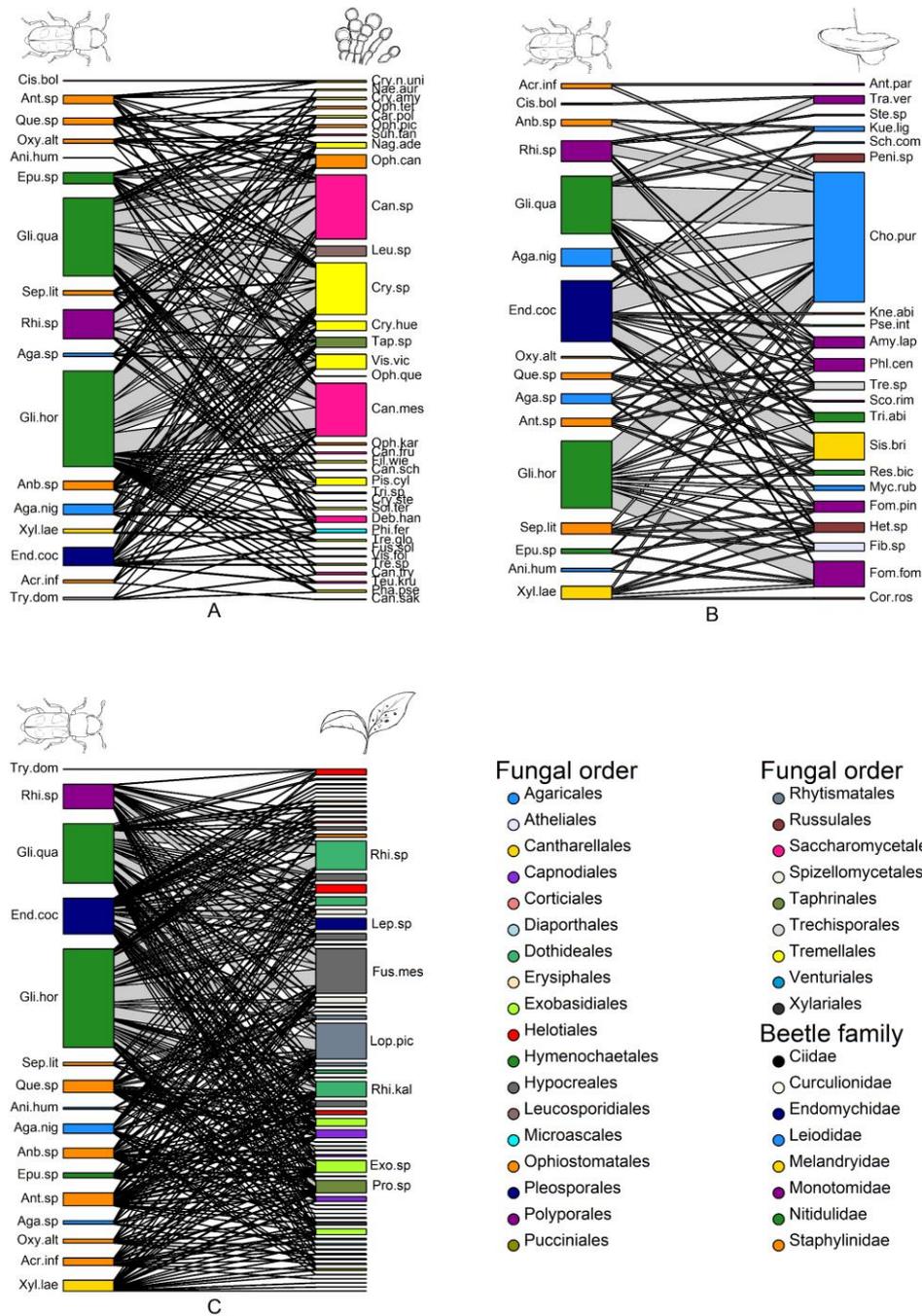
## 567 Author Contributions

568 RMJ, TB, HK and AST conceived the idea and designed the methodology. RMJ did the field  
569 work, lab work, analyses and led the writing of the manuscript. All authors contributed critically  
570 to the drafts and gave final approval for publication.

## 571 Additional Information

572 The authors declare no competing interests.

573 Figure Legends

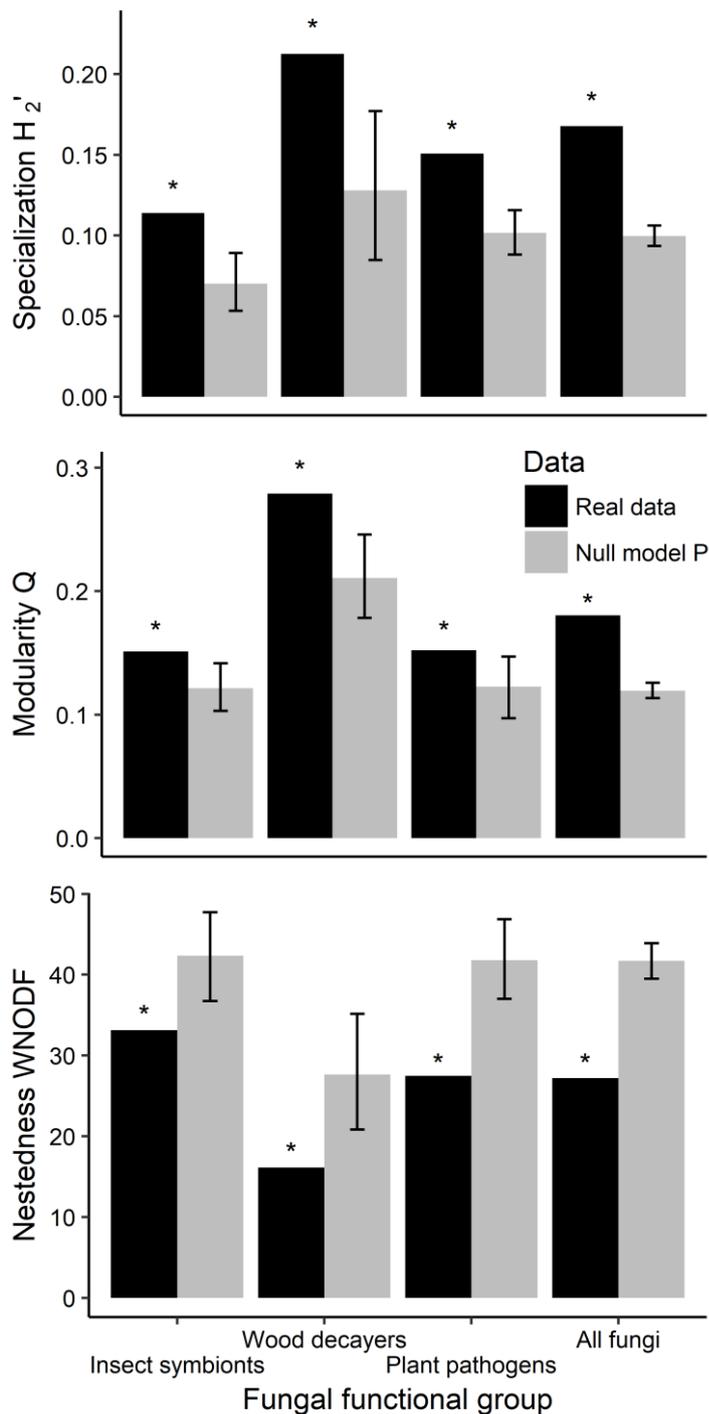


574

575 Figure 1) Networks of wood-inhabiting beetles and fungi classified as (A) insect symbionts,  
 576 (B) wood-decayers or (C) plant pathogens. Sizes of boxes and interaction lines represent  
 577 number of occurrences of the fungi in the insect samples. Colours denote taxonomic

578 grouping; order for fungi and family for insects (all in the order Coleoptera). See

579 Supplementary Tables S2-S7 for full names of abbreviations.



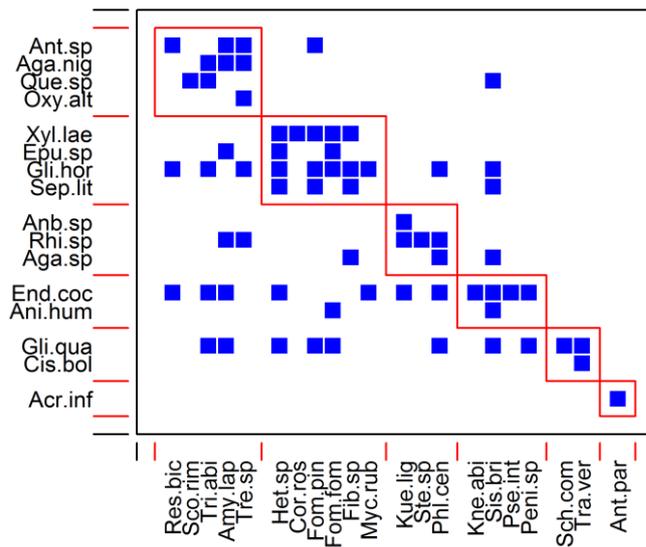
580

581 Figure 2) Network specialization, modularity and weighted nestedness for networks between

582 wood-inhabiting beetles and the fungal functional groups insect symbionts, wood-decayers

583 and plant pathogens, or all fungi annotated to species or genus. Black bars represent the

584 original networks, while grey bars represent networks randomized with constant marginal  
 585 sums according to null model P[50] with 95% confidence intervals (CI). Asterisks (\*) above  
 586 the black bars signify significant differences between the original and the randomized  
 587 networks.



588  
 589 Figure 3) Modules in the network between wood-inhabiting beetles and wood-decay fungi  
 590 with *C. purpureum* excluded, as organised by the QuanBiMo algorithm[26]. Lines demarcate  
 591 modules, squares indicate interactions between insects and fungi. See Supplementary Tables  
 592 S3 and S6 for full names of abbreviations.

593