How to build a mycelium: tradeoffs in fungal architectural traits 1 2 3 4 Anika Lehmann 1,2,*, Weishuang Zheng 3, Katharina Soutschek 1, Matthias C. 5 Rillig 1,2 6 7 1 Freie Universität Berlin, Institut für Biologie, Plant Ecology, Altensteinstr. 6, D-14195 8 Berlin, Germany; 9 2 Berlin-Brandenburg Institute of Advanced Biodiversity Research (BBIB), D-14195 Berlin, 10 Germany; 11 3 State key Laboratory of Microbial Technology, Shandong University, Qingdao 266237, 12 China 13 14 * Corresponding author, Freie Universität Berlin, Institut für Biologie, Plant Ecology, 15 Altensteinstr. 6, D-14195 Berlin, Germany. Tel.: +49 30 83853145. Fax: 49 30 83853886. 16 E-mail address: lehmann.anika@googlemail.com 17 18 Keywords: saprobic fungi, traits, tradeoff, mycelium, architecture

20 Abstract

- 21
- 22 The fungal mycelium represents the essence of the fungal lifestyle, and understanding how a
- 23 mycelium is constructed is of fundamental importance in fungal biology and ecology.
- 24 Previous studies have examined initial developmental patterns or focused on a few strains,
- 25 often mutants of model species, and frequently grown under non-harmonized growth
- 26 conditions; these factors currently collectively hamper systematic insights into rules of
- 27 mycelium architecture. To address this, we here use a broader suite of fungi (31 species
- 28 including members of the Ascomycota, Basidiomycota and Mucoromycotina), all isolated
- 29 from the same soil, and test for ten architectural traits under standardized laboratory
- 30 conditions.
- 31 We find great variability in traits among the saprobic fungal species, and detect several clear
- 32 tradeoffs in mycelial architecture, for example between internodal length and hyphal
- 33 diameter. Within the constraints so identified, we document otherwise great versatility in
- 34 mycelium architecture in this set of fungi, and there was no evidence of trait 'syndromes' as
- 35 might be expected.
- 36 Our results point to an important dimension of fungal properties with likely consequences for
- 37 coexistence within local communities, as well as for functional complementarity (e.g.
- 38 decomposition, soil aggregation).
- 39

41 Introduction

- 42
- 43 The mycelium comprises the entirety of the hyphae of a fungus, representing its nutrient-
- 44 capture and interaction interface, and the infrastructure for transport within the fungal
- 45 individual. This structure is designed for a dynamic exploratory lifestyle with its ability to
- 46 reconfigure, fragment and fuse, and represents the very essence of the fungal lifestyle¹.
- 47 Understanding how a mycelium is built is therefore of fundamental importance for gaining
- 48 insight into fungal biology and ecology.
- 49 The initial development of the mycelium starting from germinating spores has been
- 50 extensively studied (e.g.^{2,3}), revealing some general hyphal growth patterns: emerging from a
- 51 spore, a hypha extends at an exponential rate followed by a constant linear phase until the
- 52 formation of a new branch is initiated; each new branch itself follows this exponential-linear
- 53 phase pattern. Additionally, hyphae show negative autotropism and radial orientation away
- 54 from the colony center³, eventually giving rise to the characteristic circular (in 2D) or
- spherical (in 3D) shape of 'colonies' or fungal individuals.
- 56 Mycologists have also examined the kinetics and branching behavior of fungi with a focus on
- a limited suite of traits and with a typical focus on mutants of the same species or a few
- 58 species, mainly derived from fungal culture collections (e.g.⁴⁻⁶). These investigations revealed
- 59 that fungal mycelia undergo changes in growth behavior due to differentiation. Studies on
- 60 Neurospora crassa revealed that after approximately 22 h branching angles decrease while
- 61 hyphal extension rate and diameters increase. Ultimately, the mycelium establishes a
- 62 hierarchy in which hyphae of higher branching order have decreased hyphal growth rate and
- 63 diameter in relation to the parental hyphae from which they emerged. As a consequence, the
- 64 space-filling capacity of the mycelium increases⁷, leading to a maximum surface area while
- 65 investing in a minimum of hyphal length⁸.
- 66

67 Apart from such studies of development, there is no systematic comparison of a range of 68 architectural features, measured under the same, standardized laboratory conditions, on a 69 larger set of fungi from a common ecological context. This is why we currently only have 70 limited knowledge about tradeoffs governing mycelium architecture that could give insight on 71 architectural 'rules'. So far, insights into such tradeoffs come from studying single species or 72 mutants, and this frequently has not resulted in a consensus. For example, the relationship 73 between hyphal diameter and growth rate can be positive (e.g. Bortrytis) or neutral (e.g., 74 *Mucor* strains)⁹. The same discrepancy holds true for the relationship between hyphal 75 branching frequency (number of hyphal tips) and hyphal growth rate, which was examined in 76 multiple studies focusing on *Neurospora* strains and mutants: some studies found a positive

relationship¹⁰ and others did not¹¹.

78 Recently, advantages of pursuing a trait-based approach in fungal ecology have been

- introduced^{12,13}. One clear benefit of such an approach is to move beyond idiosyncratic
- 80 comparisons of a few isolates to making comparisons using larger sets of fungal isolates,
- 81 offering opportunities for more general inferences about mycelium architectural rules.
- 82 Here, we report on studies designed to collect mycelium architecture traits for a set of 31
- 83 saprobic fungal strains, containing members of the phyla Ascomycota, Basidiomycota and
- 84 Mucoromycotina. Our goal was to uncover general 'rules' of mycelium construction by
- 85 identifying tradeoffs among mycelium growth characteristics within this set of fungi, all
- 86 isolated from the same soil.
- 87

88 Materials and Methods

Fungal strains. Fungal strains were originally cultured from Mallnow Lebus, a dry grassland 89 90 in a nature conservation reserve (Brandenburg, Germany, 52° 27.778' N, 14° 29.349' E). A 91 set of 31 fungal strains were isolated from soil samples as described in Andrade-Linares et 92 al.¹⁴. Briefly, soils were diluted or washed to minimize spore abundance and increase the isolation of fungi derived from hyphae attached to soil particles^{15,16}. For isolation a variety of 93 94 media and antibiotics were used to target Ascomycota, Basidiomycota and Mucoromycotina 95 while suppressing bacterial growth. Isolates were incubated at 22°C and were cultured on 96 PDA. The fungal set comprised members of the Ascomycota (twenty strains), Basidiomycota 97 (four strains) and Mucoromycotina (seven strains) (Fig.1a, Table S1).

98

99 Architecture traits. We conducted two separate studies to collect architectural traits for the 100 31 fungal strains. All studies were performed in vitro with PDA as growth substrate. In the 101 first study, we focused on measuring hyphal branching angle (BA), internodal length (IL) and 102 diameter (D). For this, the fungal strains were grown on single concavity slides carrying 150 103 µl of PDA. We chose to reduce the concentration of PDA to 10% to obtain nutrient reduced 104 growth medium for reduced mycelial density. This was necessary to be able to identify single 105 hyphae in very densely growing fungi. To guarantee solidification of the medium, we added 106 agar (Panreac AppliChem) to reach 15 g L⁻¹ concentration. The growth medium was flattened 107 by placing a cover slip on the liquid medium drop until it solidified. A pre-sterilized poppy 108 seed carrying the target fungal strain was positioned in the center of the concavity. The slide 109 was placed in a 9 cm Petri dish filled with a 5 mm layer of water agar to maintain high air 110 humidity. Plates were sealed and stored at room temperature (22 °C) in the dark until the

111 fungal colony covered half of the concavity area. For each fungal strain five slides were 112 prepared and placed in separate Petri dishes. For the measurements, slides were examined 113 under the microscope (Leica DM2500, bright field, 200x). Per slide, we randomly chose five 114 hyphae as subsamples; for each of these hyphae we measured at the colony edge the last 115 developed branching angle, the internodal lengths between this last and second-to-last 116 branch and the hyphal diameters within this youngest internodal segment. For analyses, we 117 used the image processing software ImageJ¹⁷. For each experimental unit, we calculated a 118 mean value and a coefficient of variation (CV) from the subsample data. These represented 119 two aspects of a trait: the average value and its variability. The trait data used in statistical 120 analyses were the average of mean values and CVs of five replicates (i.e. n = 31).

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122 In the second study, we investigated the complexity and the heterogeneity of fungal mycelia 123 by applying fractal analysis - a technique used to assess self-similarity and space-filling capacity of fungal hyphae¹⁸. For this, we applied the same approach as in the first 124 125 experiment but with eight replicates per fungal strain. At harvest, the slides were examined 126 under the microscope (Leica DM2500, bright field, 200x) focusing on the outer 200 µm of the growing zone to investigate the "surface fractals"¹⁹. Camera (Leica DFC290) settings were 127 128 chosen to generate grayscale photos with high contrast (background: white, hyphae: black; 129 Fig.1b). For each slide, we photographed three fields of view at the colony edge. These settings and further image processing in ImageJ¹⁷ and Adobe Illustrator (CS6, v.16.0.0) were 130 131 necessary to guarantee comparable and unbiased photos that can be processed by image 132 analysis software. First, photos were converted to 8-bit binary images in ImageJ and 133 subsequently hyphae were skeletonized. For this, a thinning algorithm repeatedly reduced 134 pixels from the edge of the target object until a one-pixel wide shape was reached²⁰. In 135 Illustrator, the skeletonized hyphae were reconnected and image artifacts excluded, if 136 necessary. Line thickness was adjusted to mean diameter trait values derived from experiment one. The final processed images were loaded into the ImageJ plug-in "FracLac"21 137 138 to measure fractal dimensions. We chose box counting dimensions (Db) as a measure of 139 structural complexity (i.e. the degree of detail or amount of parts a pattern consists of), and 140 lacunarity (L) as a representative of structural heterogeneity (i.e. the gappiness or "rotational and translational invariance" in a pattern²¹). We used default settings but allowed for 141 142 rotational orientations in analyses. Finally, subsample data were used to calculate CVs for 143 box counting dimension and lacunarity. The subsample data were then merged to one mean 144 and CV trait value per replicate. Additionally, we verified if implementing diameter data 145 altered fractal dimension data by correlating skeletonized and adjusted diameter data for 146 both box counting dimension and lacunarity (Fig. S1). 147

148 Statistics. We analyzed the relationships between the ten architectural trait variables 149 derived from 31 saprobic fungal strains represented by both mean value and coefficients of 150 variation (CV) (n = 31). First, to evaluate fungal distribution in ten-dimensional trait space, we 151 ran a principal component analysis using the function prcomp() in the package "stats" with ztransformed data. Significance of PC axes was determined via the function testdim()²² in the 152 153 package "ade4"²³⁻²⁵. Only the first axis was significant, hence we included PC axis 1 and 2 in 154 the visualizations without losing information from the excluded axes. Next, we conducted 155 kernel density estimation to assess species occurrence probability following the procedure presented by Diaz et al.²⁶. Briefly, we used kde() function of the "ks" package²⁷ with 156 157 unconstrained bandwidth selectors by implementing the function Hpi() on our first two PC 158 axes. Using the function contourLevels() we estimated contour probabilities for 0.5 and 0.95 159 quantiles. 160 Second, to test for phylogenetic signal in our trait variables we used Moran's I statistic, a measure for phylogenetic autocorrelation, as implemented in the package "phylosignal". We 161 162 accounted for phylogenetic relatedness among species (indicated by detected phylogenetic 163 signals) by calculating phylogenetically independent contrast of our trait variables with the packages "picante"²⁸ and "ape"²⁹ using the functions pic() and match.phylo.data(). We 164 evaluated if the assumptions of the Brownian motion model were satisfied by our data³⁰. For 165 166 that, we investigated the standardization of the contrasts via diagnostic regression tests to 167 evaluate the relationship between absolute standardized contrasts and (i) the square root of their standard deviation³¹ and (ii) the node height (i.e. node age^{32,33}). Identified influential 168 169 nodes were excluded, following the threshold of absolute studentized residuals greater than 3^{31,34}. To satisfy the Brownian motion model assumption, we used log transformed trait 170 values und excluded two outlier nodes (node 49 and 61³⁵). 171 Third, multiple pairwise correlations using Pearson's rho were conducted and plotted with the 172 function corrplot() in the eponymous package³⁶. Analyses were done for original (non-173 174 transformed, n= 31) and phylogenetically corrected data (log-transformed, n= 28). 175 Fourth, we ran linear regressions and further investigated the relationships by quantile 176 regression with the package "quantreg" (https://github.com/cran/quantreg). Under most 177 ecological conditions, linear regressions tend to over- or underestimate relationships due to a 178 focus on the mean of the response distribution. Especially in wedge-shaped data 179 distributions, indicating unmeasured limiting factors, quantile regressions are more 180 informative since they test the relationship between response and predictor variable at their maxima^{37,38}. Both regression analyses were run on z-transformed data and model residuals 181 182 were tested for homogeneity and normal distribution. Additionally, we ran multiple pairwise

regressions for both original and phylogenetically corrected data to provide graphical

184 information on data distributions of all trait combinations (see Fig. S2 and S3). These were

185 generated by the function ggpairs() of the package GGally³⁹.

All analyses were conducted in R (v. 3.4.1⁴⁰) and plots were created with the graphics

187 package ggplot2⁴¹ and its extension GGally.

188

189 Results and discussion

190 Trait expression. Overall, we found high variability among strains for all traits (Fig.1). The 191 application of fractal dimensions on mycelium structure revealed that trait mean values of 192 box counting dimensions (Db) ranged between 1.2 and 1.6, where a value of 1 represents a 193 single unbranched hypha, and a value of 2 a complex space-filling mycelium. The most 194 complex mycelium was found in the Mucoromycotina, while Basidiomycota had the most 195 simply structured mycelia. For lacunarity (L), we found in our study that trait values ranged 196 between 0.4 (Basidiomycota) and 0.7 (Ascomycota). With increasing trait value, the 197 heterogeneity and hence gappiness of the mycelium increased. The investigation of hyphal 198 features revealed that the branching angle (BA) varied substantially across fungal strains 199 from 26 to 86° with Mucoromycotina having large angles and Basidiomycota rather small 200 angles. For hyphal diameter (D) trait values ranged from 2.7 to 6.5 µm across the 31 strains 201 where both extremes could be found in the Mucoromycotina. The length of the hyphal 202 internodes (IL) showed considerable differences: Within Basidiomycota internodal lengths of 203 453 µm could be reached while in Mucoromycotina the lowest value of 40 µm was measured. 204 Our values are within the range of previously reported architectural features of selected, individual saprobic filamentous fungi (e.g.^{19,42-46}). 205 206 After establishing the trait database, we investigated the trait space generated by the

207 collected fungal architectural features. To do this, we applied principal component analyses.

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209 PCA. For our 31 fungal strains, the sole significant first principal components accounted for 210 34% of the variability in the ten architecture traits (Fig. 2a). In this ten-dimensional trait 211 space, the set of our 31 fungal strains occupied the whole PC plane with a clear separation 212 of the three phyla across the plane. Considering the sole significant PC axis 1, Ascomycota 213 assembled in the center flanked by Mucoromycotina on the left, driven by large branching 214 angles and high mycelial complexity, and Basidiomycota on the right portion, primarily 215 characterized by long internodes and wide hyphal diameters (Fig. S4). Across species, some 216 clear correlations among traits became visible; hence, we further investigated the type and 217 intensity of potential architectural tradeoffs for our fungal set.

219 **Phylogenetic signal.** For this, we first tested all ten traits for a phylogenetic signal to

- 220 evaluate if the phylogenetic relatedness among fungal strains can influence any trait
- relationships we want to investigate. Applying Moran's I statistics (Table S2), we found
- phylogenetic signals in Db (I= 0.16, p= 0.02) and L_{cv} (I= 0.13, p= 0.02). Hence, we needed to
- account for phylogenetic relations for these two traits among our 31 fungal strains by
- applying phylogenetically independent contrast in the following analyses.
- 225
- 226 **Tradeoffs.** We found 14 trait pairs with significant correlations of which ten passed
- 227 phylogenetic correction (Fig. 2b, Fig.S2 and S3). The strongest correlations were detected
- between mycelium complexity and its coefficient of variation (Db Db_{cv} in **Fig.2c**), mycelium
- heterogeneity, as measured by lacunarity, (Db L in Fig.2d) and hyphal diameter (Db D in
- 230 Fig.2e), as well as between branching angle and its coefficient of variation (BA BA_{cv} in
- 231 Fig.2f). For internodal length, we detected relationships with mycelium complexity (IL Db in
- **Fig.2g**), variability in mycelium complexity (IL Db_{cv} in Fig.**2g**) and hyphal diameter (IL D in
- 233 Fig. 2i). Another strong correlation was found between the coefficients of variation of
- mycelium complexity and branching angle (Db_{cv} BA_{cv} in **Fig.2j**). In addition, weak
- 235 correlations were found for D and Db_{cv} , D and L_{cv} , as well as between IL and L_{cv} , IL and BA
- (Fig.S2 and S3). From these correlations we can deduce multiple rules for myceliumarchitecture.
- 238 For structural complexity (as represented by box counting dimensions) and branching angle
- 239 we detected a negative relationship between their mean values and CVs (Db Db_{cv} in Fig.
- 240 **2c** and BA BA_{cv} in **Fig. 2f**). Thus strains exhibiting a high trait value for BA or Db are
- restricted to this high value, while strains with low values in these traits are capable of furtheradjusting these features.
- 243 Within strains, variability in mycelial complexity itself is determined by increasing internodal
- length (IL Db_{cv} in **Fig. 2h**) and higher flexibility in branching angle measures (Db_{cv} BA_{cv} in
- **Fig. 2j**). Thus, the degree of mycelial complexity can be modulated via branching patterns
- 246 (e.g. distance between branches).
- 247 Considering the complexity – the space-filling capacity – of a mycelium, we found that more 248 complex mycelia are more heterogeneously structured (Db - L in Fig.2d). Mycelia with high 249 space-filling capacity tend to be rather heterogeneous in their structure, i.e. their mycelium is 250 not uniformly complex but rather exhibits complex zones replaced by more simple mycelium 251 structures towards the growing edge. At the colony edge, hyphae are confronted with new 252 resources and environmental conditions for which a maximum of flexibility is likely 253 advantageous. Furthermore, complex mycelia have smaller hyphal diameters (Db - D in Fig. 254 2e) and shorter internodal length (IL - Db in Fig.2g). A mycelium with long internodes is
- characterized by less branching and hence less space-filling. However, to be capable of

growing long internodes the mycelium needs to improve its structural support, i.e. its tearresistance. Long hyphae are at risk of fragmentation by shear-stresses⁴⁷. To deal with this risk, hyphal cell walls can thicken and/or hyphal diameter can increase^{5,48}. This is congruent with our finding that long internodes are linked with larger hyphal diameters (IL - D in **Fig.2i**).

- 261 **Conclusion.** One of the most fundamental decisions a growing hypha has to make is when 262 to branch. Thus, it is maybe not surprising that internodal length was a highly influential 263 variable (aligned with PC axis 1, Fig. 2) in understanding the architecture of mycelia in trait 264 space. This suggests that the trait internodal length is a main driver of mycelium architecture. 265 Mycelia with short internodes can branch more frequently thus developing a more complex 266 mycelium than those with long internodes. However, the capability of growing long 267 unbranched hyphae enables the mycelium to more flexibly adjust their mycelial modules (see 268 positive correlations between IL and Db_{cv} , L_{cv}) in response to environmental conditions.
- 269

270 It is interesting that there were no sharp boundaries in the sense of architectural 'syndromes' 271 or clear groups of traits, but rather relatively gradual changes in trait values within the set of 272 fungal isolates we examined. This illustrates the relative versatility of the mycelial growth 273 form in evolutionary terms, at least in the peripheral growth zone of the fungus, which we 274 examined here. We clearly show that there are limits to how a mycelium can be constructed, 275 since some trait combinations are evidently not favorable (e.g. long internodes and small 276 diameters). However, fungi have evidently otherwise filled the trait space within the 277 constraints of such fundamental tradeoffs, even seen in a sample of 31 species. It will be 278 interesting to compare our results to others sets of fungi once such data are available: our 279 results suggest key parameters on which to focus.

280

281 Mycelial architecture is a fundamental property of filamentous fungi, governing the way these 282 organisms explore their substrate. Using a set of fungi co-occurring in the same soil, we 283 show that architectural features vary strongly and reproducibly among different isolates 284 under the same laboratory conditions. It is therefore highly likely that such differences 285 contribute to enabling coexistence within fungal communities⁴⁹ by offering fungi different 286 ways to forage and colonize the soil environment. On the other hand, such trait divergence 287 can also mediate functional complementarity, for example in decomposition or soil 288 aggregation¹³. 289

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292 References

- Wessels, J. G. H. Fungi in their own right. *Fungal Genetics and Biology* 27, 134-145 (1999).
- Plomley, N. J. B. Formation of the colony in the fungus Chaetomium. *Australian Journal of Biological Science* 12, 53-64 (1959).
- 2973.Trinci, A. P. J. A study of the kinetics of hyphal extension and branch initiation of298fungal mycelia Journal of Genetic Microbiology 81, 225-236 (1974).
- Butler, G. M. Growth of hyphal branching systems in *Coprinus disseminatus*. *Annals of Botany* 25, 341-& (1961).
- McLean, K. M. & Prosser, J. I. Development of vegetative mycelium during colony
 growth of *Neurospora crassa*. *Transactions of the British Mycological Society* 88, 489 495 (1987).
- Robinson, P. M. & Smith, J. M. Apical branch formation and cyclic development in
 Geotrichum candidum. Transactions of the British Mycological Society **75**, 233-238
 (1980).
- 3077.Barry, D. J. Quantifying the branching frequency of virtual filamentous microbes using
fractal analysis. *Biotechnology and Bioengineering* **110**, 437-447 (2013).
- Moore, D., Robson, G. D. & Trinci, A. P. J. 97-98 (Cambridge University Press, Cambridge, UK, 2011).
- 311 9. Carlile, M. J., Watkinson, S. C. & Gooday, G. W. (Academic Press, UK, 2001).
- 10. Katz, D., Goldstein, D. & Rosenberger, R. F. Model for branch initiation in *Aspergillus nidulans* based on measurements of growth parameters. *Journal of Bacteriology* **109**, 1097-1100 (1972).
- Watters, M. K., Lindamood, E., Meunich, M. & Vetor, R. Strain-dependent
 eelationship between growth rate and hyphal branching in *Neurospora crassa*. *Proceedings of the Indiana Academy of Science* **117**, 1-6 (2008).
- 31812.Aguilar-Trigueros, C. A. *et al.* Branching out: Towards a trait-based understanding of319fungal ecology. *Fungal Biology Reviews* **29**, 34-41 (2015).
- Lehmann, A. & Rillig, M. C. Understanding mechanisms of soil biota involvement in soil aggregation: A way forward with saprobic fungi? *Soil Biology & Biochemistry* 88, 298-302 (2015).
- 32314.Andrade-Linares, D. R., Veresoglou, S. D. & Rillig, M. C. Temperature priming and
memory in soil filamentous fungi. *Fungal Ecology* **21**, 10-15 (2016).
- 325 15. Gams, W. & Domsch, K. H. Beitrage zur Anwendung der Bodenwaschtechnik für die
 326 Isolierung von Bodenpilzen. *Arch. Mikrobiol.* 58, 134-144 (1967).
- Thorn, R. G., Reddy, C. A., Harris, D. & Paul, E. A. Isolation of saprophytic
 basidiomycetes from soil. *Applied and Environmental Microbiology* 62, 4288-4292
 (1996).
- 330 17. Schneider, C. A., Rasband, W. S. & Eliceiri, K. W. NIH Image to ImageJ: 25 years of
 331 image analysis. *Nature Methods* 9, 671-675 (2012).
- 33218.Posser, J. I. in *The growing fungus* (eds N.A.R. Gow & G.M. Gadd) (Springer, UK,3331995).
- 33419.Obert, M., Pfeifer, P. & Sernetz, M. Microbial growth patterns described by fractal335geometry. Journal of Bacteriology 172, 1180-1185 (1990).
- Zhang, T. Y. & Suen, C. Y. A fast parallel algorithm for thinning digital patterns.
 Communications of the Acm 27, 236-239 (1984).
- 338 21. Karperien, A. FracLac for ImageJ (1999-2013).
- Dray, S. On the number of principal components: A test of dimensionality based on measurements of similarity between matrices. *Computational Statistics & Data Analysis* 52, 2228-2237 (2008).
- 342 23. Chessel, D., Dufour, A. B. & Thioulouse, J. The ade4 package I : One-table
 343 methods. *R News* 4, 5-10 (2004).
- 34424.Dray, S. & Dufour, A. B. The ade4 package: Implementing the duality diagram for
ecologists. *Journal of Statistical Software* 22, 1-20 (2007).

346	25.	Dray, S., Dufour, A. B. & Chessel, D. The ade4 package-II: Two-table and K-table
347		methods. <i>R News</i> 7 , 47-52 (2007).
348	26.	Diaz, S. et al. The global spectrum of plant form and function. Nature 529, 167-U173
349		(2016).
350	27.	Duong, T. ks: Kernel smoothing v. R package version 1.11.0. (2018).
351	28.	Kembel, S. W. et al. Picante: R tools for integrating phylogenies and ecology.
352		Bioinformatics 26, 1463-1464 (2010).
353	29.	Paradis, E., Claude, J. & Strimmer, K. APE: Analyses of Phylogenetics and Evolution
354	_0.	in R language. <i>Bioinformatics</i> 20 , 289-290 (2004).
355	30.	Cooper, N., Thomas, G. H. & FitzJohn, R. G. Shedding light on the "dark side' of
356	00.	phylogenetic comparative methods. <i>Methods in Ecology and Evolution</i> 7 , 693-699
357		(2016).
358	31.	Garland, T., Harvey, P. H. & Ives, A. R. Procedures for the analysis of comparative
359	51.	data using phylogenetically indpendent contrasts. Systematic Biology 41 , 18-32
360		(1992).
361	32.	Freckleton, R. P. & Harvey, P. H. Detecting non-Brownian trait evolution in adaptive
362	52.	radiations. Plos Biology 4, 2104-2111 (2006).
363	33.	Grafen, A. The phylogenetic regression. <i>Philosophical Transactions of the Royal</i>
364	55.	Society of London Series B-Biological Sciences 326 , 119-157 (1989).
	24	
365	34.	Jones, K. E. & Purvis, A. An optimum body size for mammals? Comparative evidence
366	25	from bats. <i>Functional Ecology</i> 11 , 751-756 (1997).
367	35.	Orme, D. The caper package: comparative analysis of phylogenetics and evolution in
368	00	R v. 1.0.1 (2013).
369	36.	Wei, T. & Simko, V. R package "corrplot": Visualization of a Correlation Matrix v. 0.84
370	07	
371	37.	Cade, B. S. & Noon, B. R. A gentle introduction to quantile regression for ecologists.
372		Frontiers in Ecology and the Environment 1, 412-420 (2003).
373	38.	Cade, B. S., Terrell, J. W. & Schroeder, R. L. Estimating effects of limiting factors with
374	~~	regression quantiles. <i>Ecology</i> 80 , 311-323 (1999).
375	39.	Schloerke, B. et al. GGally: Extension to 'ggplot2' v. 1.3.2 (2017).
376	40.	R Development Core Team. R: A language and environment for statistical computing
377		v. 3.4.1 (2014).
378	41.	Wickham, H. ggplot2: Elegant graphics for data analysis. (Springer, 2009).
379	42.	Henis, Y., Okon, Y. & Chet, I. Relationship between early hyphal branching and
380		formation of sclerotia in Sclerotium rolfsii. Journal of General Microbiology 79, 147-
381		150 (1973).
382	43.	Ho, H. H. Hyphal branching systems in Phytophthora and other Phcomycetes.
383		Mycopathologia 64 , 83-86 (1978).
384	44.	Hutchinson, S. A., Sharma, P., Clarke, K. R. & Macdonald, I. Control of hyphal
385		orientation in colonies of Mucor hiemalis. Transactions of the British Mycological
386		Society 75 , 177-191 (1980).
387	45.	Kotov, V., Anishchenko, I., Sirenko, I. & Reshetnikov, S. Statistical analysis of
388		structural and kinetic characteristics of fungal colony growth with Trichoderma viride
389		Pers.: S.F. Gray. <i>Microbiological Research</i> 160, 273-278 (2005).
390	46.	Trinci, A. P. J. A kinetic study of the growth of Aspergillus nidulans and other fungi.
391		Journal of Genetic Microbiology 57 , 11-24 (1969).
392	47.	Suijdam, J. C. & van Metz, B. Fungal pellet breakup as a function of shear in a
393		fermentor. Journal of fermentation technology 59, 329-333 (1981).
394	48.	Trinci, A. P. J. & Collinge, A. J. Hyphal wall growth in Neurospora crassa and
395		Geotrichum candidum. Journal of General Microbiology 91, 355-361 (1975).
396	49.	Crowther, T. W. et al. Untangling the fungal niche: the trait-based approach. Frontiers
397		in Microbiology 5 (2014).
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404 Conflicts of interest

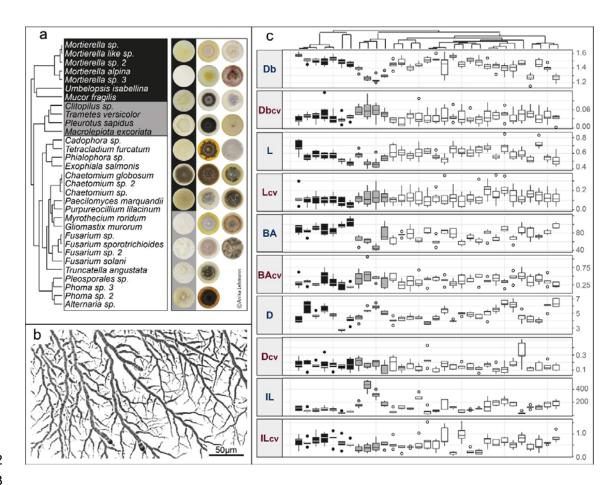
- 405 The authors declare no conflicts of interest.
- 406

407 Figure captions

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409 Fig. 1 Overview of fungal strains, mycelial image and trait distribution. (a) Phylogenetic tree of 410 the 31 fungal strains belonging to the phyla Ascomycota, Basidiomycota and Mucoromycotina with 411 colony pictures. Pictures are from four-week old cultures grown on potato dextrose agar, and their 412 order follows the order of the phylogenetic tree. Further information about phylogeny and accession 413 numbers of the 31 strains are available in Table S1. (b) An example of mycelial pictures obtained from 414 the setup introduced in the material and method section. Imaged strain is Mucor fragilis. (c) Tukey 415 boxplot of the ten architectural trait variables, mean value and their coefficient of variation (CV) 416 measured in this study: box counting dimension (unitless, Db with n= 8, Db_{cv} with n= 8), lacunarity 417 (unitless, L with n= 8, L_{cv} with n=8), branching angle (in °, BA with n= 5, BA_{cv} with n= 5), hyphal 418 diameter (in μ m, D with n= 5, D_{CV} with n= 5), internodal length (in μ m, IL with n= 5, IL_{CV} with n= 5). The 419 boxplots represent 25th and 75th percentile, median and outlying points. Information about phylum 420 affiliation is color-coded (black: Mucoromycotina, grey: Basidiomycota, white: Ascomycota).

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426 Fig. 2 Outcomes of principal component analysis, trait correlation, linear and quantile

427 regression of the investigated architectural traits. Analyses were conducted on trait data (n= 31).

428 (a) Projection of the ordinated 31 fungal strains onto ten architectural trait variables (mean and CV):

429 box counting dimension (Db), lacunarity (L), branching angle (BA), hyphal diameter (D), internodal

430 length (IL) into two-dimensional trait space represented by principal component axis 1 and 2

431 (explaining 34 and 18% of variance, respectively). Arrows indicate direction and weight of trait vectors.

432 Color gradient represents probability of species occurrence (white = low, red = high) in the trait space,

433 with the contour lines denoting the 0.50 and 0.95 quantiles of kernel density estimation (see materials

434 and methods section). (b) Correlation plot of five architectural trait variables and their coefficients of

435 variation. The upper triangle displays original while the lower triangle represents phylogenetically

436 corrected data correlations. Color gradient and square size are proportional to correlation coefficient

437 (Pearson's rho). Asterisks denote significance level: *** < 0.001, ** <0.01, * < 0.05. In grey, we

438 highlight trait combinations affected by detected phylogenetic signal (Table S2). (c-j) The eight

439 strongest trait relationships for either original and/or phylogenetically corrected data. Red lines

440 represent linear regression lines and blue lines quantile regression lines, while line type depicts

441 significance of regression lines; solid lines p-value < 0.05, dashed lines > 0.05. Corresponding

regression statistics can be found in Table S4. Adjusted R² values correspond to linear regressions.

