

1 **Research article**

2

3 **Title:** Phylogenetic placement and contamination screening of Amoebozoa genomic data from the  
4 Protist 10,000 Genomes (P10K) Database

5

6 **Authors:** Alfredo L. Porfirio-Sousa<sup>a,1</sup>, Robert E. Jones<sup>a</sup>, Matthew W. Brown<sup>b</sup>, Daniel J. G. Lahr<sup>c</sup>,  
7 Alexander K. Tice<sup>a,1</sup>

8 <sup>a</sup> Department of Biological Sciences, Texas Tech University, Lubbock, TX, USA

9 <sup>b</sup> Department of Biological Sciences, Mississippi State University, Mississippi State, MS, USA

10 <sup>c</sup> Institute of Biosciences, University of São Paulo, São Paulo 055080-090, Brazil

11 <sup>1</sup> corresponding authors: [alex.tice@ttu.edu](mailto:alex.tice@ttu.edu) (A. K. Tice) and [alfredolpsousa@gmail.com](mailto:alfredolpsousa@gmail.com) (A. L. Porfirio-Sousa)

13

14 **ORCID**

15 ALPS: <https://orcid.org/0000-0001-7490-158X>

16 REJ: <https://orcid.org/0000-0001-5227-4773>

17 MWB: <https://orcid.org/0000-0002-1254-0608>

18 DJGL: <https://orcid.org/0000-0002-1049-0635>

19 AKT: <https://orcid.org/0000-0002-3128-1867>

20

21

22

23

24

25

26

27

28

29

30

31        **Abstract**

32        **Background:** Genomic data are essential for uncovering the evolutionary history, ecological roles,  
33 and diversity of life. Yet, microbial eukaryotes like Amoebozoa, an ancient and morphologically diverse  
34 lineage, remain critically underrepresented in genomic repositories. This has limited our ability to  
35 address fundamental questions in eukaryotic evolution. The Protist 10,000 Genomes (P10K) initiative  
36 seeks to fill this gap by generating and compiling genome- and transcriptome-level data for a wide  
37 range of microbial eukaryotes. To ensure the reliability of these resources, accurate taxonomic  
38 identification and contamination screening are vital. In this study, we aimed to assess the taxonomic  
39 consistency and integrity of the P10K database with a phylogenetic-based approach using Amoebozoa  
40 as a case study.

41        **Results:** Through SSU rDNA/rRNA and COI phylogenetic reconstructions this study confirmed  
42 several initial taxonomic identifications provided in the P10K database, resolved ambiguities at higher  
43 taxonomic levels, and corrected misassignments among morphologically similar but phylogenetically  
44 distant taxa. Moreover, the contamination screening using SSU rDNA/rRNA revealed several  
45 amoebozoan data that are contaminated by sequence from other eukaryotic taxa, representing  
46 contaminated genomic assemblies.

47        **Conclusion:** Phylogenetic placement coupled with contamination screening enabled us to  
48 distinguish the higher-quality Amoebozoa datasets currently available in the P10K database from those  
49 requiring decontamination or additional sequencing before downstream use. These findings serve as a  
50 reference for the future use of these data and as a guide for further sequencing efforts aimed at  
51 expanding the taxonomic diversity of Amoebozoa represented at the genomic level. By applying a  
52 phylogenetic survey to the Amoebozoa data, we present a framework that can be extended to other  
53 microbial eukaryote lineages. Addressing imprecise taxonomic identifications and contamination in  
54 certain P10K datasets, as well as data reproducibility, will further enhance the value of this  
55 unprecedented genomic resource for protists, with significant potential to illuminate the evolution and  
56 diversification of eukaryotic life.

57        **Keywords:** Amoebozoa, Arcellinida, Eukaryotic diversity, Comparative genomics, Genomic  
58 database curation

59

60

61 **Introduction**

62

63       Genomic-level data are essential for advancing our understanding of the evolution of life on Earth  
64 [1, 2]. High-quality genome and transcriptome sequences enable comparative analyses that reveal  
65 patterns of genomic evolution, including gene family expansions, horizontal gene transfers, and  
66 changes in genome organization and regulatory systems [3, 4]. Such data also clarify phylogenetic  
67 relationships through multi-gene and genome-scale reconstructions [5, 6]. Moreover, genomic analyses  
68 uncover ecological interactions by identifying metabolic pathways, symbiotic associations, and genetic  
69 adaptations to specific environmental conditions [7]. Consequently, genomics has become central to  
70 studying life's diversity and complexity.

71       While genomics has progressed rapidly for major eukaryotic groups such as plants, animals, fungi,  
72 and other traditional model systems, most microbial eukaryotes (commonly referred to as protists)  
73 remain vastly underrepresented [1, 2]. Although they comprise most of eukaryotic diversity, the majority  
74 of lineages within this highly diverse paraphyletic assemblage still lack genomic data [2, 5]. Over the  
75 past decade, several initiatives have contributed to fill this gap, such as the Marine Microbial Eukaryote  
76 Transcriptome Sequencing Project (MMETSP) for marine microbial genomics [8], the Tree of Life  
77 Programme for eukaryotic genome sequencing [9], and the One Thousand Plant Transcriptomes  
78 initiative (1KP) for plants, including single-celled algae [10]. Similarly, the Protist 10,000 Genomes  
79 (P10K) initiative aims to address the underrepresentation of microbial eukaryotes in genomic databases  
80 by generating new genomic data (i.e., genomes and transcriptomes) and compiling previously available  
81 data from a wide array of lineages, coupled with taxonomic identification and decontamination  
82 procedures [2]. This large-scale effort, consolidated in the P10K database, represents an  
83 unprecedented genomic resource for protists and potentially provides a valuable foundation for  
84 achieving a comprehensive and integrative view of eukaryotic evolution.

85       However, to fully exploit the potential of this newly available genomic resource for protists, three  
86 key challenges must be considered: (1) accurate taxonomic identification, (2) the presence of  
87 contaminated genomic data resulting from non-target eukaryotic contamination, and (3) data  
88 interpretation and reproducibility [11, 12]. Specifically, the taxonomic identification of the P10K  
89 database relies on small subunit ribosomal DNA (SSU rDNA) retrieval and BLAST similarity searches  
90 against curated SSU databases, including SILVA, PR<sup>2</sup>, and NCBI's NT (nucleotide sequence) database

91 [2]. Although informative, a BLAST-based approach is less accurate than phylogeny-based methods  
92 for taxonomic identification and may lead to a less precise classification of the target groups. Regarding  
93 contamination, protists typically inhabit environments with high eukaryotic microbial diversity, several  
94 prey on other organisms, many host endosymbionts, and most are often difficult to isolate as  
95 monoeukaryotic cultures or single cells [13–15]. As a result, genomic-level sequencing efforts might  
96 produce contaminated assemblies that contain not only the genome of the target organism but also  
97 sequences from other eukaryotic organisms [16, 17]. While P10K performs a decontamination strategy  
98 to remove bacterial, archaeal, viral, fungal, and other eukaryotic contaminants from ciliate data; for  
99 other protist groups, only bacterial, archaeal, and viral contaminants are filtered out [2]. Finally, the  
100 P10K database currently lacks photo-documentation and detailed information, for instance, about the  
101 sequencing platforms used for each sample (i.e., Illumina, Oxford Nanopore, or PacBio Sequel II),  
102 which are essential for deeper interpretation and reproducibility of the data. In this context, accurate  
103 taxonomic identification, assessment of eukaryotic contamination, and the availability of more detailed  
104 sample information are most essential for the effective downstream use of data in the P10K database.

105 Here, we present a survey of the taxonomic identification and contamination screening of the P10K  
106 database using Amoebozoa as a case study. Amoebozoa is an ancient clade of heterotrophic  
107 organisms, estimated to have originated around 1.5 billion years ago, and it exhibits remarkable  
108 morphological and ecological diversity [13, 15, 18]. Because Amoebozoa are heterotrophic and often  
109 share habitats with a wide range of eukaryotic organisms, making them particularly prone to sequence  
110 contamination, they serve as an ideal model group for evaluating the accuracy of taxonomic  
111 identification and contamination assessment in the P10K database. In short, we retrieved all publicly  
112 available Amoebozoa genomes from the P10K database, representing the three major amoebozoan  
113 clades; Tubulinea, Evosea, and Discosea [13]. From these data, we conducted phylogenetic analyses  
114 using two commonly employed molecular markers: small subunit ribosomal DNA/RNA (SSU  
115 rDNA/rRNA or 18S) and cytochrome c oxidase subunit I (COI, also known as CO1 or cox1). This  
116 phylogenetic investigation enables the phylogenetic placement and contamination screening of  
117 Amoebozoa genomic data present in the P10K database, leading to the identification of higher-quality  
118 data, while also highlighting those that require decontamination or additional sequencing prior to  
119 downstream analyses. Additionally, we present some strategies that can mitigate contamination during  
120 sample preparation and discuss how the availability of more detailed metadata directly associated to

121 each sample can improve the interpretation, usability, and reproducibility of P10K data. Ultimately, this  
122 study serves as a proof of concept for using a phylogenetic framework to improve taxonomic  
123 identification and contamination assessment within the P10K dataset, as an approach that can be  
124 extended to other taxa represented in the database.

125

## 126 **Methods**

127

### 128 *Datasets construction*

129

130 We retrieved genomic-level assemblies for the 201 amoebozoans from the P10K database based  
131 on the taxonomic annotations provided on the platform (**Supplementary Information - Table S1**). We  
132 constructed a small subunit ribosomal DNA/RNA (SSU rDNA/rRNA) dataset considering all 201 of these  
133 assemblies. Further, we constructed a cytochrome c oxidase subunit I (COI) dataset focusing only on  
134 the testate amoebae order Arcellinida (Tubulinea), since this marker is well sampled for this lineage as  
135 it has been traditionally used for phylogenetics in arcellinids. To construct datasets for the small subunit  
136 ribosomal DNA/RNA (SSU rDNA/rRNA) and cytochrome c oxidase subunit I (COI) markers, we  
137 extracted sequences from these assemblies using similarity searches implemented in BLAST+  
138 v2.16.0+ and a custom Python script (**Supplementary Information - File S1**). For each marker, the  
139 script automated the creation of BLAST databases using makeblastdb and performed local blastn  
140 searches for each query sequence in a multi-FASTA file against each Amoebozoa genomic data.  
141 Specifically, the script executed the commands *makeblastdb -in P10KID.fasta -dbtype nucl* and *blastn*  
142 *-query marker\_query -db P10K.fasta -outfmt [script\_default\_choice]* (**Supplementary Information -**  
143 **File S1**). The script retrieved the top five hits per genomic data file, extracting each aligned region along  
144 with 1000 bp of upstream and downstream flanking sequence. This approach enabled recovery of  
145 extended SSU and COI regions suitable for downstream phylogenetic analyses and compatible with  
146 Amoebozoa SSU and COI data available in the PR<sup>2</sup> database [19] and NCBI. The orientation of each  
147 retrieved sequence was assessed, and sequences were reverse complemented when necessary to  
148 match the strand of the original query. For the query sequences, we used SSU rDNA/rRNA data from  
149 the PR<sup>2</sup> database for three representative species of the major amoebozoan clades: *Arcella vulgaris*  
150 WP (Tubulinea; GenBank: HM853762.1), *Dictyostelium discoideum* (Evosea; GenBank: AM168040.1),

151 and *Acanthamoeba castellanii* Neff (Discosea; GenBank: U07416.1). As the COI query sequence, we  
152 considered the sequence of *Arcella uspiensis* (SRR5396453).

153 To build a phylogenetically informative datasets, we combined the retrieved sequences with  
154 previously published SSU rDNA/rRNA and COI datasets for Amoebozoa, as well as sequences from  
155 the PR<sup>2</sup> database for SSU [13, 15, 19, 20]. For SSU, we curated a non-redundant dataset broadly  
156 representative of the major lineages in Amoebozoa (Tubulinea, Evosea, and Discosea) and for COI a  
157 dataset broadly representative of the major lineages in Arcellinida, for both markers excluding  
158 environmental sequences. This strategy ensured a robust and interpretable dataset for our phylogenetic  
159 framework. From a preliminary phylogenetic reconstruction, we curated the SSU and COI datasets  
160 used to generate the main trees in this study. This initial analysis allowed us to identify and remove  
161 identical or highly similar sequences that resulted from retrieving the top five BLAST+ hits per genome.  
162 It also enabled visual inspection of the tree to detect SSU sequences that either failed to cluster within  
163 Amoebozoa or formed unusually long branches. Many of these were short sequences (<200 bp) and  
164 were excluded from downstream analyses, while some long branches corresponded to full-length  
165 sequences that likely represented contaminants. These were retained for further contamination  
166 screening.

167 For contamination screening, we focused on the SSU rDNA/rRNA dataset. After the initial  
168 phylogenetic reconstruction (see *Phylogenetic reconstructions* section), any SSU sequences from  
169 P10K genome assemblies that represented long branches or did not branch within the Amoebozoa  
170 clade were selected for further analysis. These sequences were subjected to additional BLAST+  
171 searches against the PR<sup>2</sup> database, a curated SSU resource representing eukaryotic diversity, using  
172 our custom script and the same parameters described above (**Supplementary Information - File S1**).  
173 To perform this search locally, we downloaded the complete PR<sup>2</sup> database and used it as the reference  
174 for the BLAST+ similarity search. This approach allowed us to retrieve SSU sequences from the PR<sup>2</sup>  
175 database that were similar to those of the putative contaminant eukaryotes and to assign their  
176 taxonomic affiliations through subsequent phylogenetic analyses.

177

178

179 *Phylogenetic reconstructions*

180

181 All phylogenetic reconstructions in this study were based on multiple sequence alignments (MSAs)  
182 generated using MAFFT v7.490 with the E-INS-I algorithm and 1000 refinement iterations. Alignments  
183 were produced with the following command: *mafft --genadpair --maxiterate 1000 input.fasta >*  
184 *output\_aligned.fasta*. Automated alignment trimming was performed with trimAl v1.2, using the  
185 command *trimal -in input\_aligned.fasta -out output\_aligned\_trimmed.fasta -keepheader -gt [threshold]*,  
186 where the gap threshold was set to 0.3 for SSU rDNA/rRNA and 0.5 for COI. Phylogenetic trees were  
187 inferred from the trimmed alignments using the maximum likelihood method implemented in IQ-TREE  
188 v2.3.6, with ModelFinder for model selection and node support assessed via 1,000 ultrafast bootstrap  
189 replicates and 1,000 SH-aLRT tests. The analysis was executed with the command *iqtree2 -s*  
190 *aligned\_trimmed.fasta -alrt 1000 -bb 1000 -m TEST*.

191

## 192 **Results and discussion**

193

### 194 *Phylogeny-based taxonomic identification of P10K amoebozoan data*

195

196 Amoebozoan small subunit ribosomal DNA/RNA (SSU rDNA/rRNA) sequences were successfully  
197 retrieved from 151 of the 201 genomic datasets available for Amoebozoa in the P10K database  
198 (**Supplementary Information - Table S1**). Given the established use of the Cytochrome c oxidase  
199 subunit I (COI) marker in Arcellinida phylogenetics, we also specifically targeted this marker for  
200 arcellinid taxa. Arcellinid COI sequences were successfully recovered from 40 of the 59 genomic data  
201 available for Arcellinida (**Supplementary Information - Table S1**). Phylogenetic analyses based on  
202 SSU rDNA/rRNA and COI were largely congruent, supporting most of the original taxonomic  
203 assignments in the P10K database (**Figs. 1 and 2**). Moreover, they enabled a more precise  
204 identification for 43 taxa, including a refined classification at the genus and family levels for taxa initially  
205 classified only at the family or higher levels (**Figs. 1 and 2; Supplementary Information - Table S1**).  
206 As species-level identification typically requires extensive morphological and morphometric data in  
207 addition to molecular evidence, we adopted a conservative approach and assigned identifications at  
208 the genus level based on our phylogenetic results. However, it is worth noting that several genomic  
209 data from the P10K database originate from well-established cultures of widely used and shared strains,  
210 some originally available in the NCBI database, for which detailed morphological data are available in

211 the literature, allowing confident species-level identifications (**Figs. 1 and 2; Supplementary**  
212 **Information - Table S1**).

213 Notably, seven original taxonomic assignments from the P10K database were not corroborated  
214 by our phylogenetic reconstructions (**Figures 1 and 2; Supplementary Information - Table S1**). This  
215 was most apparent within Arcellinida, where misidentifications primarily involved closely related or  
216 morphologically similar genera and families. Although Arcellinida are well known for their test (shell),  
217 which provides informative taxonomic characters, convergent evolution has led to similar shell  
218 morphologies across distantly related lineages. For example, species of *Diffugia* (Diffugiidae),  
219 *Hyalosphenia* (Hyalospheniidae), and *Netzelia* (Netzelliidae) often possess rounded, ovoid, or  
220 elongated shells, which can lead to misidentification if other shell features (e.g., aperture shape,  
221 composition) or cellular characteristics are not considered [21, 22]. Consistent with this, our analyses  
222 revealed that several taxa initially assigned to Diffugiidae (infraorder Longithecina) belong to  
223 Hyalospheniidae (Hyalospheniformes) or Netzelliidae (Sphaerothecina) (**Supplementary Information**  
224 **- Table S1**). These infraorders are distantly related, with their last common ancestor estimated to have  
225 lived over 500 million years ago [15, 23], making such misassignments evolutionarily significant.  
226 Outside Arcellinida, only one notable case of misidentification was observed: *Pessonella* sp. PRA-29,  
227 a culture originally submitted to the American Type Culture Collection (ATCC) under the genus  
228 *Pessonella*, was later described as a new genus and species, *Armaparvus languidus*, representing the  
229 correct taxonomic identification for this organism [24].

230

231 *Contamination screening of the P10K amoebozoan data*

232

233

234 Non-amoebozoan SSU sequences were retrieved from 58 of the 201 genomic datasets available  
235 for Amoebozoa in the P10K database (**Supplementary Information - Table S1**). The phylogenetic  
236 analysis of these non-amoebozoan SSU and SSU sequences from the PR<sup>2</sup> database reveals a  
237 widespread contamination of the P10K amoebozoan dataset by a taxonomically diverse set of  
238 eukaryotic lineages, mirroring the ecological complexity of the environments these protists inhabit  
239 (**Figure 3 and Supplementary Information - Table S1**). Among the contaminant groups were fungi  
240 and metazoans, including sequences from arthropods and nematodes, likely introduced via soil

241 particles, organic debris, or sample handling procedures (**Figure 3**). SSU sequences affiliated with  
242 ciliates were particularly abundant, with representatives spanning multiple clades such as Vorticellidae,  
243 *Coleps*, and *Pseudomicrothorax* (**Figure 3**). Additional contaminants included lineages within the  
244 Stramenopiles such as *Paraphysomonas* and *Poterioochromonas*, as well as centrohelid heliozoans  
245 (Haptista) and cercozoans (Rhizaria) (**Figure 3**). Several photosynthetic eukaryotes represented  
246 another major group of contaminants, including land plants, especially angiosperms of the Fabaceae  
247 family, likely introduced through pollen or plant debris, as well as green algae from the Chlorophyceae  
248 (e.g., *Chlamydomonas*, *Hyalomonas*) and Zygnematophyceae, the closest relatives of land plants  
249 (**Figure 3**). Collectively, these findings demonstrate that contamination in amoebozoan assemblies is  
250 both frequent and taxonomically widespread, encompassing multiple branches of the eukaryotic tree.

251

252

253 *Quality assessment of the P10K amoebozoan data*

254

255 Based on the results from single-marker retrieval, phylogenetic inference, and quality assessment,  
256 we were able to evaluate the current state of Amoebozoa genomic data in the P10K database. As a  
257 relative measure, we can consider samples to be of relatively higher quality if they contain the SSU  
258 marker (and COI for arcellinids), show no signs of contamination based on SSU phylogenetic  
259 reconstruction, and have a BUSCO completeness score of at least 50% (**Figure 4; Supplementary**  
260 **Information - Table S1**). These samples are more likely to yield meaningful results in downstream  
261 analyses, including phylogenomics and comparative genomics (**Figure 4; Supplementary**  
262 **Information - Table S1**). It is important to note, however, that the SSU-based contamination screening  
263 we used as an exploratory assessment of the amoebozoan P10K data has limitations and cannot, on  
264 its own, detect all sources of potential contamination. Therefore, incorporating additional markers for  
265 further screening is advisable before using the data in downstream analyses. Similarly, while we chose  
266 a relatively low BUSCO threshold ( $\geq 50\%$ ) to include a broader range of potentially useful genomic data.  
267 More stringent completeness cutoffs, as well as additional quality metrics such as genome contiguity,  
268 N50 values, and assembly size, are required depending on the downstream applications of the data  
269 and should be considered accordingly. On the other hand, the P10K samples lacking key markers,  
270 flagged as contaminated based on SSU phylogenetic inference, or with BUSCO scores below 50%

271 represent more incomplete and lower-quality genomic data (**Figure 4; Supplementary Information -**  
272 **Table S1**). In this context, several of the amoebozoan genomic data available in the P10K database  
273 require decontamination or further sequencing prior to downstream analysis (**Figure 4D;**  
274 **Supplementary Information - Table S1**). Finally, this quality assessment of the P10K genomic  
275 resource highlights that several major Amoebozoa lineages remain unsampled at the genomic level  
276 and serves as a useful guide for targeted sampling efforts aimed at expanding the taxonomic diversity  
277 sampled for genomic data.

278

279

#### 280 *Widespread Contamination in Protist Genomes: Sources, Impacts, and Mitigation Strategies*

281

282 While the widespread contamination identified in amoebozoan genome assemblies is certainly  
283 undesirable, it is consistent with the natural ecological context not only of Amoebozoa but of microbial  
284 eukaryotes in general. Protists typically inhabit complex microbial communities, including soil, biofilms,  
285 freshwater and marine sediments, mosses, and decaying organic matter, where they coexist and  
286 interact with a wide diversity of organisms [13, 21, 25]. Like many free-living protists, amoebozoans are  
287 predatory and feed on bacteria, algae, fungi, and other eukaryotic cells, or form close physical  
288 associations with them, such as transient or stable endosymbiotic relationships [13, 21]. These  
289 ecological interactions, combined with the technical difficulty of isolating single amoebozoan cells free  
290 from other microbial associates, make the presence of contaminant sequences in genome assemblies  
291 not only possible but likely [16]. Many species cannot be maintained in long-term axenic or  
292 monoeukaryotic cultures, and even those that can often require extensive purification efforts to  
293 eliminate co-cultured organisms [26].

294 Contaminant sequences can significantly impact genomic-level downstream analyses. They may  
295 compromise gene prediction, reveal artifactual patterns of gene family evolution, mislead functional  
296 annotations, and introduce biases in comparative genomic studies [4, 27, 28]. More specifically,  
297 contamination can lead to overestimation of genomic complexity and distorts analyses of gene family  
298 evolution by introducing homologs or paralogs from unrelated lineages, which can result in artificial  
299 expansions or contractions of gene families and misrepresentation of evolutionary trajectory studies [4,  
300 27, 28]. Functional annotation is similarly affected, as contaminant sequences may be erroneously

301 assigned to the target genome, leading to inaccurate inferences about metabolic capabilities, signaling  
302 pathways, or ecological roles [4, 27, 28]. Similarly, assemblies that include sequences from diverse  
303 eukaryotic contaminants may cluster incorrectly in phylogenomic datasets, leading to artifactual  
304 phylogenetic trees [29]. Technically, contamination can also affect genome completeness metrics by  
305 artificially increasing the number of expected genes detected. This may create the false impression of  
306 high assembly quality and completeness, even when substantial portions of the assembly are derived  
307 from non-target organisms [16]. Ultimately, unaddressed contamination undermines efforts to draw  
308 biologically meaningful conclusions about evolution, diversity, and functional biology from genomic  
309 data.

310 Given these challenges, several strategies that are not reported to be used by the P10K initiative  
311 have been successfully used to reduce contamination in protist genomic-level data generation efforts,  
312 including for Amoebozoa [13, 15, 23]. For taxa that can be cultivated, growing cultures through multiple  
313 generations can help eliminate contaminant organisms that were initially co-isolated with the target  
314 taxon [13, 15, 23]. Other effective practices include visual inspection of cultures to detect fungi or small  
315 eukaryotes, filtration of culture media, and rigorous sterile handling during DNA and RNA extraction.  
316 For species that must be isolated as single cells from environmental samples, useful techniques include  
317 repeated transfers of the cell through filtered sterile water followed by overnight starvation in sterile  
318 medium [13, 15, 23]. These procedures allow cleaning of the cells and digestion of prey items, reducing  
319 the risk of capturing genetic material from non-target organisms derived from their environment or food  
320 source.

321 Even when applying these methods, it is not possible to guarantee that genomic data will be  
322 completely free from contamination, particularly when working with environmentally derived specimens.  
323 Therefore, comprehensive screening of genome assemblies remains essential. When contamination is  
324 detected, identifying the phylogenetic affinities of non-target sequences can guide decisions on data  
325 curation and inform subsequent analyses [29]. As corroborated in this study, combining single marker-  
326 based phylogenetic screening with genome-level examination provides a powerful and generalizable  
327 strategy for distinguishing genuine genomic content from artifactual sequences, especially SSU that  
328 have been traditionally sequenced from diverse eukaryotes and for which comprehensive curated  
329 databases like PR<sup>2</sup> are available. Importantly, this strategy couples the strengths of likelihood-based  
330 phylogenetics, which outperform similarity-based approaches such as BLAST by incorporating explicit

331 models of molecular evolution and statistically grounded inference [30, 31]. These features allow for  
332 more accurate reconstruction of evolutionary relationships, particularly among divergent or closely  
333 related taxa, where mere sequence similarity may be misleading [31, 32]. This approach serves as a  
334 guide for data curation and ensures that genomic data accurately represent the biology and evolutionary  
335 histories of the target protist lineages.

336

337 *Reproducibility of the P10K data*

338

339 Currently, the P10K database lacks photo-documentation and detailed metadata directly  
340 associated with each sample, which impairs both the reproducibility and deeper interpretation of the  
341 data. In particular, the absence of voucher images of specimens or cultures from which the data were  
342 derived prevents taxonomic confirmation and revision. Without such reference material, especially for  
343 uncultivable organisms, it becomes impossible to attempt re-isolation of the target taxa for additional  
344 sequencing efforts aimed at generating more complete genomic data. Moreover, the integration of  
345 morphological documentation with phylogenetic analyses would not only improve taxonomic accuracy  
346 and robustness, but also enable more comprehensive, integrative discoveries that combine molecular  
347 and morphological information. Ideally, the database could include voucher photographs and, when  
348 possible, images of multiple specimens of the same taxon from environmental samples or cultures. This  
349 would facilitate further morphometric analyses and potentially contribute to the utility of the P10K  
350 dataset for researchers who study protists using both molecular and morphological approaches.  
351 Another limitation affecting data accessibility and reproducibility is the lack of clear information on the  
352 specific sequencing platforms used for each sample (e.g., Illumina with short or long-insert libraries,  
353 Oxford Nanopore, or PacBio Sequel II). Since each platform has characteristic error profiles and biases,  
354 this metadata is critical for downstream analyses and informed interpretation of the genomic data. Thus,  
355 the incorporation of photo-documentation and detailed metadata for each sample could substantially  
356 contribute to the scientific value, reproducibility, and long-term impact of the P10K database.

357 **Conclusions**

358 This study presents a comprehensive phylogenetic assessment of taxonomic assignments and  
359 contamination across Amoebozoa genomic datasets in the P10K database. By using SSU rDNA/rRNA

360 and COI markers, we confirmed many of the original classifications, refined others, and identified  
361 multiple cases of misidentification within morphologically similar lineages. Additionally, we uncovered  
362 widespread contamination by diverse eukaryotic lineages, including fungi, metazoans, green algae, and  
363 other protists. These findings highlight the ecological complexity of protist-associated environments and  
364 the inherent challenges of obtaining contamination-free genomic data for target lineages. Despite these  
365 challenges, our results demonstrate that single-marker phylogenetic screening, particularly using SSU,  
366 provides a reliable and scalable strategy for verifying taxonomic identity and for an exploratory detection  
367 of contamination. By improving the taxonomic resolution and reliability of available genome assemblies,  
368 this work contributes to downstream evolutionary, ecological, and functional genomic studies of  
369 Amoebozoa enabled by the genomic resources available in the P10K database. More broadly, our  
370 framework offers a practical and generalizable approach for curating the growing volume of genomic  
371 data from protists. As genomic resources for microbial eukaryotes continue to expand, phylogenetically  
372 aware data curation efforts, alongside the strategies to minimize contamination and improve data  
373 reproducibility discussed here, will be critical to ensuring data accuracy. Accordingly, the P10K project  
374 already envisions improving the reliability of its genomic data in future developments by providing, for  
375 instance, bioinformatic tools for multiple sequence alignment and phylogenetic analysis, made available  
376 through the P10K database. Ultimately, this will maximize the impact of the genomic data available  
377 through the P10K initiative in addressing major biological questions, including the origins of complex  
378 traits, symbioses, multicellularity, and the diversification of eukaryotic life on Earth.

379 **Availability of data and materials**

380 All compiled and curated SSU and COI datasets associated with this manuscript are publicly available  
381 on FigShare at <https://doi.org/10.6084/m9.figshare.29814947>

382 **Competing interests**

383 The authors declare that they have no competing interests

384 **Funding**

385 A.L.P-S and R.E.J were supported by startup funds provided to A.K.T. by Texas Tech University. This  
386 work was supported by the National Science Foundation Division of Environmental Biology (2100888)  
387 awarded to M.W.B., D.J.G.L. is supported by a FAPESP award #2019/22815-2.

388        **Authors' contributions**

389        **Alfredo L. Porfirio-Sousa:** Conceptualization, Data curation, Visualization, Formal analysis,  
390        Investigation, Writing – original draft, Writing – review & editing. **Robert E. Jones:** Investigation, Data  
391        curation, Writing – review & editing. **Matthew W. Brown:** Investigation, Writing – review & editing.  
392        **Daniel Lahr:** Investigation, Writing – review & editing. **Alexander K. Tice:** Conceptualization, Data  
393        curation, Visualization, Formal analysis, Investigation, Writing – original draft, Writing – review & editing,  
394        Resources, Supervision, Funding acquisition.

395        **Acknowledgments**

396        The authors acknowledge the High Performance Computing Center (HPCC) at Texas Tech University  
397        for providing computational resources that have contributed to the research results reported within this  
398        paper. URL: <http://www.hpcc.ttu.edu>

399        **Declaration of generative AI in scientific writing**

400        During the preparation of this work the authors used ChatGPT GPT-4o to improve the readability  
401        and language of the first draft of the manuscript. After using this tool, the authors reviewed and edited  
402        the content as needed and take full responsibility for the content of the published article.

403        **Ethics approval and consent to participate:** Not applicable

404        **Consent for publication:** Not applicable

405        **References**

406        1. Lewin HA, Robinson GE, Kress WJ, Baker WJ, Coddington J, Crandall KA, et al. Earth BioGenome  
407        Project: Sequencing life for the future of life. *Proceedings of the National Academy of Sciences*.  
408        2018;115:4325–33. <https://doi.org/10.1073/pnas.1720115115>.

409        2. Gao X, Chen K, Xiong J, Zou D, Yang F, Ma Y, et al. The P10K database: a data portal for the  
410        protist 10 000 genomes project. *Nucleic Acids Research*. 2024;52:D747–55.  
411        <https://doi.org/10.1093/nar/gkad992>.

412        3. Brown MW, Tice AK. A genetic toolbox for marine protists. *Nat Methods*. 2020;17:469–70.  
413        <https://doi.org/10.1038/s41592-020-0794-z>.

414        4. Schoenle A, Francis O, Archibald JM, Burki F, Vries J de, Dumack K, et al. Protist genomics: key to  
415        understanding eukaryotic evolution. *Trends in Genetics*. 2025;0.  
416        <https://doi.org/10.1016/j.tig.2025.05.004>.

417 5. Burki F, Roger AJ, Brown MW, Simpson AGB. The New Tree of Eukaryotes. *Trends in Ecology &*  
418 *Evolution*. 2020;35:43–55. <https://doi.org/10.1016/j.tree.2019.08.008>.

419 6. Williamson K, Eme L, Baños H, McCarthy CGP, Susko E, Kamikawa R, et al. A robustly rooted tree  
420 of eukaryotes reveals their excavate ancestry. *Nature*. 2025;640:974–81.  
421 <https://doi.org/10.1038/s41586-025-08709-5>.

422 7. López-García P, Moreira D. The Syntrophy hypothesis for the origin of eukaryotes revisited. *Nat*  
423 *Microbiol*. 2020;5:655–67. <https://doi.org/10.1038/s41564-020-0710-4>.

424 8. Keeling PJ, Burki F, Wilcox HM, Allam B, Allen EE, Amaral-Zettler LA, et al. The Marine Microbial  
425 Eukaryote Transcriptome Sequencing Project (MMETSP): Illuminating the Functional Diversity of  
426 Eukaryotic Life in the Oceans through Transcriptome Sequencing. *PLOS Biology*. 2014;12:e1001889.  
427 <https://doi.org/10.1371/journal.pbio.1001889>.

428 9. The Darwin Tree of Life Project Consortium. Sequence locally, think globally: The Darwin Tree of  
429 Life Project. *Proceedings of the National Academy of Sciences*. 2022;119:e2115642118.  
430 <https://doi.org/10.1073/pnas.2115642118>.

431 10. Leebens-Mack JH, Barker MS, Carpenter EJ, Deyholos MK, Gitzendanner MA, Graham SW, et  
432 al. One thousand plant transcriptomes and the phylogenomics of green plants. *Nature*.  
433 2019;574:679–85. <https://doi.org/10.1038/s41586-019-1693-2>.

434 11. Lahr DJ. An emerging paradigm for the origin and evolution of shelled amoebae, integrating  
435 advances from molecular phylogenetics, morphology and paleontology. *Mem Inst Oswaldo Cruz*.  
436 2021;116:e200620. <https://doi.org/10.1590/0074-02760200620>.

437 12. Ribeiro GM, Lahr DJG. Survival in a Changing World: The role of transcriptomics and the urgent  
438 need for genomes to understand Arcellinida's adaptive capabilities. *Acta Protozoologica*. 2025;2024  
439 Volume 63, Special Issue / Early View.

440 13. Kang S, Tice AK, Spiegel FW, Silberman JD, Pánek T, Čepička I, et al. Between a Pod and a  
441 Hard Test: The Deep Evolution of Amoebae. *Molecular Biology and Evolution*. 2017;34:2258–70.  
442 <https://doi.org/10.1093/molbev/msx162>.

443 14. Onsbring H, Tice AK, Barton BT, Brown MW, Ettema TJG. An efficient single-cell transcriptomics  
444 workflow for microbial eukaryotes benchmarked on *Giardia intestinalis* cells. *BMC Genomics*.  
445 2020;21:448. <https://doi.org/10.1186/s12864-020-06858-7>.

446 15. Porfirio-Sousa AL, Tice AK, Morais L, Ribeiro GM, Blandenier Q, Dumack K, et al. Amoebozoan  
447 testate amoebae illuminate the diversity of heterotrophs and the complexity of ecosystems throughout  
448 geological time. *Proceedings of the National Academy of Sciences*. 2024;121:e2319628121.  
449 <https://doi.org/10.1073/pnas.2319628121>.

450 16. Parks DH, Imelfort M, Skennerton CT, Hugenholtz P, Tyson GW. CheckM: assessing the quality  
451 of microbial genomes recovered from isolates, single cells, and metagenomes. *Genome Res*.  
452 2015;25:1043–55. <https://doi.org/10.1101/gr.186072.114>.

453 17. Astashyn A, Tvedte ES, Sweeney D, Sapochnikov V, Bouk N, Joukov V, et al. Rapid and sensitive  
454 detection of genome contamination at scale with FCS-GX. *Genome Biol*. 2024;25:60.  
455 <https://doi.org/10.1186/s13059-024-03198-7>.

456 18. Eme L, Sharpe SC, Brown MW, Roger AJ. On the Age of Eukaryotes: Evaluating Evidence from  
457 Fossils and Molecular Clocks. *Cold Spring Harb Perspect Biol*. 2014;6:a016139.  
458 <https://doi.org/10.1101/cshperspect.a016139>.

459 19. Guillou L, Bachar D, Audic S, Bass D, Berney C, Bittner L, et al. The Protist Ribosomal Reference  
460 database (PR2): a catalog of unicellular eukaryote Small Sub-Unit rRNA sequences with curated  
461 taxonomy. *Nucleic Acids Research*. 2013;41:D597–604. <https://doi.org/10.1093/nar/gks1160>.

462 20. Ribeiro GM, Useros F, Dumack K, González-Miguéns R, Siemensma F, Porfírio-Sousa AL, et al.  
463 Expansion of the cytochrome C oxidase subunit I database and description of four new lobose testate  
464 amoebae species (Amoebozoa; Arcellinida). European Journal of Protistology. 2023;91:126013.  
465 <https://doi.org/10.1016/j.ejop.2023.126013>.

466 21. Kosakyan A, Gomaa F, Lara E, Lahr DJG. Current and future perspectives on the systematics,  
467 taxonomy and nomenclature of testate amoebae. European Journal of Protistology. 2016;55:105–17.  
468 <https://doi.org/10.1016/j.ejop.2016.02.001>.

469 22. González-Miguéns R, Todorov M, Blandenier Q, Duckert C, Porfirio-Sousa AL, Ribeiro GM, et al.  
470 Deconstructing *Diffugia*: The tangled evolution of lobose testate amoebae shells (Amoebozoa:  
471 Arcellinida) illustrates the importance of convergent evolution in protist phylogeny. Molecular  
472 Phylogenetics and Evolution. 2022;175:107557. <https://doi.org/10.1016/j.ympev.2022.107557>.

473 23. Lahr DJG, Kosakyan A, Lara E, Mitchell EAD, Morais L, Porfirio-Sousa AL, et al. Phylogenomics  
474 and Morphological Reconstruction of Arcellinida Testate Amoebae Highlight Diversity of Microbial  
475 Eukaryotes in the Neoproterozoic. Current Biology. 2019;29:991–1001.e3.  
476 <https://doi.org/10.1016/j.cub.2019.01.078>.

477 24. Schuler GA, Brown MW. Description of *Armaparvus languidus* n. gen. n. sp. Confirms  
478 Ultrastructural Unity of Cutosea (Amoebozoa, Evosea). Journal of Eukaryotic Microbiology.  
479 2019;66:158–66. <https://doi.org/10.1111/jeu.12640>.

480 25. Burki F, Sandin MM, Jamy M. Diversity and ecology of protists revealed by metabarcoding.  
481 Current Biology. 2021;31:R1267–80. <https://doi.org/10.1016/j.cub.2021.07.066>.

482 26. Kosakyan A. Towards testate amoebae genomics and beyond, a wish list.... Acta Protozoologica.  
483 2025;2024 Volume 63, Special Issue:41–7.

484 27. Francois CM, Durand F, Figuet E, Galtier N. Prevalence and Implications of Contamination in  
485 Public Genomic Resources: A Case Study of 43 Reference Arthropod Assemblies. G3 (Bethesda).  
486 2020;10:721–30. <https://doi.org/10.1534/g3.119.400758>.

487 28. Cornet L, Baurain D. Contamination detection in genomic data: more is not enough. Genome Biol.  
488 2022;23:60. <https://doi.org/10.1186/s13059-022-02619-9>.

489 29. Tice AK, Žihala D, Pánek T, Jones RE, Salomaki ED, Nenarokov S, et al. PhyloFisher: A  
490 phylogenomic package for resolving eukaryotic relationships. PLOS Biology. 2021;19:e3001365.  
491 <https://doi.org/10.1371/journal.pbio.3001365>.

492 30. Yang Z, Rannala B. Molecular phylogenetics: principles and practice. Nat Rev Genet.  
493 2012;13:303–14. <https://doi.org/10.1038/nrg3186>.

494 31. Emms DM, Kelly S. SHOOT: phylogenetic gene search and ortholog inference. Genome Biology.  
495 2022;23:85. <https://doi.org/10.1186/s13059-022-02652-8>.

496 32. Smith SA, Pease JB. Heterogeneous molecular processes among the causes of how sequence  
497 similarity scores can fail to recapitulate phylogeny. Brief Bioinform. 2017;18:451–7.  
498 <https://doi.org/10.1093/bib/bbw034>.

499

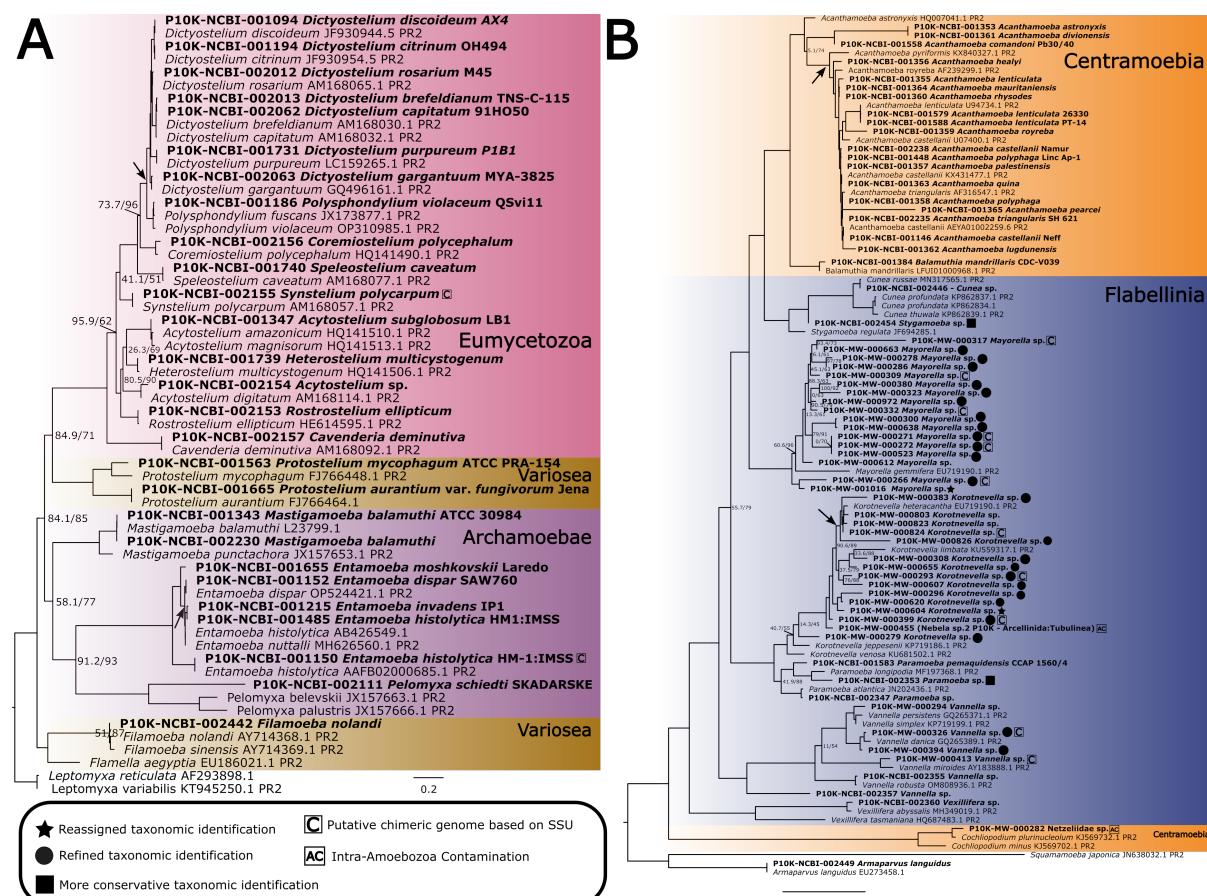
500

501

502

503

504 **Figures**



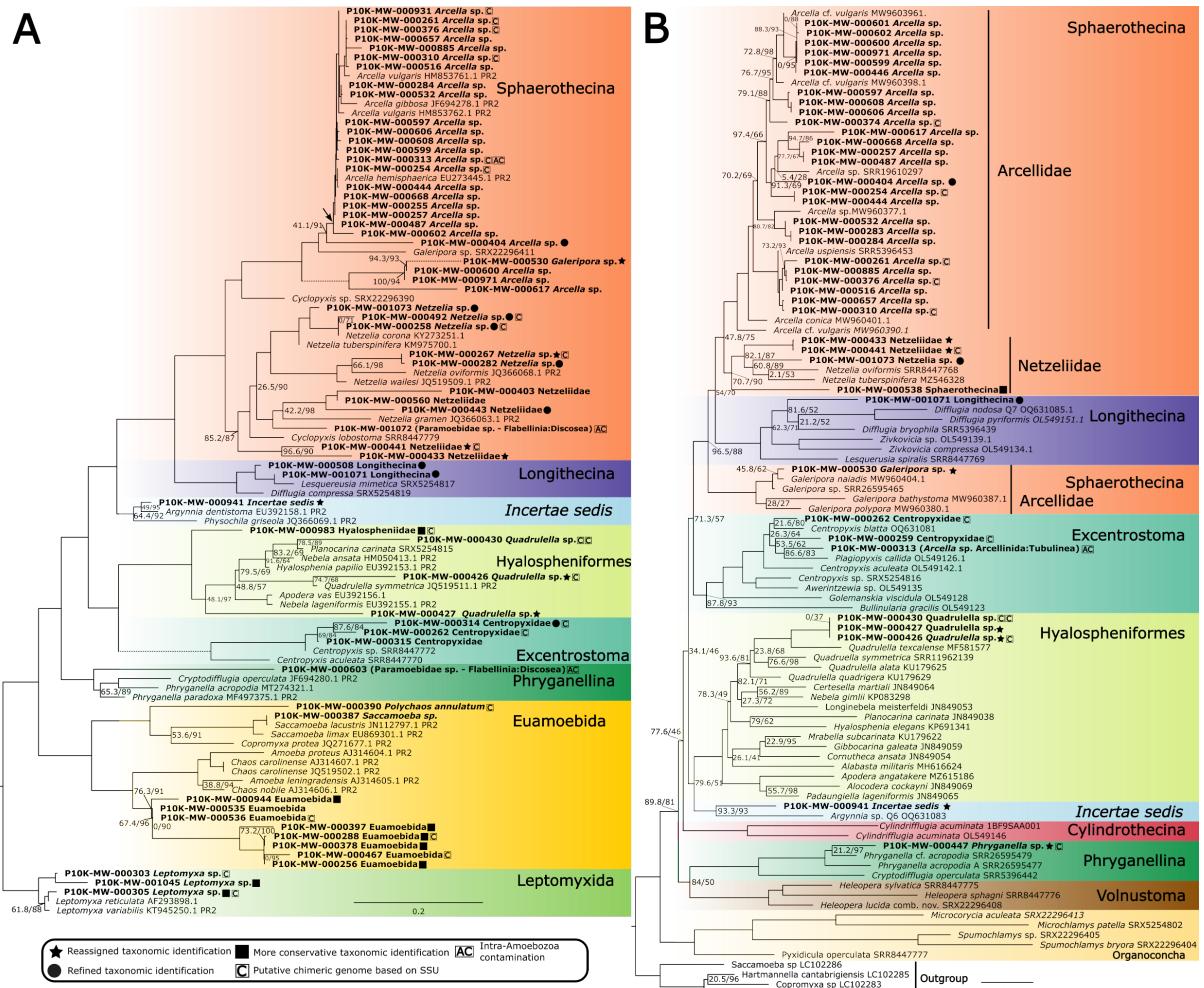
505

506 **Figure 1. The phylogenetic placement of Amoebozoa genomic data from the P10K focused on**  
507 **the major groups Evosea and Discosea.** Maximum-likelihood phylogenetic trees constructed from  
508 the Small Subunit ribosomal RNA (SSU) inferred from a subset of the curated dataset presented in  
509 Figure S1. **A.** Focuses on the major Amoebozoa group Evosea. Phylogenetic reconstruction was  
510 conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model  
511 (TIM2+F+R4). **B.** Focuses on the major Amoebozoa group Discosea. Phylogenetic reconstruction was  
512 conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model  
513 (GTR+F+G4). Node support was assessed using both ultrafast bootstrap (UFBoot) and the  
514 Shimodaira-Hasegawa approximate likelihood ratio test (SH-aLRT). Support values are reported as  
515 SH-aLRT / UFBoot, with values  $\geq 80/95$  considered indicative of strong support. For clarity, high-support  
516 values are omitted in this figure, as well as support values for nodes above the nodes indicated by the  
517 arrows, which are represented mostly by flat branches. The complete tree, including all support values,  
518 is shown in Figures S2 and S3. Stars indicate genomic data reassigned to a different taxonomic identity  
519 than reported in the P10K database. Filled circles mark indicate cases with refined taxonomic  
520 resolution. Filled squares denote more conservative identifications (e.g., genus or family level) rather  
521 than the more specific genus or species-level identification originally provided in the P10K database.  
522 'C' indicates putative chimeric genomes containing sequences from multiple eukaryotes and 'AC'  
523 denotes intra-Amoebozoa contamination.

524

525

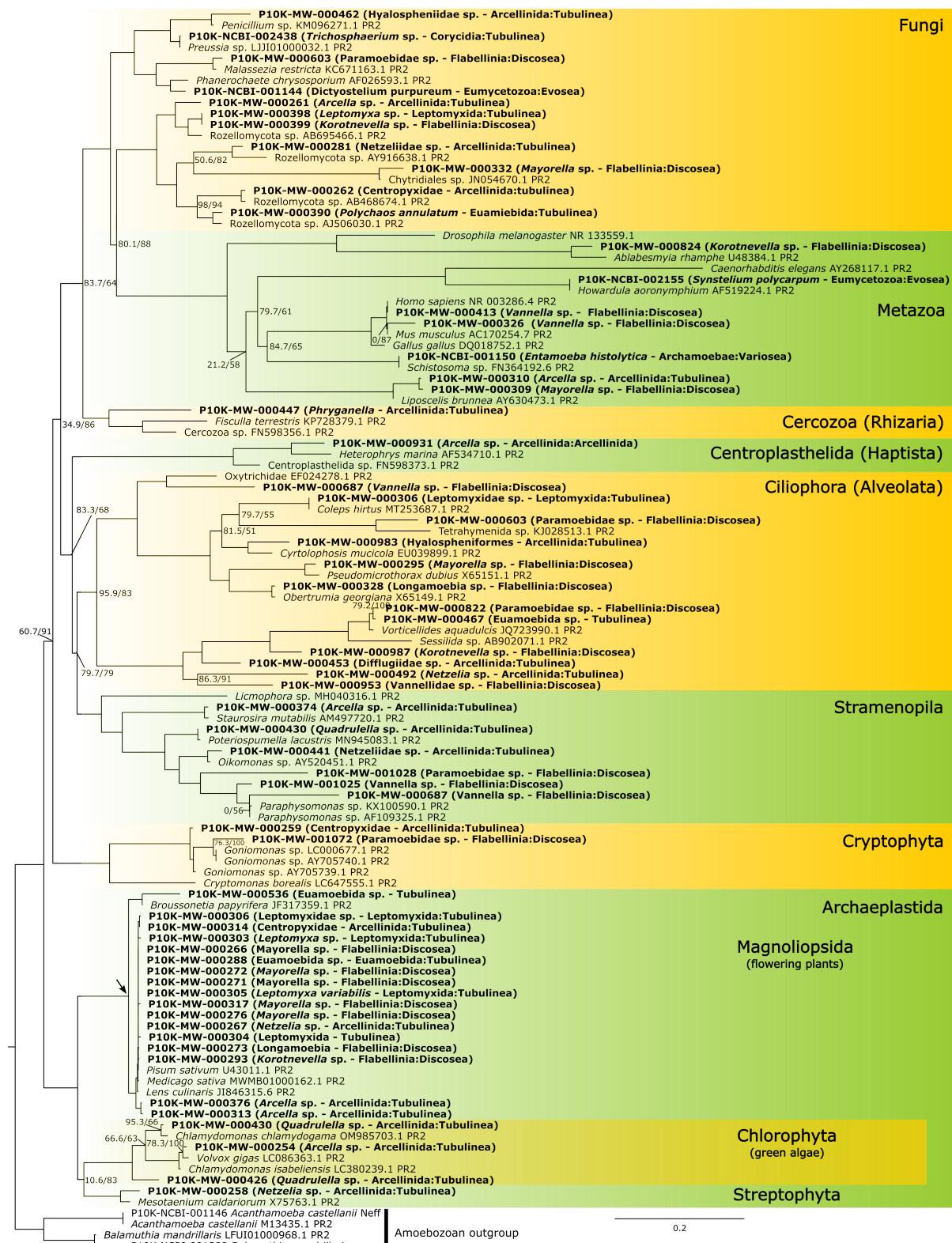
526



527

528 **Figure 2. The phylogenetic placement of Amoebozoa genomic data from the P10K focused on**  
 529 **the major group Tubulinea. A. The maximum-likelihood phylogenetic tree constructed from the Small**  
 530 **Subunit ribosomal RNA (SSU) inferred from a subset of the curated dataset presented in Figure S1,**  
 531 **focuses on the major Amoebozoa group Tubulinea. Phylogenetic reconstruction was conducted using**  
 532 **IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model (TIM3e+G4). The length of**  
 533 **branches depicted as dashed lines have been reduced by 50% for presentation purposes. B. The**  
 534 **maximum-likelihood phylogenetic tree of cytochrome c oxidase subunit I (COI) inferred from a curated**  
 535 **dataset generated in the present study, focusing on Arcellinida order (Tubulinea:Amoebozoa)**  
 536 **comprising COI sequences retrieved from genomes and transcriptomes available in the P10K**  
 537 **database, along with reference sequences made available by previous. Phylogenetic reconstruction**  
 538 **was conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model**  
 539 **(GTR+F+I+G4). Node support was assessed using both ultrafast bootstrap (UFBoot) and the**  
 540 **Shimodaira-Hasegawa approximate likelihood ratio test (SH-aLRT). Support values are reported as**  
 541 **SH-aLRT / UFBoot, with values  $\geq 80/95$  considered indicative of strong support. For clarity, high-support**  
 542 **values are omitted in this figure, as well as support values for nodes above the one indicated by the**  
 543 **arrow, which are represented mostly by flat branches. The complete tree, including all support values,**  
 544 **is shown in Figures S4 and S5. Stars indicate genomic data reassigned to a different taxonomic identity**  
 545 **than reported in the P10K database. Filled circles mark indicate cases with refined taxonomic**  
 546 **resolution. Filled squares denote more conservative identifications (e.g., genus or family level) rather**  
 547 **than the more specific genus or species-level identification originally provided in the P10K database.**  
 548 **'C' indicates putative chimeric genomes containing sequences from multiple eukaryotes and 'AC'**  
 549 **denotes intra-Amoebozoa contamination.**

550



552 **Figure 3. Contamination-screening phylogenetic tree.** The maximum-likelihood phylogenetic tree  
 553 constructed from the Small Subunit ribosomal RNA (SSU), inferred from a curated dataset generated  
 554 in the present study, comprises SSU sequences of putative non-Amoebozoan contaminants retrieved  
 555 from genomes and transcriptomes available in the P10K database, along with reference sequences  
 556 from the PR<sup>2</sup> database (indicated by PR<sup>2</sup> IDs), identified through BLAST+ similarity searches using the  
 557 putative contaminant SSU sequences as queries. Phylogenetic reconstruction was conducted using  
 558 IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model (TN+G+F+I+G4). Node

559 support was assessed using both ultrafast bootstrap (UFBoot) and the Shimodaira–Hasegawa  
560 approximate likelihood ratio test (SH-aLRT). Support values are reported as SH-aLRT / UFBoot, with  
561 values  $\geq 80/95$  considered indicative of strong support. For clarity, high-support values are omitted in  
562 this figure, as well as support values for nodes above the one indicated by the arrow, which are  
563 represented mostly by flat branches. The complete tree, including all support values, is shown in Figure  
564 S6. The taxonomic identification of Amoebozoa taxa corresponding to the displayed P10K ID codes is  
565 shown in parentheses.

566

567

568

569

570

571

572

573

574

575

576

577

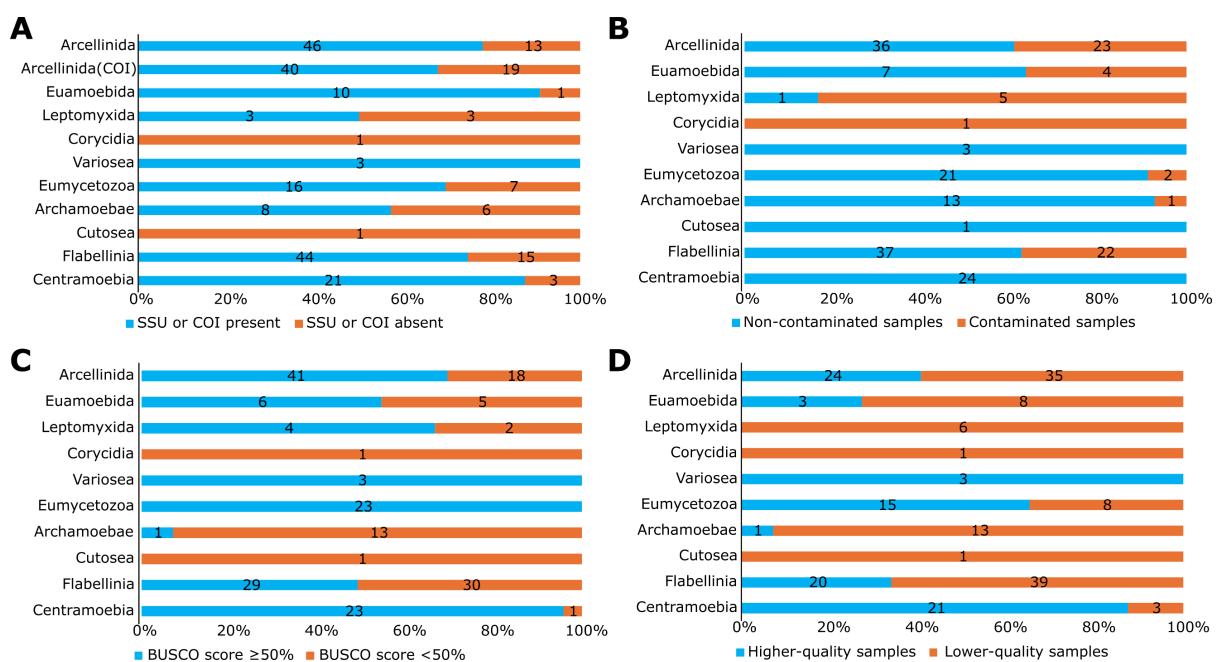
578

579

580

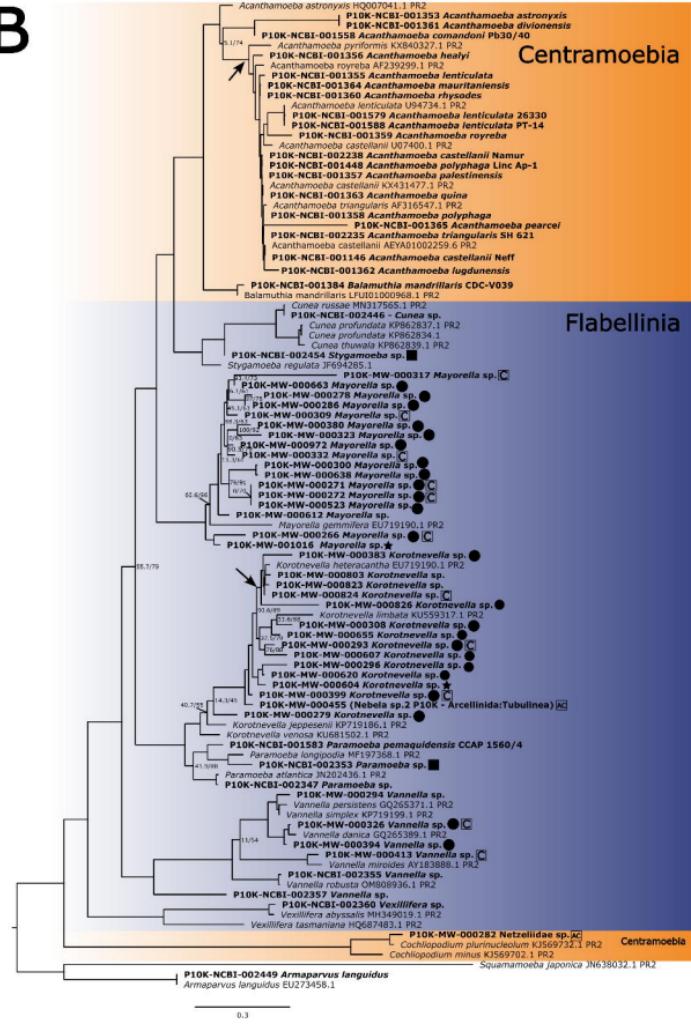
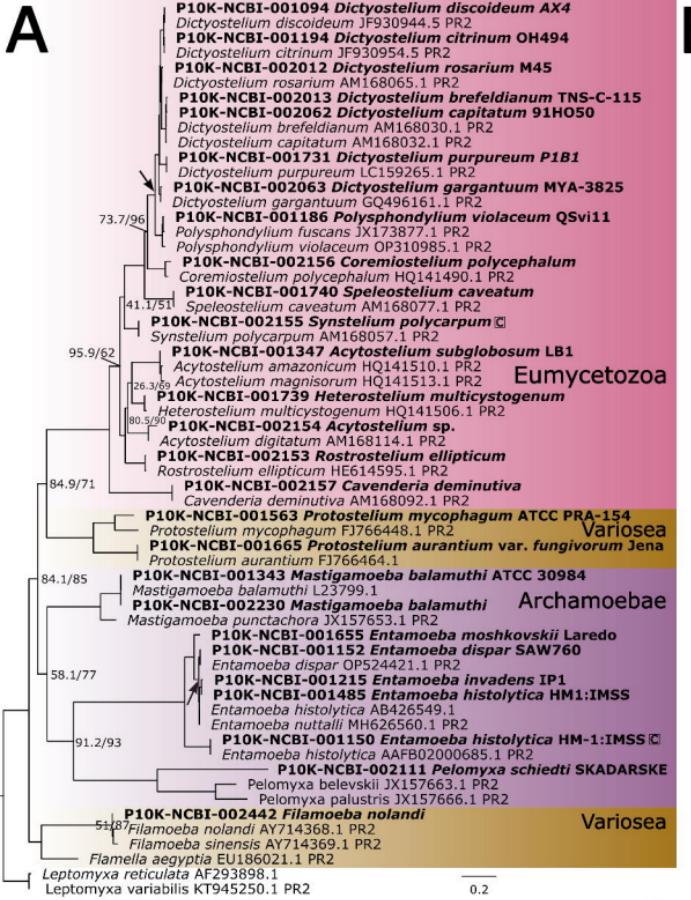
581

582

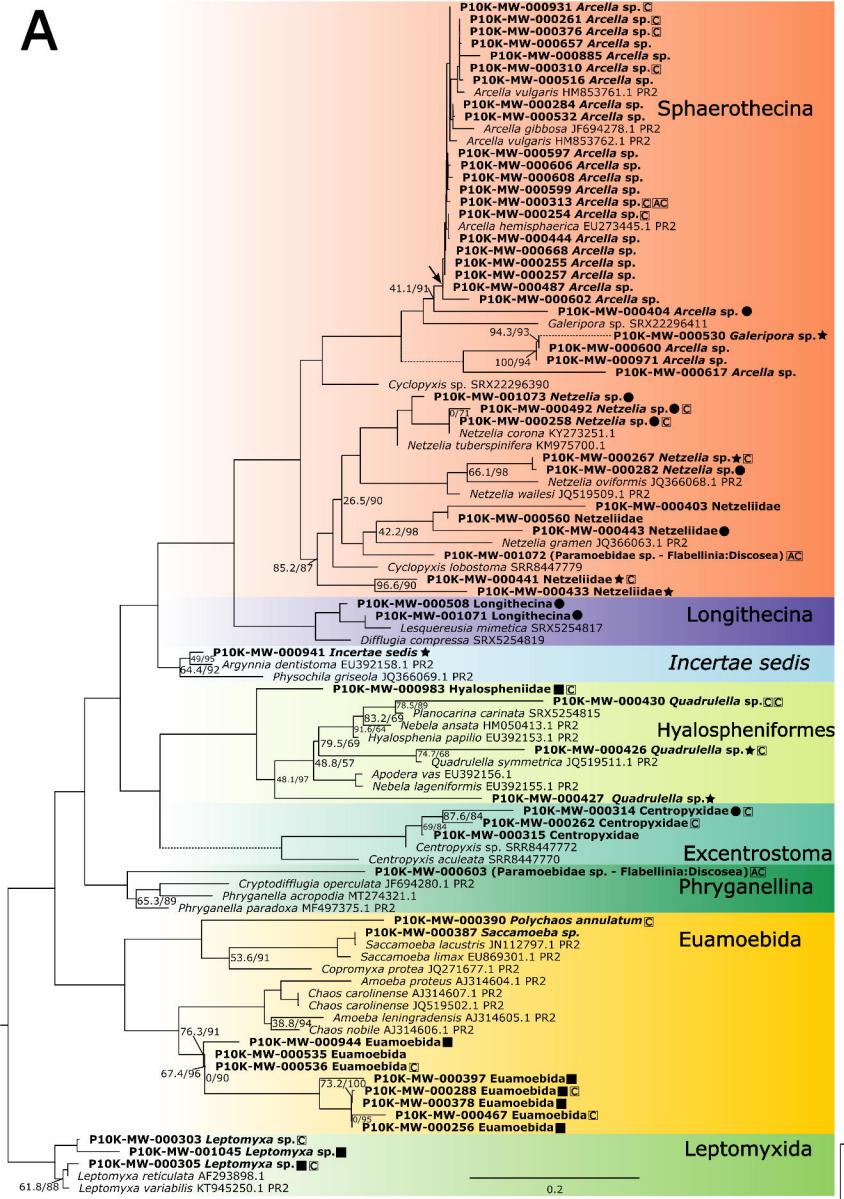


583

584 **Figure 4. Summary of Amoebozoa P10K database sample counts by taxonomic group, focusing**  
 585 **on the key parameters used to evaluate the data. A.** presence of SSU or COI. **B.** BUSCO  
 586 completeness score. We considered the BUSCO score of each sample as originally provided in the  
 587 P10K database and reported in Gao et al. (2024). **C.** contamination by another eukaryotic lineage. **D.**  
 588 sample quality assessment. Higher-quality samples are defined as those with SSU (and COI in the  
 589 case of Arcellinida), a BUSCO score  $\geq 50\%$ , and no contamination identified based on the SSU. Lower-  
 590 quality samples are those that require decontamination or further sequencing prior to reliable  
 591 downstream analysis. The bars represent the total number of Amoebozoa samples in the P10K  
 592 database available for each taxonomic group. The samples considered as higher quality are highlighted  
 593 in the Supplementary Information – Table S1.

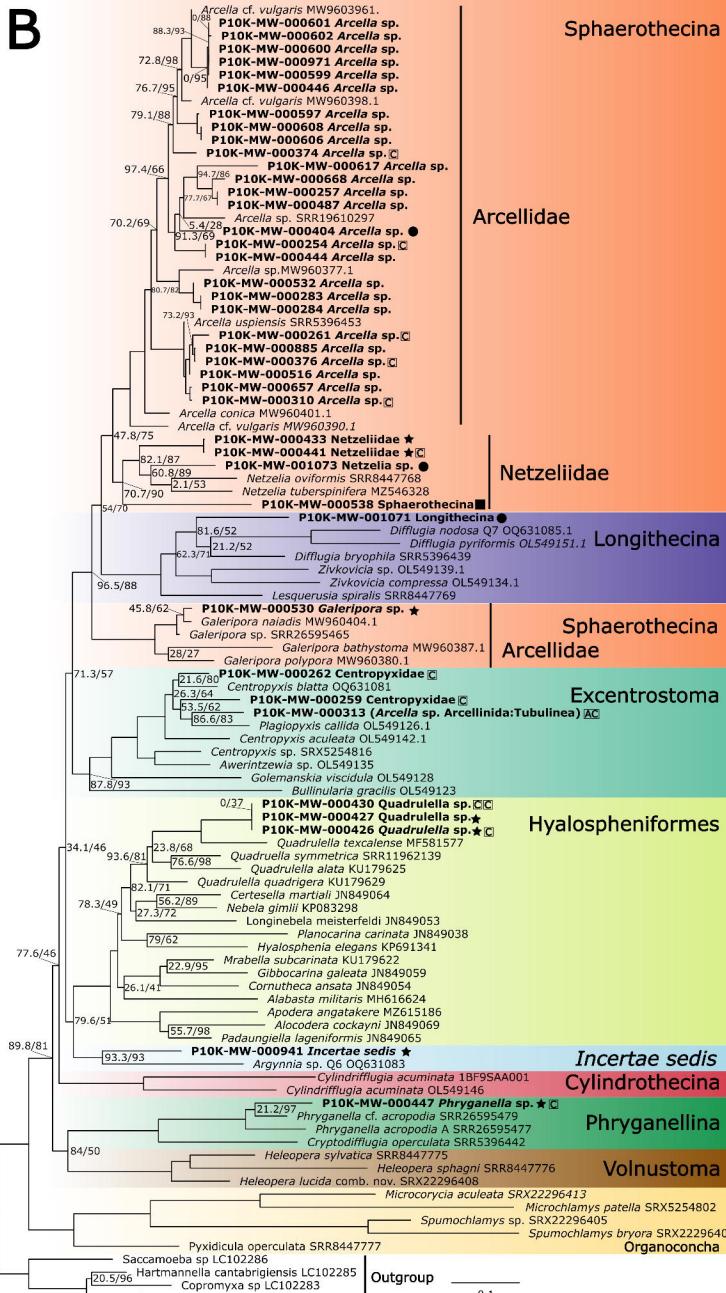


A



★ Reassigned taxonomic identification ■ More conservative taxonomic identification AC Intra-Amoebozoa contamination  
 ● Refined taxonomic identification C Putative chimeric genome based on SSU

B



Sphaerothecina

Arcellidae

Netzeliidae

Longithecina

Sphaerothecina  
Arcellidae

Excentrostoma

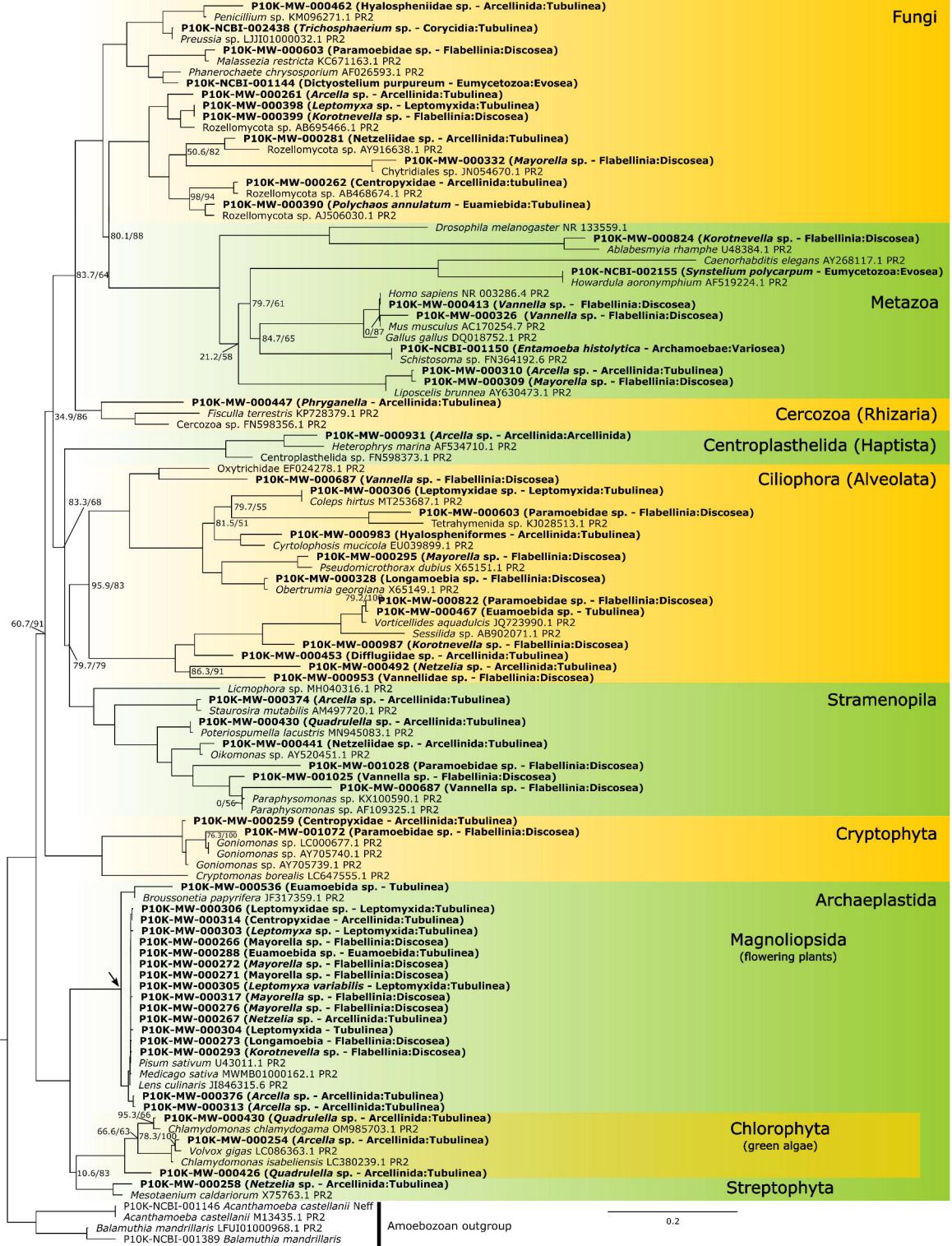
Hyalospheniformes

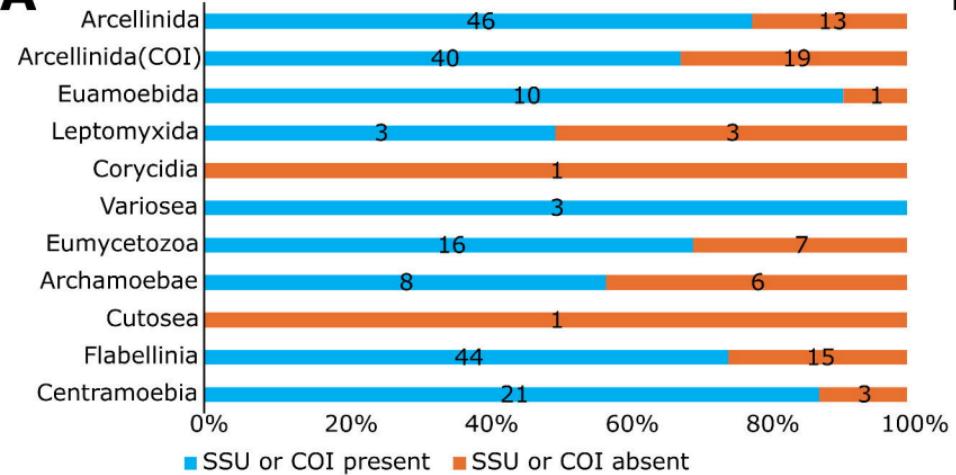
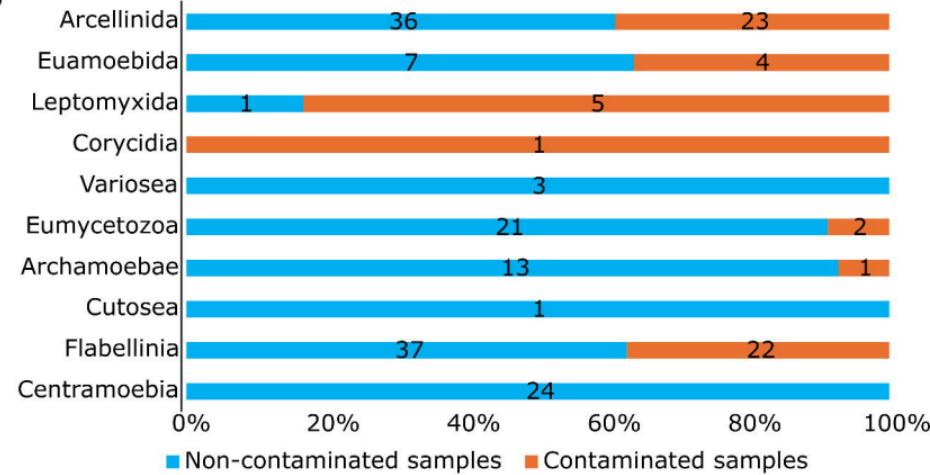
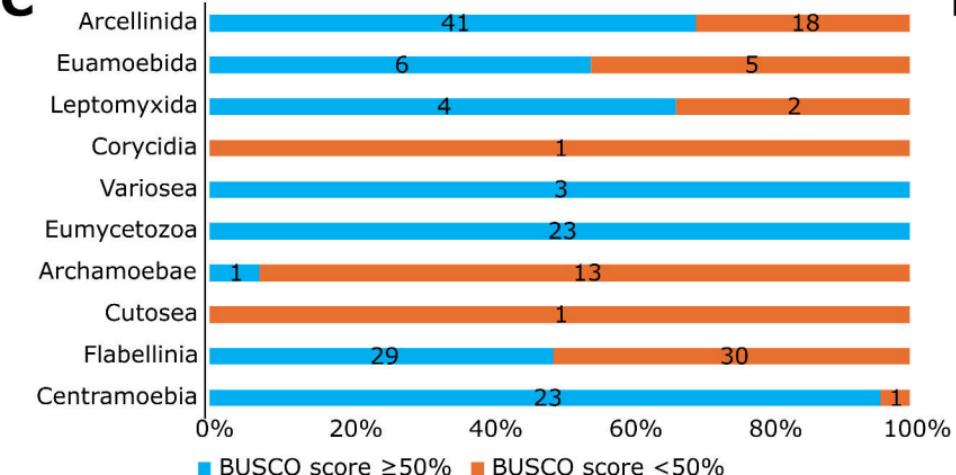
Incertae sedis  
Cylindrothecina

Phryganellina

Volnustoma

Organoncocha



**A****B****C****D**