Research article

Title: CSI-SSU: Phylogenetic contamination screening of genomic datasets, demonstrated on the

Protist 10,000 Genomes (P10K) database

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Abstract

Background: Genomic data are essential for uncovering the evolutionary history, ecological

roles, and diversity of life. Yet, diverse microbial eukaryotes, predominantly unicellular and

traditionally referred to as protists, remain critically underrepresented in genomic repositories, limiting

our ability to address fundamental questions in eukaryotic evolution. The Protist 10,000 Genomes

(P10K) initiative seeks to fill this gap by generating and compiling genomic and transcriptomic data for

a wide range of microbial eukaryotes. However, large-scale sequencing efforts face persistent

challenges, including contamination and imprecise taxonomic identification, particularly for poorly

studied taxa that require specialized taxonomic expertise. To ensure the reliability of these resources,

robust and scalable approaches for taxonomic identification and contamination screening are

essential.

Results: We developed CSI-SSU (https://github.com/AlexTiceLab/CSI-SSU), a command-line

tool for Contaminant Sequence Investigation (CSI) that uses small subunit ribosomal RNA (SSU)

sequences, chimeric sequence detection, and phylogenetic placement to rapidly identify, retrieve, and

classify SSU sequences from eukaryotic genomic-level assemblies. CSI-SSU incorporates a curated

SSU reference dataset representing the major known eukaryotic supergroups, with sequences and

taxonomic nomenclature derived from the Protist Ribosomal Reference (PR2) database. In addition to

detecting contaminant sequences, CSI-SSU enables approximate taxonomic assignment of the target

lineage in each assembly, with resolution constrained by the current diversity represented in PR2. To

further assess potential bacterial contamination, CSI-SSU employs bacterial BUSCO searches as a

proxy. We demonstrate CSI-SSU utility and performance by screening 2,960 genomic-level

assemblies spanning a broad diversity of eukaryotes from P10K. CSI-SSU efficiently detected non-

target eukaryotic SSU sequences, revealing cross-group contamination. Classifications also

corroborated or refined the original taxonomic assignments, with resolution depending on PR2

representation. Bacterial BUSCO searches indicated bacterial contamination. Independent SSU and

COI phylogenies of Amoebozoa supported CSI-SSU classifications, highlighting its accuracy and

sensitivity.

Conclusion: CSI-SSU provides a scalable and reproducible framework for phylogenetically

informed contamination screening and taxonomic validation of genomic and transcriptomic data.

Coupling phylogenetic placement with contamination detection enabled us to distinguish high-quality

P10K datasets from those requiring decontamination or additional sequencing before downstream

use. These findings serve as a reference for future analyses and guide further sequencing efforts to

expand the taxonomic diversity of microbial eukaryotes at the genomic level. Addressing imprecise

taxonomic assignments, contamination, and reproducibility in genomic-level datasets will enhance the

value of these resources and facilitate studies illuminating the evolution and diversification of

eukaryotic life.

Keywords: Amoebozoa, Arcellinida, Eukaryotic diversity, Comparative genomics, Genomic

database curation

Introduction

Genomic-level data are essential for advancing our understanding of the evolution of life on

Earth [1, 2]. High-quality genome and transcriptome sequences enable comparative analyses that

reveal patterns of genomic evolution, including gene family expansions, horizontal gene transfers,

and changes in genome organization and regulatory systems [3, 4]. Such data also clarify

phylogenetic relationships through multi-gene and genome-scale reconstructions [5, 6]. Moreover,

genomic analyses uncover ecological interactions by identifying metabolic pathways, symbiotic

associations, and genetic adaptations to specific environmental conditions [7]. Consequently,

genomics has become central to studying life's diversity and complexity.

While genomics has progressed rapidly for major eukaryotic groups such as plants, animals,

fungi, and other traditional model systems, most microbial eukaryotes (commonly referred to as

protists) remain vastly underrepresented [1, 2]. Although they comprise most of eukaryotic diversity,

the majority of lineages within this highly diverse paraphyletic assemblage still lack genomic data [2,

5]. Over the past decade, several initiatives have contributed to fill this gap, such as the Marine

Microbial Eukaryote Transcriptome Sequencing Project (MMETSP) for marine microbial genomics [8],

the Tree of Life Programme for eukaryotic genome sequencing [9], and the One Thousand Plant

Transcriptomes initiative (1KP) for plants, including single-celled algae [10]. Similarly, the Protist

10,000 Genomes (P10K) initiative aims to address the underrepresentation of microbial eukaryotes in

genomic databases by generating new genomic data (i.e., genomes and transcriptomes) and

compiling previously available data from a wide array of lineages, coupled with taxonomic

identification and decontamination procedures [2]. This large-scale effort, consolidated in the P10K

database, represents an unprecedented genomic resource for protists and potentially provides a

valuable foundation for achieving a comprehensive and integrative view of eukaryotic evolution.

However, to fully exploit the potential of this newly available genomic resource for protists, three

key challenges must be considered: (1) accurate taxonomic identification, (2) the presence of

contaminated genomic data resulting from non-target eukaryotic contamination, and (3) data

interpretation and reproducibility [11, 12]. Specifically, the taxonomic identification of the P10K

database relies on small subunit ribosomal DNA (SSU rDNA) retrieval and BLAST similarity searches

against curated SSU databases, including SILVA, PR^2, and NCBI's NT (nucleotide sequence)

database [2]. Although informative, a BLAST-based approach is less accurate than phylogeny-based

methods for taxonomic identification and may lead to a less precise classification of the target groups.

Regarding contamination, protists typically inhabit environments with high eukaryotic microbial

diversity, several prey on other organisms, many host endosymbionts, and most are often difficult to

isolate as monoeukaryotic cultures or single cells [13–15]. As a result, genomic-level sequencing

efforts might produce contaminated assemblies that contain not only the genome of the target

organism but also sequences from other eukaryotic organisms [16, 17]. While P10K applies a

decontamination strategy to remove bacterial, archaeal, viral, fungal, and other eukaryotic

contaminants from ciliate data, for other protist groups only bacterial, archaeal, and viral contaminants

are filtered [2]. Finally, the P10K database currently lacks photo-documentation and detailed

information, for instance, about the sequencing platforms used for each sample (i.e., Illumina, Oxford

Nanopore, or PacBio Sequel II), which are essential for deeper interpretation and reproducibility of the

data. In this context, accurate taxonomic identification, assessment of eukaryotic contamination, and

the availability of more detailed sample information are most essential for the effective downstream

use of data in the P10K database.

Here, we introduce CSI-SSU, a command-line tool for Contaminant Sequence Investigation

(CSI) that uses small subunit ribosomal RNA (SSU) sequences, chimeric sequence detection via

VSEARCH, bacterial BUSCO searches, and phylogenetic placement via pplacer to rapidly identify,

classify, and screen eukaryotic genomic resources. We applied CSI-SSU to the Protist 10,000

Genomes (P10K) database to evaluate taxonomic identification and contamination across diverse

eukaryotic lineages. Incorporating a curated SSU reference dataset representing the major known

eukaryotic supergroups, CSI-SSU detects non-target and chimeric sequences, enables approximate

taxonomic assignment of assemblies, and assesses potential bacterial contamination using BUSCO

searches. Phylogenetic analyses focused on Amoebozoa as a case study, using SSU and

cytochrome c oxidase subunit I (COI) markers, validated CSI-SSU classifications and identification of

high-quality assemblies from those requiring decontamination or additional sequencing. This

framework provides a scalable, reproducible approach for improving the reliability, usability, and

interpretability of large genomic datasets, offering a proof-of-concept for phylogenetically informed

contamination screening and taxonomic validation across all major eukaryotic lineages using CSI-

SSU tool (https://github.com/AlexTiceLab/CSI-SSU).

Methods

CSI-SSU tool

The full mode of the CSI-SSU tool performs automated SSU (18S) screening on genomic-level

data using BLAST v2.17.0, MAFFT v7.526, pplacer v1.1.alpha20, and VSEARCH v2.30.5 [18–21]

(**Figure S1**). CSI-SSU requires a single input: a nucleotide assembly (e.g., genome, transcriptome,

metagenome, or metatranscriptome) to be screened. To identify and retrieve SSU sequences from

target assemblies, CSI-SSU implements a similarity search using BLAST+ v2.17.0, based on a

curated reference SSU query dataset. Query sequences were selected from the Protist Ribosomal

Reference (PR2) database to capture the broad diversity of major eukaryotic lineages

(https://github.com/AlexTiceLab/CSI-SSU/tree/main/csi_ssu/data/queries). For each input assembly,

CSI-SSU first constructs a local BLAST database using the *makeblastdb* command. The reference

SSU queries are then used in a *blastn* search to identify candidate SSU sequences within the

assembly. BLAST results are subsequently parsed, and SSU sequences are retrieved using the

Python script *parse_blast.py*, which extracts full subject sequences and corrects their orientation

based on BLAST coordinates. Once retrieved, SSU sequences are subjected to phylogenetic

placement.

To perform phylogenetic placement, CSI-SSU provides curated reference alignments and

phylogenetic trees for nine major eukaryotic supergroups (Amoebozoa, Excavata, TSAR,

Archaeplastida, Cryptista, Haptista, CRuMs, Provora, and Obazoa), derived from the PR2 database.

To construct these reference datasets, we downloaded the PR2 SSU database (version 5.1.0) on

September 9th, 2025. Redundancy was reduced using CD-HIT v4.8.1 with the parameter "-c 0.95,"

corresponding to a global similarity threshold of 95%. For each target supergroup, the reference

dataset included: (i) all sequences assigned to the focal supergroup in PR^2, (ii) sequences labeled as

Eukaryota_X, and (iii) five representative sequences from each class of all other supergroups

available in PR^2. Taxonomic assignments followed the PR2 classification as accessed on September

9th, 2025. For each supergroup-specific FASTA file, sequences were aligned using MAFFT *(--auto)*

and trimmed with trimAl v1.2 using a gap threshold of 0.3 (*trimal -gt 0.3*). Preliminary phylogenetic

trees were inferred using RAxML with 100 bootstrap replicates. These trees were manually inspected

to remove putative contaminants and long-branch sequences. The untrimmed, curated datasets were

then realigned using MAFFT (*--auto*) and trimmed again with trimAl (*-gt 0.3*). The resulting curated

alignments and phylogenetic trees constitute the reference datasets used for SSU sequence

placement.

CSI-SSU performs phylogenetic placement of retrieved SSU sequences using pplacer.

Retrieved sequences are added to the selected reference alignment using MAFFT (*mafft --add --*

keeplength), based on the user-selected focal supergroup. The extended alignment is then used for

phylogenetic placement onto the corresponding reference tree using pplacer likelihood (ML) mode.

Placement results are stored as SQLite databases. CSI-SSU uses the script *summarize_results.py* to

extract and summarize taxonomic classifications from these databases, generating a structured

summary output. For visualization, the script *plot_tree.py* is used to generate midpoint-rooted trees

with placed SSU sequences highlighted, using the ETE3 toolkit.

To identify putative chimeric sequences in large-scale datasets, CSI-SSU implements

VSEARCH v2.30.5 using the command *vsearch --uchime_ref*. As a reference database, we used the

PR2 SSU dataset (version 5.1.0) downloaded on September 9th, 2025, and clustered with CD-HIT at

95% similarity ("-c 0.95"). Based on this reference, CSI-SSU identifies putative chimeric sequences

among the retrieved SSU sequences. The *plot_tree.py* script also incorporates VSEARCH results,

highlighting the branches of detected chimeric sequences in the final tree visualization.

In addition to the full mode, CSI-SSU provides a placement mode for pre-extracted SSU

sequences and a retrieval mode to extract SSU sequences without phylogenetic placement. In the

placement mode, the user supplies a FASTA file containing SSU sequences, and the pipeline

proceeds directly to the phylogenetic placement and chimeric sequence identification steps, including

MAFFT (*--add*) alignment, pplacer placement, VSEARCH, and result summarization with the scripts

summarize_results.py and *plot_tree.py*, as described above. In the retrieval mode, the user supplies a

nucleotide assembly. SSU sequences are then identified and collected as detailed above, but

phylogenetic placement is skipped. This allows users to take the identified SSU sequences and

perform their own analyses.

Phylogenetic placement-based contamination screening of the P10K database

We implemented the CSI-SSU full mode to screen the complete Protist 10,000 Genomes (P10K)

database available as of September 9, 2025 (**Supplementary Information, Table S1**). To do so, we

downloaded all assemblies available in P10K for each of the major eukaryotic supergroups

represented in the database (i.e., Amoebozoa, Archaeplastida, Excavata, Haptista, Obazoa, and

TSAR), totaling 2,960 genome-level assemblies. For each assembly, we performed contamination

screening using CSI-SSU full mode, following the steps described above and the guidelines provided

in the CSI-SSU GitHub repository (https://github.com/AlexTiceLab/CSI-SSU). Analyses were

conducted separately for each supergroup, using the corresponding focal supergroup reference

alignment and phylogenetic tree provided with the CSI-SSU tool. The P10K database includes

assemblies representing the Cryptista supergroup within the Archaeplastida dataset. Therefore, for

Archaeplastida screening, we used CSI-SSU with Archaeplastida as the focal group and accounted

for this when parsing and presenting the results.

SSU and COI Phylogenetic reconstruction

To further investigate Amoebozoa as a case study and validate CSI-SSU, we performed

phylogenetic reconstructions of 201 amoebozoan taxa from the P10K database (Supplementary

Information, Table S1). This approach allows us to refine taxonomic identifications and to assess the

performance of CSI-SSU. We constructed a small subunit ribosomal (18S) dataset considering all 201

amoebozoan assemblies. Further, we constructed a cytochrome c oxidase subunit I (COI) dataset

focusing only on the testate amoebae order Arcellinida (Tubulinea), since this marker is well sampled

for this lineage as it has been traditionally used for phylogenetics in arcellinids [22, 23]. To construct

datasets for the 18S and COI markers, we extracted sequences from these assemblies using

similarity searches implemented in BLAST+ v2.16.0+ and a custom Python script (**Supplementary**

Information - File S1). For each marker, the script automated the creation of BLAST databases

using makeblastdb and performed local blastn searches for each query sequence in a multi-FASTA

file against each Amoebozoa genomic data. Specifically, the script executed the commands

makeblastdb -in P10KID.fasta -dbtype nucl and *blastn -query marker_query -db P10K.fasta -outfmt*

[script_default_choice] (**Supplementary Information - File S1**). Aiming for a manageable number of

SSU sequences for manual and visual curation, the script retrieved the top five hits per genomic data

file, extracting each aligned region along with 1,000 bp of upstream and downstream flanking

sequence. This approach enabled recovery of extended SSU and COI regions suitable for

downstream phylogenetic analyses and compatible with Amoebozoa SSU and COI data available in

the PR^2 database [24] and NCBI. The orientation of each retrieved sequence was assessed, and

sequences were reverse complemented when necessary to match the strand of the original query.

For the query sequences, we used 18S data from the PR^2 database for three representative species

of the major amoebozoan clades: *Arcella vulgaris* WP (Tubulinea; GenBank: HM853762.1),

Dictyostelium discoideum (Evosea; GenBank: AM168040.1), and *Acanthamoeba castellanii* Neff

(Discosea; GenBank: U07416.1). As the COI query sequence, we considered the sequence of *Arcella*

uspiensis (SRR5396453).

To build a phylogenetically informative datasets, we combined the retrieved sequences with

previously published 18S and COI datasets for Amoebozoa, as well as sequences from the PR^2

database for SSU [13, 15, 23, 24]. For SSU, we curated a non-redundant dataset broadly

representative of the major lineages in Amoebozoa (Tubulinea, Evosea, and Discosea) and for COI a

dataset broadly representative of the major lineages in Arcellinida, for both markers excluding

environmental sequences. This strategy ensured a robust and interpretable dataset for our

phylogenetic framework. From a preliminary phylogenetic reconstruction, we curated the SSU and

COI datasets used to generate the main trees in this study. This initial analysis allowed us to identify

and remove identical or highly similar sequences that resulted from retrieving the top five BLAST+ hits

per genome. It also enabled visual inspection of the tree to detect SSU sequences that either failed to

cluster within Amoebozoa or formed unusually long branches. Many of these were short sequences

(<200 bp) and were excluded from downstream analyses, while some long branches corresponded to

full-length sequences that likely represented contaminants. These were retained for further
contamination screening.

For contamination screening, we focused on the 18S dataset. After the initial phylogenetic
reconstruction (see *Phylogenetic reconstructions* section), any SSU sequences from P10K genome
assemblies that represented long branches or did not branch within the Amoebozoa clade were
selected for further analysis. These sequences were subjected to additional BLAST+ searches
against the PR^2 database, a curated SSU resource representing eukaryotic diversity, using our
custom script and the same parameters described above (**Supplementary Information - File S1**). To
perform this search locally, we downloaded the complete PR^2 database and used it as the reference
for the BLAST+ similarity search. This approach allowed us to retrieve SSU sequences from the PR^2
database that were similar to those of the putative contaminant eukaryotes and to assign their
taxonomic affiliations through subsequent phylogenetic analyses.

All phylogenetic reconstructions were based on multiple sequence alignments (MSAs) generated
using MAFFT v7.490 with the E-INS-I algorithm and 1000 refinement iterations. Alignments were
produced with the following command: *mafft --genadpair --maxiterate 1000 input.fasta >*
output_aligned.fasta. Automated alignment trimming was performed with trimAl v1.2, using the
command *trimal -in input_aligned.fasta -out output_aligned_trimmed.fasta -keepheader -gt*
[threshold], where the gap threshold was set to 0.3 for 18S and 0.5 for COI. Phylogenetic trees were
inferred from the trimmed alignments using the maximum likelihood method implemented in IQ-TREE
v2.3.6, with ModelFinder for model selection and node support assessed via 1,000 ultrafast bootstrap
replicates and 1,000 SH-aLRT tests. Support values of SH-aLRT/UFBoot ≥ 80/95 are considered
indicative of strong support, following the recommendations of the original papers describing the
development of these node support measures [31, 32]. The analysis was executed with the command
iqtree2 -s aligned_trimmed.fasta -alrt 1000 -bb 1000 -m TEST.

Results and discussion

The CSI-SSU tool

The full mode of CSI-SSU begins with a user-provided nucleotide genomic-level assembly in

FASTA format for a target organism (**Figure S1**). Based on the organism's supergroup, users can

select the appropriate reference group from those available in CSI-SSU (Amoebozoa, Excavata,

TSAR, Archaeplastida, Cryptista, Haptista, CRuMs, Provora, Obazoa), reflecting the current diversity

sampling and taxonomy of the PR2 database. In full mode, the tool performs SSU sequence

identification and retrieval, chimeric sequence detection, and phylogenetic placement onto a

reference SSU tree for the selected supergroup using pplacer. In addition, CSI-SSU conducts

bacterial contamination screening using BUSCO searches. The CSI-SSU tool generates a structured

output directory containing intermediate and final files, including a summary folder with

comprehensive reports and visualizations. All retrieved SSU sequences are provided for downstream

analyses, along with taxonomic assignments and placement confidence values based on the

likelihood weight ratio (LWR) from pplacer. The tool also produces an annotated phylogenetic tree in

which placed sequences are highlighted using a color scheme indicating both taxonomic assignment

and chimeric status. Each sequence is annotated across multiple taxonomic levels (supergroup,

order, family, and genus), together with corresponding LWR values, facilitating interpretation and

manual inspection. Although no universal LWR threshold is defined, simulation studies have shown

that higher LWR values are associated with lower placement error. In our empirical analysis of the

P10K dataset using CSI-SSU (see results below), most placements fell within LWR values of 0.9–1,

corresponding to high-confidence assignments. Lower LWR values were often associated with

shorter sequences (e.g., ~500 bp) or chimeric sequences, whereas higher-confidence placements

typically involved longer sequences (~1,000 bp) positioned within well-supported clades (e.g.,

bootstrap support >60) and exhibiting consistently high LWR values across taxonomic levels. It is

important to note that the resolution of the reference trees is constrained by the current SSU sampling

and by the phylogenetic signal of this single marker. Therefore, interpretation of placement results

must take into account the taxonomic representation of the target group (or the group to which the

SSU sequence is assigned) within the reference dataset. In addition, assembly complexity,

contamination, uneven taxonomic sampling, and intrinsic limitations of SSU-based phylogenetic

resolution may vary across lineages and influence placement precision. Accordingly, careful visual

inspection of the resulting phylogenetic trees and summary outputs is strongly recommended and is

explicitly supported by the tool's automated organization of results, designed to facilitate this process.

Finally, all identified SSU sequences are compiled into a summary FASTA file
(parsed_sequences.fasta), which can be used for downstream analyses, including phylogenetic
reconstruction.

In addition to the full workflow, CSI-SSU provides retrieval and placement modes. The retrieval
mode performs SSU sequence identification and extraction, along with bacterial BUSCO searches
and chimeric sequence detection, but does not include phylogenetic placement. This allows users to
recover SSU sequences from assemblies for downstream analyses without placing them. The
placement mode performs phylogenetic placement of user-provided SSU sequences. In this mode,
users can supply SSU sequences in FASTA format from any source, including primer-based 18S
amplification of target organisms or metabarcoding projects. Additionally, after running the full
workflow, retrieved SSU sequences identified as contaminants can be reanalyzed using the CSI-SSU
placement mode with a reference tree corresponding to the identified contaminant supergroup,
enabling more precise taxonomic placement and confirmation.

Benchmarking across eukaryotic supergroups, using 2,960 genomic-level assemblies from P10K
(see results below), demonstrated that the tool performs efficiently, with generally low runtime, CPU
usage, and memory consumption (**Fig. S2**). Median runtimes ranged from 3.9 to 9.2 minutes, median
CPU times from 2.9 to 7.4 minutes, and median memory usage from 332 to 459 MB. While most
analyses completed quickly, a subset of datasets required substantially more resources, with
runtimes reaching up to 154 minutes and memory usage up to 19.9 GB (**Fig. S2**). These higher
resource demands were primarily associated with large genome assemblies, with BUSCO accounting
for a significant portion of the computational cost due to its mapping and gene annotation steps. All
CSI-SSU analyses were conducted on the Quanah partition of the Texas Tech University High
Performance Computing Center (HPCC) RedRaider cluster, using a single thread on one node. The
Quanah partition is built on Intel Xeon E5-2695 v4 processors and runs Rocky Linux 9.2/8.10, with 36
cores and 192 GB of memory per node. To test the CSI-SSU tool and demonstrate its compatibility
with both Linux and macOS, we also performed analyses on a separate workstation: a Mac mini with
an Apple M4 Pro chip, 14 cores (10 performance, 4 efficiency), and 48 GB of RAM.

Contamination Screening of the P10K Database Using CSI-SSU

CSI-SSU retrieved and phylogenetically placed 13,513 SSU sequences from 2,960 genome-

level assemblies representing six eukaryotic supergroups obtained from the P10K database (**Figure**

1A–B). In most assemblies, CSI-SSU recovered multiple SSU sequences and placed with high

confidence (pplacer likelihood weight ratio ≥ 0.9) within the target supergroup or non-target

supergroups (**Figure 1C**). Across supergroups, Amoebozoa exhibited the highest levels of

contamination, whereas Obazoa showed the lowest. This pattern is consistent with differences in data

origin: Amoebozoa are largely represented by non-model organisms, many of which are derived from

environmental single-cell isolations, while Obazoa include several well-established model systems

with genome data typically generated from cultured material and listed as Reference level on NCBI

(**Supplementary Information - Table S1**). Based on VSEARCH searches, CSI-SSU identified

numerous chimeric SSU sequences across genomic assemblies from all surveyed supergroups

(**Figure 1D**). Many contaminant SSU sequences were assigned to a wide range of supergroups

(**Figure 1E–I**). Among these, TSAR was the most frequent source of contamination, followed by

Obazoa, Archaeplastida, and Amoebozoa. One Archaeplastida assembly contained an SSU

sequence assigned to Eukaryota_X and placed within Ancyromonas, a group classified as

Eukaryota_X in the PR2 database. A small subset of SSU sequences identified and retrieved was

placed with a low confidence (likelihood weight ratio as low as 0.2) and often forming long branches

(**Supplemental material – Data S1**). These patterns were associated with short or chimeric

sequences, which negatively influences the placement accuracy. Bacterial BUSCO scores further

indicate that, although many assemblies contain fewer than 30 of the 116 searched bacterial markers

(bacteria_odb12), some assemblies show scores exceeding 60 bacterial BUSCO genes, with a few

approaching near-complete bacterial gene representation (**Figure 1J**). This pattern is particularly

evident in datasets from Amoebozoa, Archaeplastida, Haptista, and TSAR, suggesting substantial

bacterial contamination despite the bacterial decontamination effort implemented by P10K project. It

is important to note that the presence of bacterial BUSCO genes does not necessarily indicate

bacterial contamination, as the possibility of genuine horizontal gene transfer (HGT) events cannot be

excluded. Therefore, bacterial BUSCO scores should be considered a proxy and interpreted

accordingly. Overall, the results derived from CSI-SSU highlight both the scale and heterogeneity of

contamination across protist genomic assemblies available in P10K, as well as the effectiveness of

CSI-SSU as a large-scale screening tool. CSI-SSU provides a robust and efficient first-pass

framework for SSU recovery and contamination detection. A detailed summary of CSI-SSU results is

presented on **Supplementary Information - Table S1** for each of the 2,960.

Phylogeny-based taxonomic identification of P10K amoebozoan data

Amoebozoan SSU sequences were successfully retrieved from 151 of the 201 genomic datasets

available for Amoebozoa in the P10K database (**Supplementary Information - Table S1**). Given the

established use of the Cytochrome c oxidase subunit I (COI) marker in Arcellinida phylogenetics, one

of the main amoebozoan groups represented in P10K database, we also specifically targeted this

marker for arcellinid taxa. Arcellinid COI sequences were successfully recovered from 40 of the 59

genomic data available for Arcellinida (**Supplementary Information - Table S1**). Phylogenetic

analyses based on SSU and COI enabled precise taxonomic identification, corroborating CSI-SSU

results and supporting most of the original taxonomic assignments in the P10K database

(**Supplementary Information - Table S1; Figs. 2 and 3**). They also allowed more accurate

classification for 43 taxa, including refined genus- and family-level assignments for taxa previously

classified only at higher taxonomic levels. A more precise taxonomic assignment was achieved using

more complete SSU and COI datasets curated for Amoebozoa, incorporating not only PR2

sequences but also SSU and COI sequences previously retrieved from transcriptomic data published

in the literature but not available on NCBI (**Figs. 2 and 3; Supplementary Information - Table S1**).

As species-level identification typically requires extensive morphological and morphometric data in

addition to molecular evidence, we adopted a conservative approach and assigned identifications at

the genus level based on our phylogenetic results. However, it is worth noting that several genomic

data from the P10K database originate from well-established cultures of widely used and shared

strains, some originally available in the NCBI database, for which detailed morphological data are

available in the literature, allowing confident species-level identifications (**Figs. 2 and 3**;

Supplementary Information - Table S1).

Notably, seven original taxonomic assignments from the P10K database were not corroborated

by our phylogenetic reconstructions (**Figures 2 and 3**; **Supplementary Information - Table S1**). This

was most apparent within Arcellinida, where misidentifications primarily involved closely related or

morphologically similar genera and families. Although Arcellinida are well known for their test (shell),

which provides informative taxonomic characters, convergent evolution has led to similar shell

morphologies across distantly related lineages. For example, species of *Difflugia* (Difflugiidae),

Hyalosphenia (Hyalospheniidae), and *Netzelia* (Netzeliidae) often possess rounded, ovoid, or

elongated shells, which can lead to misidentification if other shell features (e.g., aperture shape,

composition) or cellular characteristics are not considered. Consistent with this, our analyses revealed

that several taxa initially assigned to Difflugiidae (infraorder Longithecina) belong to Hyalospheniidae

(Hyalospheniformes) or Netzeliidae (Sphaerothecina) (**Supplementary Information - Table S1**).

These infraorders are distantly related, with their last common ancestor estimated to have lived over

500 million years ago [15, 25], making such misassignments evolutionarily significant. By comparison,

CSI-SSU correctly placed arcellinid SSU sequences within Arcellinida with high confidence (LWR ≥

0.9, most LWR = 1), consistent with results from SSU and COI phylogenetic reconstruction

(**Supplementary Information - Table S1; DataS1**). Outside Arcellinida, only one notable case of

misidentification was observed: *Pessonella* sp. PRA-29, a culture originally submitted to the American

Type Culture Collection (ATCC) under the genus *Pessonella*, was later described as a new genus

and species, *Armaparvus languidus*, representing the correct taxonomic identification for this

organism [26].

Contamination screening of the P10K amoebozoan data

Non-amoebozoan SSU sequences were retrieved from 58 of the 201 genomic datasets available

for Amoebozoa in the P10K database (**Supplementary Information - Table S1**). The phylogenetic

analysis of these non-amoebozoan SSU and SSU sequences from the PR^2 database reveals a

widespread contamination of the P10K amoebozoan dataset by a taxonomically diverse set of

eukaryotic lineages, mirroring the ecological complexity of the environments these protists inhabit

(**Figure 3 and Supplementary Information - Table S1**). Among the contaminant groups were fungi

and metazoans, including sequences from arthropods and nematodes, likely introduced via soil

particles, organic debris, or sample handling procedures (**Figure 4**). SSU sequences affiliated with

ciliates were particularly abundant, with representatives spanning multiple clades such as

Vorticellidae, *Coleps*, and *Pseudomicrothorax* (**Figure 4**). This reflects the high diversity and

abundance of ciliates across environments, leading to their consistent identification as contaminants.

Additional contaminants included lineages within the Stramenopiles such as *Paraphysomonas* and
Poterioochromonas, as well as centrohelid heliozoans (Haptista) and cercozoans (Rhizaria) (**Figure**
4). Several photosynthetic eukaryotes represented another major group of contaminants, including
land plants, especially angiosperms of the Fabaceae family, likely introduced through pollen or plant
debris, as well as green algae from the Chlorophyceae (e.g., *Chlamydomonas*, *Hyalomonas*) and
Zygnematophyceae, the closest relatives of land plants (**Figure 4**). It is worth noting that symbiotic
associations may also contribute to the high levels of apparent contamination observed in
Amoebozoa and other eukaryotic groups [27]. For example, several taxa of testate amoebae are
known to be consistently associated with green algae such as Chlorella (Chlorophyta) [27]. The
identification of some contaminant sequences may therefore reflect biological associations rather than
external contamination. By comparison, CSI-SSU assignments were consistent with the phylogenetic
reconstruction results (**Supplementary Information - Table S1**). Notably, for the phylogenetic
reconstruction approach we implemented, only the top five BLAST hits were retained. In contrast,
CSI-SSU retrieves all sequences identified by BLAST hits, resulting in a more comprehensive
identification and classification of contaminant SSU sequences. This is particularly important for
assemblies contaminated by multiple non-target supergroups or containing multiple SSU copies per
group (**Figure 1; Supplementary Information - Table S1; Data S1**).

Quality assessment of the P10K data

Collectively, CSI-SSU and the comparative reconstruction approach findings corroborates that
contamination in eukaryotic assemblies is both frequent and taxonomically widespread,
encompassing multiple branches of the eukaryotic tree. Based on our contamination screening results
and other parameter, we can evaluated the current state of microbial eukaryote genomic data in the
P10K database. Samples can be considered relatively higher quality if they contain the SSU marker
for the target organisms, show no evidence of contamination based on SSU phylogenetic
reconstruction, and have a eukaryotic BUSCO completeness score of at least 50% (**Supplementary**
Information - Table S1). The eukaryotic BUSCO scores, already included in the P10K metadata,
were used here to assess completeness. Such samples are more likely to yield meaningful results in
downstream analyses, including phylogenomics and comparative genomics (**Supplementary**

Information - Table S1). It is important to note that SSU-based contamination screening has

limitations and cannot detect all sources of potential contamination. Specifically, while the presence of

a contaminant SSU informs the chimeric characteristics of a given assembly, the absence of SSU

sequences does not necessarily indicate that a sample is free of contamination, as other contaminant

sequences may still be present. While we are considering a relatively low BUSCO threshold (≥50%)

to include a broader range of genomic data, more stringent completeness cutoffs and additional

quality metrics, such as genome contiguity, N50, and assembly size, may be required depending on

the intended analyses. Similarly, P10K samples lacking key markers, flagged as contaminated by

SSU phylogenetic inference, or with BUSCO scores below 50% represent lower-quality or incomplete

genomic data (**Supplementary Information - Table S1**). Several assemblies require

decontamination or additional sequencing prior to downstream analyses (**Supplementary**

Information - Table S1). This assessment also highlights that several major lineages remain

unsampled (or poorly sampled) at the genomic level and provides a guide for targeted sequencing

efforts aimed at expanding taxonomic coverage in the P10K database.

Contamination in protist genomic data, is expected and widespread due to their ecological

context: these organisms inhabit complex microbial communities, engage in predation, and often form

close associations with other microbes, making isolation of pure genomic material challenging [13, 28,

29]. Contaminant sequences can compromise downstream analyses by inflating genome complexity,

distorting gene family evolution, misleading functional annotation, and generating artifactual

phylogenetic relationships [4, 30, 31]. While cultivation and single-cell isolation techniques, combined

with careful sterile handling, filtration, and repeated transfers, can reduce contamination, complete

elimination is rarely possible, particularly for environmentally derived specimens. Therefore,

comprehensive contamination screening, such as single-marker phylogenetic analysis (e.g., SSU)

coupled with genome-level examination, is essential for accurate data curation and taxonomic

validation. Additionally, the P10K database currently lacks detailed metadata and specimen

documentation, including voucher images and information on sequencing platforms, which limits

reproducibility, taxonomic verification, and interpretation of genomic data. Incorporating such

metadata, alongside rigorous contamination screening, would enhance the reliability, usability, and

long-term scientific value of microbial eukaryotic genomic resources.

Conclusions

This study introduces CSI-SSU, a scalable and reproducible tool for phylogenetically informed

contamination screening and taxonomic validation of genomic datasets, and demonstrates its

performance across the microbial eukaryotic assemblies from the P10K database. By integrating

SSU-based sequence retrieval, chimeric detection, and phylogenetic placement within a unified

workflow, CSI-SSU enables rapid identification of target lineages while simultaneously detecting non-

target eukaryotic and bacterial contamination. Applied at scale, CSI-SSU tool revealed widespread

and taxonomically diverse contamination, reflecting the ecological complexity of protist-associated

environments and the inherent difficulty of generating contamination-free genomic data. Importantly,

our results highlight that single-marker phylogenetic screening, especially using SSU given its

traditional marker and existence of curated databases, sampling in diverse, provides a robust and

efficient strategy for both taxonomic validation and exploratory contamination detection. CSI-SSU

offers a critical entry point for large-scale data curation, allowing researchers to rapidly assess

dataset quality and prioritize assemblies for downstream analyses. By improving the accuracy and

interpretability of genomic resources, CSI-SSU contributes directly to enabling reliable evolutionary,

ecological, and functional genomic studies. More broadly, it establishes a generalizable framework for

curating the rapidly expanding volume of microbial eukaryotic genomic data. As initiatives like P10K

continue to grow, integrating CSI-SSU tool into standard pipelines, alongside improved metadata,

contamination control strategies, and reproducible workflows, will be essential to maximize the

scientific value of these resources and to support robust inferences about the evolution and

diversification of eukaryotic life.

Availability of data and materials

The CSI-SSU tool is hosted on GitHub (https://github.com/AlexTiceLab/CSI-SSU/). All compiled and

curated SSU and COI datasets associated with this manuscript are publicly available on FigShare at

https://doi.org/10.6084/m9.figshare.29814947.

Competing interests

The authors declare that they have no competing interests

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Authors' contributions

Alfredo L. Porfirio-Sousa: Conceptualization, Data curation, Visualization, Formal analysis,

Investigation, Writing – original draft, Writing – review & editing. **Robert E. Jones**: Investigation, Data

curation, Writing – review & editing. **Matthew W. Brown:** Investigation, Writing – review & editing.

Daniel Lahr: Investigation, Writing – review & editing. **Alexander K. Tice:** Conceptualization, Data

curation, Visualization, Formal analysis, Investigation, Writing – original draft, Writing – review &

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Declaration of generative AI in scientific writing

523 During the preparation of this work the authors used ChatGPT GPT-4o to improve the readability

524 and language of the first draft of the manuscript. After using this tool, the authors reviewed and edited

525 the content as needed and take full responsibility for the content of the published article.

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Figures

Figure 1. Summary of contamination screening of the P10K database using CSI-SSU tool. A.
Bar chart showing the total number of assemblies per supergroup in the P10K database screened
with CSI-SSU. Values are displayed on a $\log_{10}$ scale for improved visualization. **B.** Bar chart showing
the total number of SSU sequences identified, retrieved, and placed by CSI-SSU tool per supergroup.
C. Stacked bar chart showing the percentage of assemblies per supergroup by SSU supergroup
assignment category: no SSU identified; SSU sequences assigned only to the target supergroup;
SSU sequences assigned to the target and one or multiple non-target supergroups; and SSU
sequences assigned only to non-target supergroups (single or multiple). The bar chart shows the total
number of assemblies containing within-supergroup contamination. Colors correspond to assignment
categories shown in the legend. **D.** Stacked bar chart showing the percentage of assemblies per
supergroup with SSU sequences classified as chimeric or non-chimeric. **E.** Bar chart showing the
number of Amoebozoa assemblies with SSU sequences assigned to contaminant supergroups. **F.**
Bar chart showing the number of Archaeplastida assemblies with SSU sequences assigned to
contaminant supergroups. **G.** Bar chart showing the number of Excavata assemblies with SSU
sequences assigned to contaminant supergroups. **H.** Bar chart showing the number of Haptista
assemblies with SSU sequences assigned to contaminant supergroups. **I.** Bar chart showing the
number of TSAR assemblies with SSU sequences assigned to contaminant supergroups. **J.** Boxplot
showing the distribution of bacterial markers identified in assemblies from each supergroup. The
number of complete BUSCOs represents the count of complete bacterial BUSCO genes detected out
of the 116 markers in the bacteria_odb12 database. All displayed results represent assignments with
an LWR of at least 0.9 and SSU sequences retrieved and placed with a minimum length of 500 bp.

Figure 2. The phylogenetic placement of Amoebozoa genomic data from the P10K focused on the major groups Evosea and Discosea. Maximum-likelihood phylogenetic trees constructed from the Small Subunit ribosomal RNA (SSU) inferred from a subset of the curated dataset presented in Figure S2. **A.** Focuses on the major Amoebozoa group Evosea. Phylogenetic reconstruction was conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model (TIM2+F+R4). **B.** Focuses on the major Amoebozoa group Discosea. Phylogenetic reconstruction was conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model (GTR+F+G4). Node support was assessed using both ultrafast bootstrap (UFBoot) and the Shimodaira–Hasegawa approximate likelihood ratio test (SH-aLRT). Support values are reported as SH-aLRT / UFBoot, with values ≥80/95 considered indicative of strong support [32, 33]. For clarity, high-support values are omitted in this figure, as well as support values for nodes above the nodes indicated by the arrows, which are represented mostly by flat branches. The complete tree, including all support values, is shown in Figures S4 and S5. Stars indicate genomic data reassigned to a different taxonomic identity than reported in the P10K database. Filled circles mark indicate cases with refined taxonomic resolution. Filled squares denote more conservative identifications (e.g., genus or family level) rather than the more specific genus or species-level identification originally provided in the P10K database. 'C' indicates putative chimeric genomes containing sequences from multiple eukaryotes and 'AC' denotes intra-Amoebozoa contamination.

Figure 3. The phylogenetic placement of Amoebozoa genomic data from the P10K focused on the major group Tubulinea. A. The maximum-likelihood phylogenetic tree constructed from the Small Subunit ribosomal RNA (SSU) inferred from a subset of the curated dataset presented in Figure S2, focuses on the major Amoebozoa group Tubulinea. Phylogenetic reconstruction was conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model (TIM3e+G4). The lengths of branches depicted as dashed lines and marked with an asterisk have been reduced by 50% for presentation purposes. **B.** The maximum-likelihood phylogenetic tree of cytochrome c oxidase subunit I (COI) inferred from a curated dataset generated in the present study, focusing on Arcellinida order (Tubulinea:Amoebozoa) comprising COI sequences retrieved from genomes and transcriptomes available in the P10K database, along with reference sequences made available by previous. Phylogenetic reconstruction was conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model (GTR+F+I+G4). Node support was assessed using both ultrafast bootstrap (UFBoot) and the Shimodaira–Hasegawa approximate likelihood ratio test (SH-aLRT). Support values are reported as SH-aLRT / UFBoot, with values ≥80/95 considered indicative of strong support [32, 33]. For clarity, high-support values are omitted in this figure, as well as support values for nodes above the one indicated by the arrow, which are represented mostly by flat branches. The complete tree, including all support values, is shown in Figures S6 and S7. Stars indicate genomic data reassigned to a different taxonomic identity than reported in the P10K database. Filled circles mark indicate cases with refined taxonomic resolution. Filled squares denote more conservative identifications (e.g., genus or family level) rather than the more specific genus or species-level identification originally provided in the P10K database. 'C' indicates putative chimeric genomes containing sequences from multiple eukaryotes and 'AC' denotes intra-Amoebozoa contamination.

Figure 4. Contamination-screening phylogenetic tree. The maximum-likelihood phylogenetic tree constructed from the Small Subunit ribosomal RNA (SSU), inferred from a curated dataset generated in the present study, comprises SSU sequences of putative non-Amoebozoan contaminants retrieved from genomes and transcriptomes available in the P10K database, along with reference sequences from the PR^2 database (indicated by PR^2 IDs), identified through BLAST+ similarity searches using the putative contaminant SSU sequences as queries. Phylogenetic reconstruction was conducted using IQ-TREE v2.3.6, with ModelFinder identifying the best-fit substitution model (TN+F+I+G4). Node support was assessed using both ultrafast bootstrap (UFBoot) and the Shimodaira–Hasegawa approximate likelihood ratio test (SH-aLRT). Support values are reported as SH-aLRT / UFBoot, with values ≥80/95 considered indicative of strong support [32, 33]. For clarity, high-support values are omitted in this figure, as well as support values for nodes above the one indicated by the arrow, which are represented mostly by flat branches. The complete tree, including all support values, is shown in Figure S8. The taxonomic identification of Amoebozoa taxa corresponding to the displayed P10K ID codes is shown in parentheses.

A

P10K-NCBI-001094 *Dictyostelium discoideum AX4*
Dictyostelium discoideum JF930944.5 PR2
P10K-NCBI-001194 *Dictyostelium citrinum OH494*
Dictyostelium citrinum JF930954.5 PR2
P10K-NCBI-002012 *Dictyostelium rosarium M45*
Dictyostelium rosarium AM168065.1 PR2
P10K-NCBI-002013 *Dictyostelium brefeldianum TNS-C-115*
P10K-NCBI-002062 *Dictyostelium capitatum 91HO50*
Dictyostelium brefeldianum AM168030.1 PR2
Dictyostelium capitatum AM168032.1 PR2
P10K-NCBI-001731 *Dictyostelium purpureum P1B1*
Dictyostelium purpureum LC159265.1 PR2
P10K-NCBI-002063 *Dictyostelium gargantuum MYA-3825*
Dictyostelium gargantuum GQ496161.1 PR2
P10K-NCBI-001186 *Polysphondylium violaceum QSvi11*
Polysphondylium fuscans JX173877.1 PR2
Polysphondylium violaceum OP310985.1 PR2
P10K-NCBI-002156 *Coremiostelium polycephalum*
Coremiostelium polycephalum HQ141490.1 PR2
P10K-NCBI-001740 *Speleostelium caveatum*
Speleostelium caveatum AM168077.1 PR2
P10K-NCBI-002155 *Synstelium polycarpum* ⓒ
Synstelium polycarpum AM168057.1 PR2

73.7/96
41.1/51

Eumycetozoa

P10K-NCBI-001347 *Acytostelium subglobosum LB1*
Acytostelium amazonicum HQ141510.1 PR2
Acytostelium magnisorum HQ141513.1 PR2
P10K-NCBI-001739 *Heterostelium multicystogenum*
Heterostelium multicystogenum HQ141506.1 PR2
P10K-NCBI-002154 *Acytostelium* sp.
Acytostelium digitatum AM168114.1 PR2
P10K-NCBI-002153 *Rostrostelium ellipticum*
Rostrostelium ellipticum HE614595.1 PR2
P10K-NCBI-002157 *Cavenderia deminutiva*
Cavenderia deminutiva AM168092.1 PR2

95.9/62
26.3/89
80.5/90

Variosea

P10K-NCBI-001563 *Protostelium mycophagum ATCC PRA-154*
Protostelium mycophagum JF766448.1 PR2
P10K-NCBI-001665 *Protostelium aurantium var. fungivorum Jena*
Protostelium aurantium FJ766464.1

84.9/71

Archamoebae

P10K-NCBI-001343 *Mastigamoeba balamuthi ATCC 30984*
Mastigamoeba balamuthi L23799.1
P10K-NCBI-002230 *Mastigamoeba balamuthi*
Mastigamoeba punctachora JX157653.1 PR2

84.1/85
58.1/77

P10K-NCBI-001655 *Entamoeba moshkovskii Laredo*
P10K-NCBI-001152 *Entamoeba dispar SAW760*
Entamoeba dispar OP524421.1 PR2
P10K-NCBI-001215 *Entamoeba invadens IP1*
P10K-NCBI-001485 *Entamoeba histolytica HM1:IMSS*
Entamoeba histolytica AB426549.1
Entamoeba nuttalli MH626560.1 PR2
P10K-NCBI-001150 *Entamoeba histolytica* HM-1:IMSS ⓒ
Entamoeba histolytica AAFB02000685.1 PR2

91.2/93

P10K-NCBI-002111 *Pelomyxa schiedti SKADARSKE*
Pelomyxa belevskii JX157663.1 PR2
Pelomyxa palustris JX157666.1 PR2

Variosea

P10K-NCBI-002442 *Filamoeba nolandi*
Filamoeba nolandi AY714368.1 PR2
Filamoeba sinensis AY714369.1 PR2
Flamella aegyptia EU186021.1 PR2

51.1/87

Leptomyxa reticulata AF293898.1
Leptomyxa variabilis KT945250.1 PR2

0.2

Legend box:

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

B

Acanthamoeba astronyxis HQ007041.1 PR2
P10K-NCBI-001353 *Acanthamoeba astronyxis*
P10K-NCBI-001361 *Acanthamoeba divionensis*
P10K-NCBI-001356 *Acanthamoeba comandoni Ph30/40*
Acanthamoeba royreba AF239299.1 PR2
P10K-NCBI-001356 *Acanthamoeba healyi*
P10K-NCBI-001355 *Acanthamoeba mauritaniensis*
P10K-NCBI-001360 *Acanthamoeba rhysodes*
Acanthamoeba lenticulata U94734.1 PR2
P10K-NCBI-001579 *Acanthamoeba lenticulata 26330*
P10K-NCBI-001588 *Acanthamoeba lenticulata PT-14*
P10K-NCBI-001359 *Acanthamoeba royreba*
Acanthamoeba castellanii U07400.1 PR2
P10K-NCBI-002238 *Acanthamoeba castellanii Namur*
P10K-NCBI-001448 *Acanthamoeba polyphaga Linc Ap-1*
P10K-NCBI-001357 *Acanthamoeba palestinensis*
Acanthamoeba castellanii KX431477.1 PR2
P10K-NCBI-001363 *Acanthamoeba quina*
Acanthamoeba triangularis AF316647.1 PR2
P10K-NCBI-001358 *Acanthamoeba polyphaga*
P10K-NCBI-001547 *Acanthamoeba pearcei*
P10K-NCBI-002235 *Acanthamoeba triangularis SH 621*
Acanthamoeba castellanii AEYA01002259.6 PR2
P10K-NCBI-001546 *Acanthamoeba castellanii Neff*
P10K-NCBI-001362 *Acanthamoeba hatchetti*

Centramoebia

P10K-NCBI-001384 *Balamuthia mandrillaris CDC-V039*
Balamuthia mandrillaris LFUD11000968.1 PR2

Cunea russae MN317565.1 PR2
P10K-NCBI-002446 = *Cunea* sp.
Cunea profundata KP862837.1 PR2
Cunea profundata KP862834.1
Cunea thuwala KP862839.1 PR2
P10K-NCBI-002454 *Stygamoeba* sp. ■
Stygamoeba regulata JF694349.1 PR2

P10K-MW-000663 *Mayorella* sp. ●
P10K-MW-000317 *Mayorella* sp. ⓒ
P10K-MW-000278 *Mayorella* sp. ●
P10K-MW-000286 *Mayorella* sp. ●
P10K-MW-000309 *Mayorella* sp. ●
P10K-MW-000380 *Mayorella* sp. ●
P10K-MW-000323 *Mayorella* sp. ●
P10K-MW-000972 *Mayorella* sp. ●
P10K-MW-000332 *Mayorella* sp. ●
P10K-MW-000300 *Mayorella* sp. ●
P10K-MW-000638 *Mayorella* sp. ●
P10K-MW-000271 *Mayorella* sp. ● ⓒ
P10K-MW-000272 *Mayorella* sp. ● ⓒ
P10K-MW-000523 *Mayorella* sp. ●
P10K-MW-000612 *Mayorella* sp. ●
Mayorella gemmifera EU719190.1 PR2
P10K-MW-000266 *Mayorella* sp. ●
P10K-MW-001016 *Mayorella* sp. ●

Flabellinia

P10K-MW-000383 *Korotnevella* sp. ●
P10K-MW-000803 *Korotnevella* sp. ●
P10K-MW-000823 *Korotnevella* sp. ●
P10K-MW-000824 *Korotnevella* sp. ●
Korotnevella heterocantha EU273190.1 PR2
P10K-MW-000826 *Korotnevella* sp. ●
Korotnevella limbata KU550317.1 PR2
P10K-MW-000308 *Korotnevella* sp. ●
P10K-MW-000293 *Korotnevella* sp. ●
P10K-MW-000607 *Korotnevella* sp. ●
P10K-MW-000620 *Korotnevella* sp. ●
P10K-MW-000399 *Korotnevella* sp. ●
P10K-MW-000455 (Nebela sp.2 P10K-) ● = Arcellinida:Tubulinea) AC
Korotnevella jeppesenii KP719186.1 PR2
Korotnevella venosa KU681502.1 PR2
P10K-MW-000279 *Korotnevella* sp. ●

P10K-NCBI-001583 *Paramoeba pemaquidensis CCAP 1560/4*
Paramoeba longipoda MF197368.1 PR2
P10K-NCBI-002253 *Paramoeba* sp. ■
Paramoeba atlantica JN202436.1 PR2
P10K-NCBI-002347 *Paramoeba* sp.

P10K-NCBI-002497 *Vannella* sp.
Vannella persistens GQ265371.1 PR2
Vannella simplex KP719199.1 PR2
P10K-MW-000326 *Vannella* sp. ● ⓒ
Vannella danica GQ265389.1 PR2
P10K-NCBI-000013 *Vannella* sp. ⓒ
Vannella miroides AY183888.1 PR2
P10K-NCBI-002355 *Vannella* sp.
Vannella robusta OM508936.1 PR2
P10K-NCBI-002357 *Vexillifera* sp.
P10K-NCBI-002360 *Vexillifera* sp.
Vexillifera abyssalis MH349019.1 PR2
Vexillifera tasmaniana HQ687483.1 PR2

P10K-MW-000282 *Netzeliidae* sp. AC

Centramoebia

Cochliopodium plurinucleolum KJ569732.1 PR2
Cochliopodium minus KJ569702.1 PR2
Squamoeba japonica JN638032.1 PR2

P10K-NCBI-002449 *Armparvus languidus*
Armparvus languidus EU273458.1

0.3