

1 **Sample size shapes metabarcoding-driven biodiversity assessments across body sizes**
2 **in soil**

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11

12 **Abstract**

13 Understanding how sample size influences biodiversity detection across taxonomic
14 groups differing in body size is critical for designing robust and cost-efficient
15 metabarcoding studies of soil eukaryotes. Using a soil mass gradient (0.25-32 g)
16 combined with a universal 18S rRNA metabarcoding approach, we quantified how
17 sample mass shapes diversity estimates across eukaryotic taxa. Diversity metrics
18 (richness and inverse Simpson diversity; $q = 0, 2$) and community dispersion exhibited
19 clear body size-dependent responses. Larger-bodied taxa (e.g., *Nematoda*, *Collembola*,
20 *Insecta*) showed pronounced increases in detected diversity and reduced community
21 dispersion with increasing soil mass, indicating that small soil samples fail to capture
22 their full diversity. Conversely, microeukaryotic groups such as fungi and protists
23 displayed weak or even negative relationships with increasing soil mass, implying limited
24 improvement in detection with increased sampling effort. These findings demonstrate

25 that soil sample sizes are strongly modulated by organismal body size and spatial
26 distribution. We propose that a taxon-specific analytical framework can enhance both the
27 ecological representativeness and cost efficiency in metabarcoding-based soil
28 biodiversity assessments and monitoring.

29 **Introduction**

30 Soil eukaryotes encompass a diverse array of organisms, including fungi, protists, nematodes,
31 arthropods, and various other micro- to macro-faunal groups that collectively drive key
32 ecosystem functions such as nutrient cycling, organic matter decomposition, and soil structure
33 formation (Blouin et al., 2013; Eisenhauer et al., 2026; Geisen et al., 2018; Lee and Foster,
34 1991; Lehmann et al., 2017). Information about the diversity of soil biota provides critical
35 insights into the complexity of ecosystem functions and trophic networks, and supports a
36 mechanistic understanding of ecosystem resistance and resilience under environmental change
37 (Phillips et al., 2024).

38 The advent of high-throughput DNA metabarcoding has opened new avenues to characterize
39 these complex communities, offering insights into both their taxonomic and functional diversity
40 across different spatial and temporal scales (Aslani et al., 2022; De Gruyter et al., 2020;
41 Köninger et al., 2023; Seppey et al., 2017), and allowing for the simultaneous analysis of most
42 groups of soil biota. Yet, methodological factors, especially aspects of sampling design, continue
43 to shape the accuracy and representativeness of such biodiversity surveys (Jurburg et al., 2021;
44 Lara et al., 2022; Liu et al., 2017). Among these, sample size (i.e., soil mass) is a critical
45 determinant of detected diversity patterns, largely due to the heterogeneous and patchy
46 distribution of different groups of soil biota (Bahram et al., 2016; Caruso et al., 2012; Ettema and
47 Wardle, 2002; Liu et al., 2019; Quist et al., 2019). Smaller soil sample masses may fail to
48 capture rare or spatially aggregated taxa, which can skew community profiles and lead to
49 underestimates of overall biodiversity, while increasing compositional variability between
50 samples (Dopheide et al., 2019; Kageyama and Toju, 2022; Nascimento et al., 2018; Wiesel et
51 al., 2015).

52 Importantly, the effect of sample mass on the detectability of soil biota could be affected by the
53 target organism's body size, which is a phylogenetically conserved trait that reflects an

54 organism's abundance, life history, metabolic demands, and trophic role (Brown et al., 2004;
55 Speakman, 2005; Woodward et al., 2005). By influencing resource use and energy flow, body
56 size plays a central role in the structure and stability of food webs and ecosystems (Brown and
57 Maurer, 1986; De Bie et al., 2012; Sheridan and Bickford, 2011; Woodward et al., 2005).
58 Because larger-bodied eukaryotes tend to occur at lower densities and are more spatially
59 patchy, increasing sample mass should disproportionately increase the likelihood of
60 encountering these taxa. Consequently, alpha diversity is expected to rise more steeply with
61 sample mass for larger-bodied groups than for smaller ones. Beyond effects on alpha diversity,
62 spatial aggregation also influences variation in beta diversity, as smaller sample masses amplify
63 among-replicate variability in community composition, or dispersion. We therefore expect
64 increasing sample mass to reduce compositional dispersion most strongly for large-bodied taxa,
65 whereas small-bodied taxa should remain comparatively heterogeneous. Nevertheless, despite
66 these mechanistic expectations, the extent to which body size modulates sample-mass effects
67 on both alpha diversity and compositional dispersion remains poorly understood in soil
68 eukaryote metabarcoding studies. Clarifying this relationship is key to refining sampling
69 strategies and ensuring that biodiversity assessments are both accurate and comparable across
70 different studies and ecological contexts.

71 In this study, we employed a DNA metabarcoding approach to systematically explore how
72 varying sample masses influence the detection and characterization of soil eukaryotic taxa
73 across multiple body sizes. We hypothesized that (a) alpha diversity responses to soil mass
74 increase with body size, (b) sample dispersion declines with soil mass, particularly in larger-
75 bodied taxa. By examining taxonomic groups spanning the body size spectrum from 10 to
76 20000 μm , we aimed to uncover mass-dependent sampling effects and provide practical,
77 empirically grounded guidance for designing future soil biodiversity surveys.

78 **Methods and materials**

79 **Study site and sampling design**

80 Sampling was conducted at the PhytOakmeter grassland field site (Bad Lauchstädt, Germany,
81 51°23'30"N, 11°52'49"E, 118 m a.s.l.), characterized by a Haplic Chernozem soil and a sub-
82 continental, temperate climate. To isolate the technical effects of sample mass from broad-scale
83 environmental heterogeneity (e.g., inter-tree variability), we established five independent
84 sampling plots around a 2-m radius of a *Quercus robur*. From each plot, 3-5 5 cm soil cores
85 were pooled, and thoroughly homogenized with a 2 mm sieve to create five composite soil
86 replicates (~1 kg each). Subsequently, subsamples of each composite were weighed to create a
87 soil mass gradient consisting of eight levels (0.25, 0.5, 1, 2, 4, 8, 16, and 32 g) for each of the
88 five replicates, resulting in a total of 40 samples.

89 **DNA extraction, PCR amplification, and Illumina sequencing**

90 DNA was extracted from all 40 samples using FastDNA™ SPIN Kit for Soil, 50 mL tubes (MP
91 Biomedicals, USA), according to the manufacturer's instructions. Importantly, to accommodate
92 samples exceeding the kit's capacity (max. 10 g), we implemented a split-and-pool strategy: for
93 16 g and 32 g samples, soil was split into 8 g aliquots (two and four aliquots, respectively),
94 extracted in parallel, and the resulting lysates were pooled prior to downstream purification to
95 ensure representativeness.

96 DNA extracts were quantified with Qubit dsDNA HS assay fluorometric system (Invitrogen, USA)
97 and diluted to 5 ng/μL for PCR amplifications. The nuclear small subunit ribosomal RNA gene
98 (18S rRNA gene) was chosen as the marker gene for eukaryotic metabarcoding and a universal
99 18S rRNA primer set (1391f/EukBr) from the Earth Microbiome Project was used to amplify the
100 target V9 region (Amaral-Zettler et al., 2009). The 5' ends of the primers were tagged with
101 Illumina adapters required for multiplexing samples. The first PCR was performed in a reaction

102 mixture of 25 μ L consisting of 12.5 μ L MyTaq 2x mix buffer (Meridian Bioscience), 1 μ L BSA
103 (2.5 mM), 1 μ L of each primer (25 μ M), 3 μ L DNA template and 6.5 μ L PCR water. The PCR
104 conditions consisted of an initial denaturation at 94 °C for 3 min, 35 cycles of denaturation at
105 94 °C for 45 s, annealing at 57 °C for 60 s, and elongation at 72 °C for 90 s, followed by a final
106 elongation at 72 °C for 10 min. PCR amplicons were purified using AMPure XP beads and
107 subsequently used as DNA templates in the index PCR step, following the manufacturer's
108 suggested protocols. Index PCR amplicons were purified using AMPure XP beads, quantified
109 with the Qubit fluorometric system, diluted to 4 nM, and subsequently pooled. Paired-end
110 sequencing (2 \times 300 bp) was performed on the Illumina MiSeq platform (Illumina Inc., San
111 Diego, CA, USA).

112 **Sequence processing and data analyses**

113 Raw data processing was done with the dada2 pipeline (Callahan et al., 2016) in R software v
114 4.5.0 (R Core Team, 2025). Reads were truncated with *FilterAndTrim* function of dada2, with
115 parameters set as trimLeft = 14, 24; truncLen= (165,155); maxN = 0; maxEE = (4, 4); truncQ = 2.
116 Paired reads were merged using *mergePairs* function with minOverlap parameter as default.
117 Taxonomy was assigned using PR2 database v 5.0.0 (Guillou et al., 2013) using the dada2
118 algorithm. Tables with the amplicon sequence variants (ASVs) were analyzed in R using the
119 phyloseq (McMurdie and Holmes, 2013) and vegan (Oksanen et al., 2025) packages.

120 Missing ranks were labeled as "unidentified" in the taxonomy table. ASVs were grouped into
121 taxonomic groups according to taxonomy and representative body size (Luan et al., 2020;
122 Zinger et al., 2019). Fungi, algae, plants, bacteria, and archaea were grouped using domain- or
123 class-level taxonomy. Metazoans were classified at the phylum level (e.g., *Nematoda*, *Annelida*,
124 *Mollusca*), with Arthropoda and Annelida further classified by family when available. Within
125 Annelida, *Enchytraeidae* were treated as intermediate-bodied taxa, whereas Lumbricidae and
126 all other annelid taxa and unidentified taxa were grouped as large-bodied taxa ("Annelida:

127 Lumbricidae&others”). Protists were grouped based on eukaryotic supergroups and divisions
128 (e.g., *Rhizaria*, *Amoebozoa*, *Alveolata*). Oomycetes (e.g., *Peronosporales*, *Saprolegniales*) were
129 classified separately as *Oomycota*. Bacterial, archaeal, and plant taxa were excluded from
130 downstream analyses, as their presence was not relevant to the primer target or ecological
131 focus of this study.

132 To evaluate how soil sample size influences soil eukaryotic diversity across different body sizes,
133 we measured Hill numbers (Chao et al., 2014; Jost, 2006) including observed richness ($q = 0$),
134 exponential Shannon diversity ($q = 1$), inverse Simpson index ($q = 2$), and the dispersion of
135 Bray-Curtis dissimilarities, using the *betadisper* function of the *vegan* package.

136 For each defined taxonomic group, taxa filtered to retain only those present in at least one
137 sample. Rarefaction was performed within each group to the minimum sequencing depth among
138 samples, and groups were excluded if the five lowest-depth samples had fewer than 100 reads.
139 For alpha diversity, observed richness, exponential Shannon, and inverse Simpson indices were
140 calculated using the *microbiome* package (Lahti and Shetty, 2017), and exponential Shannon
141 diversity ($q = 1$) was obtained by exponentiating the Shannon index.

142 GLM families were selected based on the data distributions of the different diversity
143 measurements. A Poisson generalized linear model (GLM with family = poisson) was used to
144 model observed richness, whereas Gaussian GLMs (family = gaussian) were applied for Hill
145 numbers with $q = 1$ and $q = 2$. For beta dispersion, a linear model ($\text{lm}(\text{Dispersion} \sim \log_2(\text{soil}_g))$)
146 was fitted. Taxonomic groups were annotated with representative body size information (Luan et
147 al., 2020; Zinger et al., 2019) and assigned to small-, intermediate-, or large-bodied groups.

148 Group-level slope estimates were then analyzed using random-effects meta-regression using
149 *rma* function in the *metafor* package (Viechtbauer, 2010), with body size as a moderator. This
150 allowed evaluation of size-dependent responses to sample size across diversity metrics.

151 Outputs included group-specific slope plots, meta-analysis forest plots, and summary tables.

152 To assess how similarity in community structure between taxonomic groups varied along a soil
153 mass gradient, we performed pairwise Mantel tests (999 permutations) of Bray-Curtis
154 dissimilarities within each soil mass level between taxonomic groups.

155 Saturation dynamics for Richness, exponential Shannon, and inverse Simpson were modeled
156 using the Michaelis-Menten equation ($y = V_{\max} \cdot m / (K_m + m)$) via Non-linear Least Squares
157 (NLS). The saturation threshold (Sat_{95}) was calculated as $19 \cdot K_m$, corresponding to the soil mass
158 required to reach 95% of the theoretical maximum (V_{\max}).

159 **Results**

160 Across all taxonomic groups, richness (observed ASVs, $q = 0$) responded to soil sample mass,
161 but this relationship varied across body size groups. For several intermediate- and large-bodied
162 taxa (e.g., *Nematoda*, *Arachnida*, *Collembola*, *Insecta*), increasing sample mass led to a higher
163 detected richness (Fig. 1A, left), while small taxa displayed weaker or negative responses. A
164 meta-regression supported this pattern, indicating larger-bodied groups had stronger and more
165 significant positive relationships between sample mass and richness than smaller-bodied taxa
166 ($p < 0.01$; Fig. 1A, right). The exponential Shannon ($q = 1$; Fig. S1) and inverse Simpson ($q = 2$;
167 Fig. 1B) indices showed similar body size-dependent trends. Larger-bodied groups showed
168 consistently stronger positive responses to increasing soil mass, indicating that common and
169 dominant taxa were more effectively detected when larger sample sizes were used. In contrast,
170 for small-bodied taxa such as fungi and protists, diversity estimates tended to plateau or decline
171 as soil mass increased.

172 In comparison, community dispersion decreased with increasing soil mass across all body size
173 groups, with the exception of *Protura*, which showed high variability (Fig. 2). Notably, larger-
174 bodied groups such as *Annelida (Lumbricidae)* exhibited the strongest negative relationships,
175 indicating a substantial reduction in sample-to-sample variability with increased sample mass.

176 To identify how soil sample mass influences inter-group compositional relationships (i.e.,
177 interdependent ecological structures), we performed pairwise Mantel tests between taxonomic
178 groups across all soil mass levels. Few significant correlations ($p \leq 0.05$) were detected (Table
179 S1), occurring occasionally across soil masses and varying among taxa. Positive correlations
180 were observed between certain fungal and arthropod groups, but no consistent pattern with soil
181 mass was evident.

182 To determine the minimum soil mass required for biodiversity saturation across different body
183 sizes, we modeled the accumulation dynamics of species richness, exponential Shannon, and
184 inverse Simpson indices using Michaelis-Menten equations. The results revealed that the soil
185 mass required for richness saturation was strongly dependent on organism body size (Fig. 3).
186 Specifically, small-bodied taxa (e.g., Fungi and Protists) reached saturation at minimal soil
187 masses (often < 0.25 g), whereas intermediate and large-bodied taxa required significantly
188 higher masses to achieve saturation. This size-dependent pattern was consistent across
189 diversity metrics, with exponential Shannon and inverse Simpson indices exhibiting similar
190 accumulation dynamics (Table S2).

191 **Discussion**

192 Understanding how organismal traits shape biodiversity detection is essential for developing
193 effective sampling strategies for soil ecosystems and monitoring. Here, we experimentally
194 examined how sample size influences biodiversity estimates across taxonomic groups differing
195 in body size.

196 Our study demonstrates a clear body size-related response of alpha diversity estimates to
197 increasing soil sample mass across a wide range of taxonomic groups. Specifically, alpha
198 diversity ($q = 0, 1, \text{ and } 2$) increased more steeply with soil mass in larger-bodied taxa such as
199 *Nematoda*, *Arachnida*, *Collembola*, and *Insecta*, indicating that larger-bodied organisms are

200 underrepresented in low soil sample masses. This pattern likely stems from the lower population
201 density, greater spatial heterogeneity, and higher motility of larger soil biota (Bardgett and van
202 der Putten, 2014; Decaëns, 2010; Ettema and Wardle, 2002), which together reduce detection
203 probability at fine spatial scales, necessitating greater sampling effort (e.g. larger soil mass) to
204 capture community composition effectively. In contrast, smaller-bodied taxa such as fungi and
205 protists showed weak or even negative relationships between diversity and soil mass, with most
206 relationships being statistically non-significant. This suggests that increasing sample mass does
207 not substantially enhance the detectable alpha diversity estimates of microeukaryotic taxa. In
208 addition to alpha diversity, community dispersion, i.e., the variance in composition among
209 replicates, decreased consistently with increasing sample mass across all taxonomic groups.
210 This suggests that larger samples improve representativeness by integrating across spatial and
211 biological heterogeneity, thereby reducing stochastic variation (Decaëns, 2010; Ettema and
212 Wardle, 2002; Zinger et al., 2019). Pairwise Mantel tests revealed no consistent pattern,
213 rejecting the notion that compositional correlations between small and larger-bodied taxa
214 increase with larger soil masses. We acknowledge that our experimental design (5 replicates
215 per soil mass) likely reduced our ability to detect these subtle trends.

216 Previous research on soil biodiversity assembly supports our finding that larger sample sizes
217 are needed to capture the full community composition of larger-bodied taxa, whose assembly is
218 more strongly shaped by stochastic processes. Zinger et al. (2019) demonstrated that body size
219 predicts assembly mechanisms in tropical soils, with mesofauna (e.g., *Arthropoda*, *Annelida*)
220 exhibiting greater stochasticity and spatial heterogeneity than microbial taxa, such as bacteria
221 and protists. Similarly, Aslani et al. (2022) reported that as body size increases, the influence of
222 environmental selection weakens, whereas the contribution of ecological drift becomes more
223 pronounced. Beyond organismal traits, methodological constraints, particularly sampling effort,
224 sequencing coverage, and read depth can also amplify apparent stochasticity in community

225 assembly. Limited sampling or insufficient sequencing depth may underrepresent rare or
226 spatially dispersed taxa, thereby enhancing the observed randomness of community patterns.
227 Together, these biological and methodological factors likely explain why larger-bodied taxa
228 exhibit a stronger positive response of alpha diversity to increasing soil mass.

229 In molecular analyses of soil microbial communities, researchers commonly utilize relatively
230 small sample quantities, often between 0.25 to 1 g, since previous evidence has demonstrated
231 that such amounts can yield representative and reproducible results. For fungal communities,
232 even larger soil amounts, generally exceeding 1 g, are recommended to apply to adequately
233 capture their more heterogeneous spatial distribution and community diversity (Kageyama and
234 Toju, 2022; Li et al., 2023; Penton et al., 2016; Ranjard et al., 2003). In contrast, molecular
235 eDNA studies targeting micro- to mesofaunal assemblages, such as nematodes and
236 collembolans, remain less standardized with respect to soil input. Some studies have used
237 small quantities comparable to those applied in microbiome studies (Gong et al., 2023; Seppey
238 et al., 2017), often improving representativeness by pooling several replicates (Hermans et al.,
239 2022; Köninger et al., 2023). Others, however, have employed larger sample sizes, up to 10 g,
240 to obtain a more comprehensive representation of soil community diversity (Donhauser et al.,
241 2023; Gong et al., 2024).

242 This variability reflects ongoing methodological uncertainty regarding the optimal sampling size,
243 which partly arises from trade-offs inherent in molecular procedures. Larger soil quantities can
244 improve representativeness but may also increase the concentration of PCR inhibitors such as
245 humic acids and metals, reducing amplification efficiency (Schrader et al., 2012; Zinger et al.,
246 2016). Moreover, working with large soil masses can raise both extraction difficulty and
247 processing costs, making it an inefficient strategy for routine or long-term biodiversity surveys.
248 For taxa with rigid or resistant body structures, such as chitinous or sclerotized fauna, cell walls
249 and cuticles can hinder complete lysis during extraction, resulting in low DNA release efficiency

250 (Martoni et al., 2022; Waeyenberge et al., 2019). This intrinsic physiological limitation constrains
251 DNA recovery even when soil mass is increased, underscoring the need to improve extraction
252 protocols rather than simply maximizing sample size. Therefore, an optimal soil mass should
253 balance ecological representativeness, extraction efficiency, and technical feasibility.

254 Our findings align with earlier evidence linking sample size to biodiversity estimates, and
255 provide guidance for future studies. Dopheide et al. (2019) reported higher arthropod diversity
256 with larger soil volumes, whereas DNA extraction size had a minimal effect on prokaryote and
257 microeukaryote diversity. Likewise, Kageyama and Toju (2022) found that bacterial and fungal
258 diversity metrics were largely independent of soil mass, while nematode diversity increased
259 significantly with sample size, recommending over 20 g of soil for comprehensive cross-taxa
260 diversity analyses. Our study extends these findings by applying a consistent DNA
261 metabarcoding workflow across a continuous range of soil masses (0.25-32 g) and body-size
262 categories. Our primer selection reflects a generalist approach that is increasingly common in
263 soil biology (Jurburg et al., 2021), and the consistent analytical framework used allows robust
264 comparison among groups, especially in the context of large-scale monitoring frameworks. We
265 acknowledge that this study was conducted in a single soil type. This design was intentionally
266 chosen to minimize environmental heterogeneity, allowing us to clearly isolate the technical
267 impact of sample mass. While future work is needed to validate specific mass thresholds across
268 diverse biomes, the body-size-dependent detection bias observed here highlights a fundamental
269 methodological challenge that likely affects soil monitoring globally.

270 Our non-linear saturation model (Michaelis-Menten) demonstrates that the sample mass
271 required to capture species richness is strongly dependent on organism body size. While micro-
272 eukaryotes (e.g., fungi and protists) reach saturation rapidly with minimal input, larger taxa
273 exhibit more gradual saturation profiles, with Nematoda stabilizing at approximately 2-3 g and
274 larger groups (e.g., Arthropoda, Annelida) often requiring 3-5 g. This distinct taxon-specific

275 sensitivity indicates that standard low-input protocols (e.g., 0.25 g) systematically underestimate
276 the richness of intermediate- and large-bodied taxa. These findings highlight that in studies
277 targeting specific soil taxa, the choice of sample should be tailored to the body size and
278 ecological characteristics of the target groups. Conversely, in more general studies targeting all
279 soil biota (e.g., employing a universal 18S rRNA marker region), we recommend a stratified
280 analytical approach that evaluates taxa separately based on their body size/specific saturation
281 thresholds rather than applying a universal analytical standard (Jurburg et al., 2021).

282 **Acknowledgements**

283 This research was conducted as part of the PhytOakmeter consortium, funded by the Deutsche
284 Forschungsgemeinschaft (DFG, German Research Foundation) – 507084794, and the Swiss
285 National Science Foundation (SNSF). We thank the Helmholtz Centre for Environmental
286 Research – UFZ for logistic support and permission to do research on their sites. We would like
287 to especially acknowledge local support from Experimental field station Bad Lauchstädt (Dr.
288 Ines Merbach). We thank Nicole Steinbach for technical support and Microbial Interaction
289 Ecology Group, Helmholtz Centre for Environmental Research GmbH – UFZ for providing
290 research facilities. We thank Dr. Ulrike Schlägel for valuable statistical consultation at UFZ. To
291 ensure data FAIRness, we utilized the DataPLANT personal assistance network and its tool
292 stack, centered around the DataHUB.

293 **CRedit authorship contribution statement**

294 Lu Wang: Writing – original draft, Writing – review & editing, Visualization, Validation, Software,
295 Methodology, Formal analysis, Data curation, Investigation. April Lyn Leonar: Writing – review &
296 editing. Simone Cesarz: Writing – review & editing, Supervision, Resources. Nico Eisenhauer:
297 Writing – review & editing, Supervision, Resources. Antonis Chatzinotas: Writing – review &
298 editing, Supervision, Resources. Stephanie Jurburg: Writing – review & editing, Supervision,
299 Resources, Project administration, Methodology, Funding acquisition, Conceptualization.

300 **Data Accessibility**

301 Sequencing data generated in this study are available in the NCBI Sequence Read Archive
302 under BioProject accession number PRJNA1359375. Downstream analytical workflows are
303 accessible via GitLab
304 (https://git.nfdi4plants.org/lu.wang.env4all/euk_metabarcoding_body_size).

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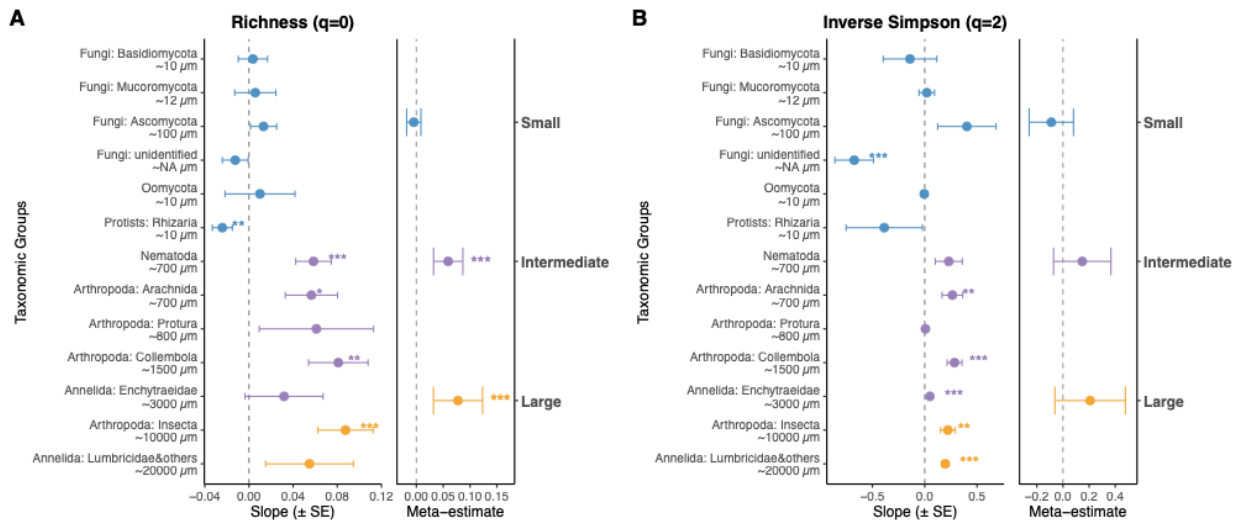
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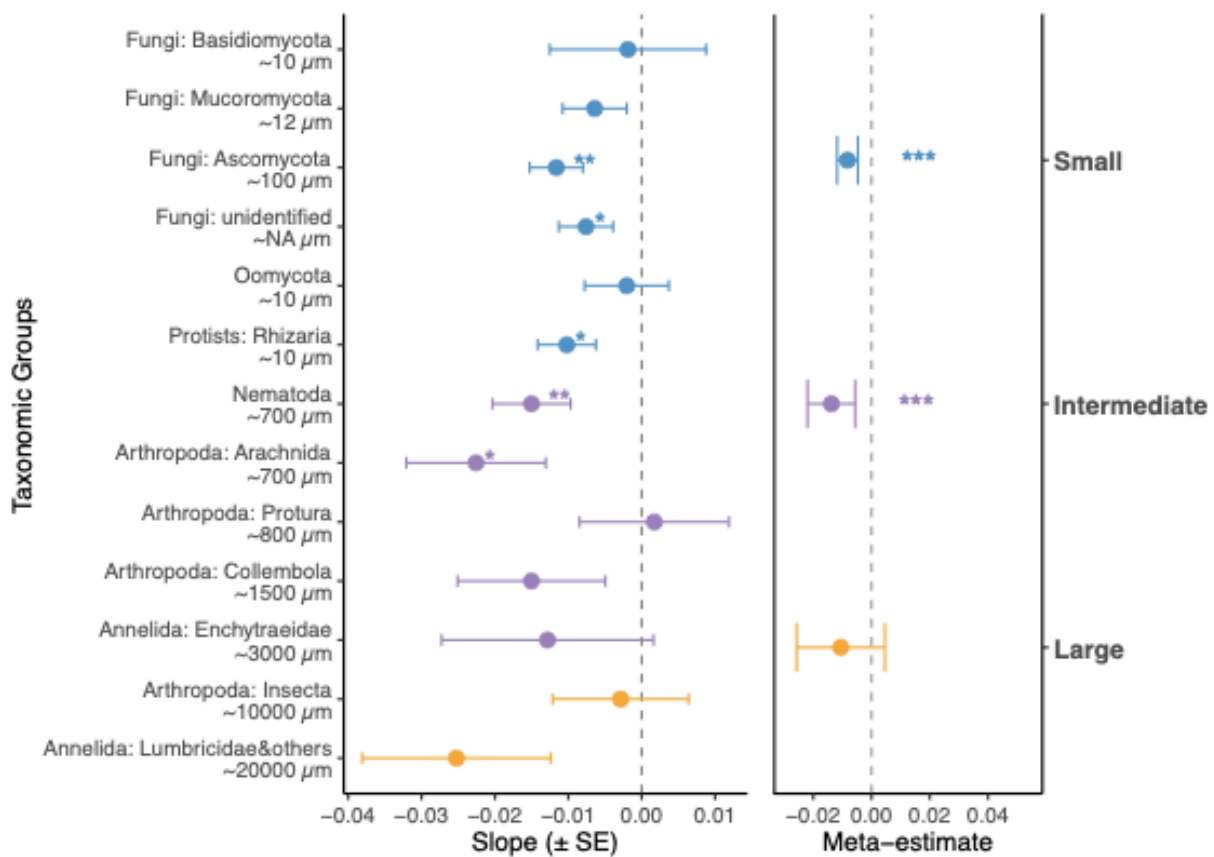
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489 **Figures**



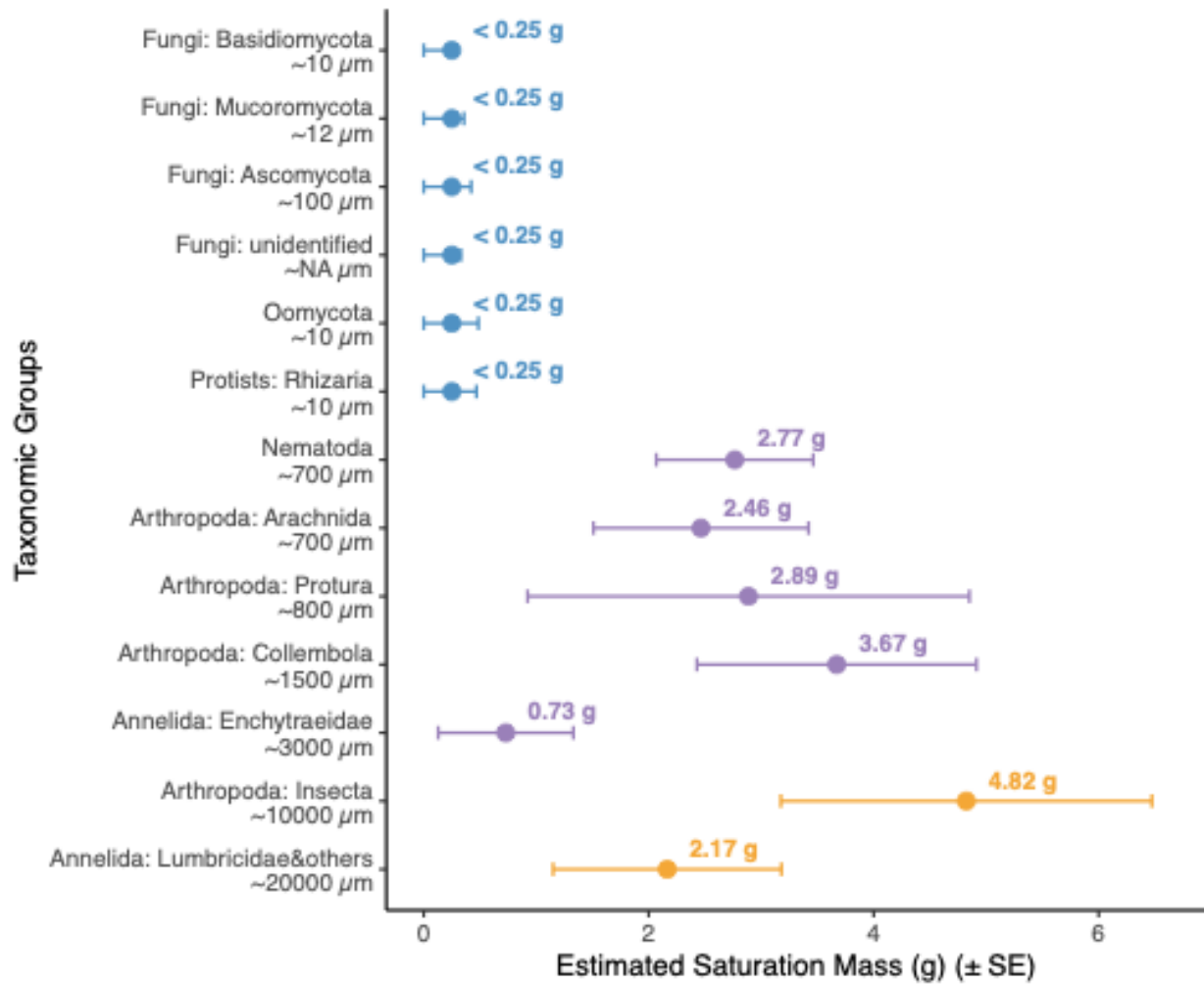
490

491 Fig 1. Soil sample mass effects on alpha diversity estimates across taxonomic groups categorized by
 492 body size. (A) Richness (observed ASVs, $q = 0$) and (B) Inverse Simpson diversity ($q = 2$) were modeled
 493 using GLMs with log-transformed soil mass as predictor. Each point in the main panels represents a
 494 taxonomic group; error bars indicate standard errors; asterisks indicate group-level significance of the soil
 495 mass slope (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$). Groups are ordered by average body size rank from
 496 small to large. Meta-regression results (right panels) summarize effect sizes across body-size categories
 497 (small-, intermediate-, and large-bodied taxa). Asterisks here indicate whether the estimated slope for
 498 each size group is significantly different from zero.



499
 500 Fig 2. Soil sample mass effects on community dispersion (Bray-Curtis distance to group centroids) across
 501 taxonomic groups categorized by body size. Each point in the main panels represents a taxonomic group;
 502 error bars indicate standard errors; asterisks indicate group-level significance of the soil mass slope (* $p <$
 503 0.05 , ** $p < 0.01$, *** $p < 0.001$). Groups are ordered by average body size rank from small to large. Meta-

504 regression results (right panels) summarize effect sizes across body-size categories (small-,
 505 intermediate-, and large-bodied taxa). Asterisks here indicate whether the estimated slope for each size
 506 group is significantly different from zero.



507
 508 Fig 3. Minimum soil mass requirements for species richness saturation (Sat_{95}) across taxonomic groups
 509 categorized by body size. Each point represents a taxonomic group; error bars indicate standard
 510 errors (SE) of the estimated saturation point derived from Michaelis-Menten models. Groups are ordered by
 511 average body size from small to large. Text labels display the specific estimated saturation mass. Labels
 512 of "< 0.25 g" indicate saturation occurred at or below the minimum experimental sampling unit. Colors
 513 differentiate body size categories.

514

515 **Supporting Information**516 **Tables and Figures**

517 Table S1. Significant Mantel correlations among taxonomic groups across soil mass gradients. Pairwise
 518 Mantel tests were performed between community dissimilarity matrices of all taxonomic groups within
 519 each soil mass level. Only correlations with $p \leq 0.05$ are shown. Each row represents a group pair that
 520 exhibited a statistically significant structural correlation (Mantel r) at a given soil mass.

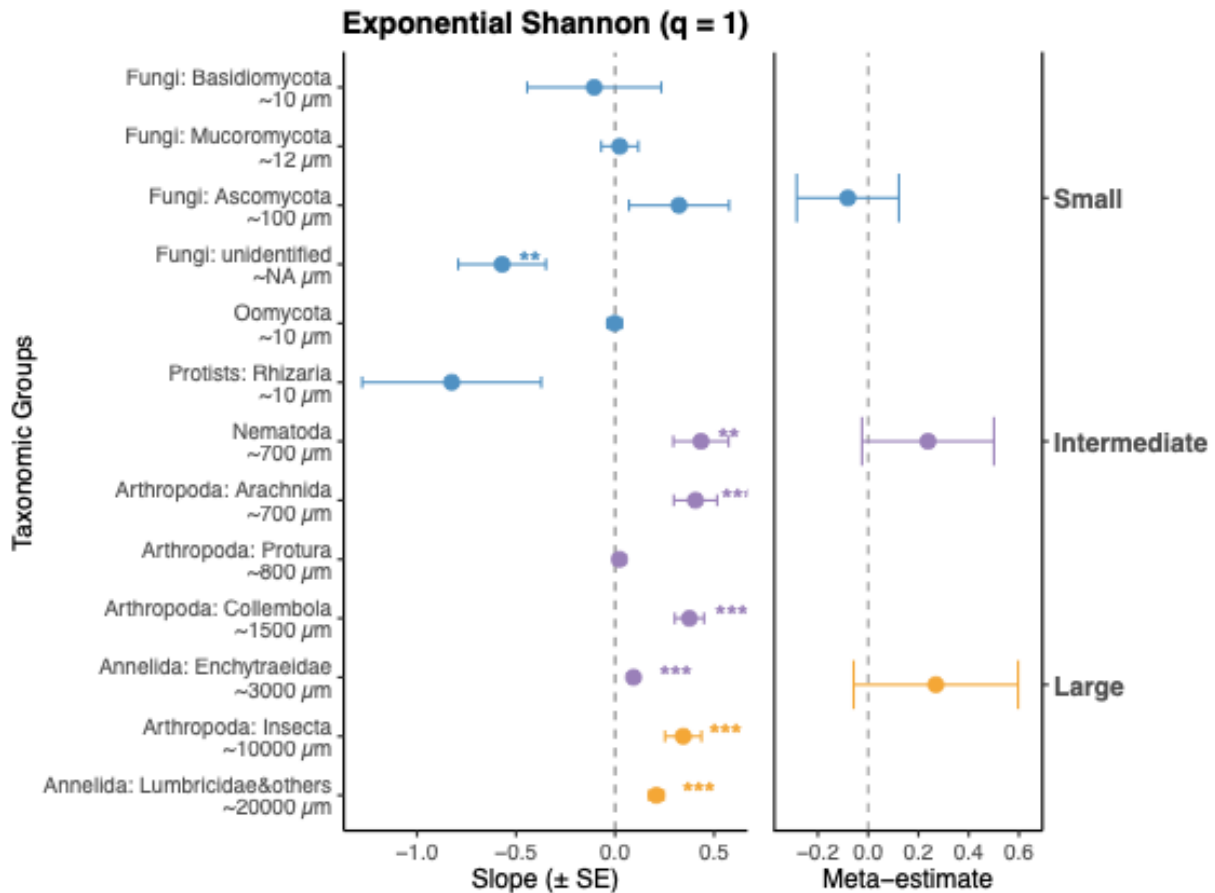
Base_Group	Compare_Group	Soil_Mass	Mantel_r	P_value	N
Annelida_Enchytraeidae	Nematoda	1	0.66	0.050	5
Annelida_Enchytraeidae	Oomycota	32	0.76	0.042	5
Annelida_Enchytraeidae	Protists_Rhizaria	0.5	0.71	0.025	5
Annelida_Lumbricidae&others	Arthropoda_Insecta	0.25	0.56	0.008	5
Fungi_Basidiomycota	Annelida_Lumbricidae&others	8	0.77	0.017	5
Annelida_Lumbricidae&others	Fungi_Mucoromycota	16	0.57	0.017	5
Arthropoda_Arachnida	Arthropoda_Collembola	2	0.56	0.008	5
Arthropoda_Protura	Arthropoda_Arachnida	8	0.91	0.033	5
Fungi_Ascomycota	Arthropoda_Collembola	1	0.51	0.033	5
Fungi_Basidiomycota	Arthropoda_Collembola	0.25	0.90	0.017	5
Nematoda	Arthropoda_Insecta	4	0.67	0.033	5
Oomycota	Arthropoda_Insecta	2	0.48	0.042	5
Fungi_Basidiomycota	Arthropoda_Protura	0.5	0.86	0.050	5
Fungi_Basidiomycota	Arthropoda_Protura	2	0.37	0.033	5
Fungi_Basidiomycota	Arthropoda_Protura	32	0.91	0.017	5
Arthropoda_Protura	Oomycota	0.5	0.49	0.050	5
Fungi_Mucoromycota	Fungi_Ascomycota	0.25	0.68	0.050	5
Fungi_Basidiomycota	Oomycota	0.5	0.51	0.008	5
Fungi_Basidiomycota	Oomycota	1	0.77	0.033	5
Fungi_Basidiomycota	Oomycota	4	0.47	0.008	5
Oomycota	Fungi_Mucoromycota	1	0.58	0.050	5
Oomycota	Protists_Rhizaria	2	0.72	0.025	5
Oomycota	Protists_Rhizaria	8	0.74	0.033	5

521

522 Table S2. Minimum soil sampling mass required for biodiversity saturation (Sat_{95}) across body size
 523 groups. Saturation points were estimated using Michaelis-Menten models based on rarefied amplicon
 524 data. Data represent the estimated saturation mass (g), with R^2 in parentheses. Sat. (< 0.25): Saturation
 525 occurred at or below the minimum experimental sampling unit (indicating rapid stabilization). Proposed
 526 minimum composite sampling mass to ensure adequate coverage of the respective taxon.

Group	Richness	Shannon	Simpson	Mass_Suggest
Fungi_Basidiomycota	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	< 0.25 g
Fungi_Mucoromycota	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	< 0.25 g
Fungi_Ascomycota	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	0.34 (<.01)	~ 1 g
Fungi_unidentified	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	< 0.25 g
Oomycota	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	< 0.25 g
Protists_Rhizaria	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	Sat.<(0.25) (<.01)	< 0.25 g
Nematoda	2.77 (0.41)	2.21 (0.25)	1.83 (0.13)	~ 3 g
Arthropoda_Arachnida	2.46 (0.24)	4.89 (0.22)	3.71 (0.12)	~ 5 g
Arthropoda_Protura	2.89 (0.10)	1.20 (0.06)	0.87 (0.04)	~ 3 g
Arthropoda_Collembola	3.67 (0.31)	4.07 (0.37)	3.94 (0.31)	~ 5 g
Annelida_Enchytraeidae	0.73 (0.04)	1.93 (0.35)	1.16 (0.25)	~ 2 g
Arthropoda_Insecta	4.82 (0.32)	3.76 (0.19)	2.88 (0.13)	~ 5 g
Annelida_Lumbricidae&others	2.17 (0.15)	4.75 (0.34)	5.15 (0.36)	~ 6 g

527



528

529 Fig. S2. Soil sample mass effects on Exponential Shannon diversity (q = 1) across taxonomic groups
 530 categorized by body size. Exponential Shannon diversity was modeled using generalized linear models
 531 (GLMs) with log-transformed soil mass as the predictor variable. Each point in the main panels represents
 532 a taxonomic group; error bars indicate standard errors; asterisks denote significance levels for the soil
 533 mass slope compared to zero (* p < 0.05, ** p < 0.01, *** p < 0.001). Groups are ordered by average body
 534 size rank from small to large. Meta-regression results (right panels) summarize effect sizes across body-
 535 size categories (small-, intermediate-, and large-bodied taxa), with asterisks indicating whether each size-
 536 group slope differs significantly from zero.

537