



Long-read metabarcoding of the eukaryotic rDNA operon to phylogenetically and taxonomically resolve environmental diversity

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Abstract

High-throughput DNA metabarcoding of amplicon sizes below 500 bp has revolutionized the analysis of environmental microbial diversity. However, these short regions contain limited phylogenetic signal, which makes it impractical to use environmental DNA in full phylogenetic inferences. This lesser phylogenetic resolution of short amplicons may be overcome by new long-read sequencing technologies. To test this idea, we amplified soil DNA and used PacBio Circular Consensus Sequencing (CCS) to obtain an ~4500-bp region spanning most of the eukaryotic small subunit (18S) and large subunit (28S) ribosomal DNA genes. We first treated the CCS reads with a novel curation workflow, generating 650 high-quality operational taxonomic units (OTUs) containing the physically linked 18S and 28S regions. To assign taxonomy to these OTUs, we developed a phylogeny-aware approach based on the 18S region that showed greater accuracy and sensitivity than similarity-based methods. The taxonomically annotated OTUs were then combined with available 18S and 28S reference sequences to infer a well-resolved phylogeny spanning all major groups of eukaryotes, allowing us to accurately derive the evolutionary origin of environmental diversity. A total of 1,019 sequences were included, of which a majority (58%) corresponded to the new long environmental OTUs. The long reads also allowed us to directly investigate the relationships among environmental sequences themselves, which represents a key advantage over the placement of short reads on a reference phylogeny. Together, our results show that long amplicons can be treated in a full phylogenetic framework to provide greater taxonomic resolution and a robust evolutionary perspective to environmental DNA.

KEYWORDS

metabarcoding, PacBio, phylogeny, protists, rDNA operon, taxonomy

1 | INTRODUCTION

Sequencing of environmental DNA (eDNA), here encompassing DNA contained in cells as well as cell-free DNA, is a popular approach to study the diversity and ecology of microbial eukaryotes, including small animals, fungi and protists. eDNA has catalysed the discoveries of novel lineages at all taxonomic ranks from abundant to rare taxa, and revealed that most, if not all, known groups of microbes are genetically much more diverse than anticipated (Heger et al., 2018; Massana et al., 2015; Pawlowski et al., 2012; de Vargas et al., 2015). For protists, recent global molecular surveys revealed that they can account for up to 80% of the total diversity of eukaryotes in the environment (Logares et al., 2014; Massana et al., 2015; Pawlowski et al., 2012; de Vargas et al., 2015). Initially, these molecular environmental studies relied on cloning the small subunit (SSU) ribosomal RNA gene (18S rDNA) followed by Sanger sequencing, thereby generating reads of sufficient length to enable reasonably accurate phylogenetic interpretation of the results (Amaral Zettler et al., 2002; Bass & Cavalier-Smith, 2004; Dawson & Pace, 2002; Diez et al., 2001; Edgcomb, Kysela, Teske, de Vera Gomez, & Sogin, 2002; Lopez-Garcia, Philippe, Gail, & Moreira, 2003; López-García, Rodríguez-Valera, Pedrós-Alió, & Moreira, 2001; Massana, Balagué, Guillou, & Pedrós-Alió, 2004; Massana, Castresana, et al., 2004; Moon-Van Der Staay, De Wachter, & Vaulot, 2001; Stoeck & Epstein, 2003; Stoeck, Taylor, & Epstein, 2003). Today, however, the overwhelming majority of eDNA data correspond to much shorter reads produced by Illumina, which routinely generates several million reads (Bates et al., 2013; Geisen, 2016; de Vargas et al., 2015). This enables sequencing a large fraction of the species present in an environment, even including extremely rare organisms (Logares et al., 2014; de Vargas et al., 2015). The drawback of this method is that only genetic regions limited to a few hundred nucleotides (typically < 500) can be sequenced at a time, for example the hypervariable V4 or V9 regions of the 18S rDNA or the internal transcribed spacer (ITS) (Mahé et al., 2015; Pawlowski et al., 2012; Stoeck et al., 2010).

Short amplicons contain relatively low phylogenetic signal (Dunthorn et al., 2014), which complicates taxonomic identification especially when environmental reads are only distantly related to reference sequences. To address the issue of low phylogenetic signal in high-throughput data, a range of tools have been developed to provide reasonable taxonomic identification of environmental OTUs (operational taxonomic units). Given the mass number of reads available, the most straightforward approach is to use pairwise sequence similarity searches against reference databases (e.g., as done in Mahé et al., 2017; de Vargas et al., 2015). Although fast, this approach is highly sensitive to the taxon sampling and annotation accuracy of the reference database. If a taxonomic group is absent or sequences are misannotated in the reference database, the corresponding queries will be only approximately annotated, remain unidentified, or worse, wrongly identified (Berger, Krompass, & Stamatakis, 2011). Recognizing the limitations of similarity-based

methods, new tools have been developed that place short sequences into a phylogenetic context. The Evolutionary Placement Algorithm (EPA; implemented in RAxML, or more recently in EPA-NG) (Barbera et al., 2019; Berger et al., 2011) or pplacer (Matsen, Kodner, & Armbrust, 2010) are two such tools. They are becoming popular methods that use a reference tree of carefully selected (often long) sequences to successively score the optimal insertion position of every query sequence or OTU. These methods perform well, and have contributed to the discovery of novel eukaryotic lineages from environments where poor references exist (Bass et al., 2018; Mahé et al., 2017). However, the phylogenetic placement of short reads still requires the independent construction of a reference data set, which by definition does not include the short reads themselves. Thus, methods such as EPA rely on the availability of reference sequences generally produced by the less efficient and more expensive Sanger sequencing, or on genome or transcriptome sequencing projects. Furthermore, references are often based on cell cultures, which are available only for a small fraction of the diversity.

To better exploit the phylogenetic signal of the rDNA operon in environmental metabarcoding studies, newer long-read sequencing technologies such as the Pacific Biosciences platform (PacBio) hold great promise. PacBio has lower throughput and higher error rates than Illumina but can produce reads that are over 20 kb long at a fraction of the cost of Sanger sequencing. In the last few years, PacBio sequencing has started to be applied to metabarcoding studies, primarily on prokaryotic 16S rDNA (Mosher et al., 2014; Schloss, Junior, Koumpouras, Westcott, & Highlander, 2016; Wagner et al., 2016) and most recently on larger amplicons also including the 23S rDNA (Martijn et al., 2019). For eukaryotes, the 18S rDNA was nearly fully sequenced for targeted microbial groups (Orr et al., 2018), whilst longer regions also spanning the ITS and the large subunit (LSU) 28S gene were used to analyse fungal diversity (Heeger et al., 2018; Tedersoo & Anslan, 2019; Tedersoo, Tooming-Klunderud, & Anslan, 2018). These studies showed that in spite of the high error rates of PacBio, when applying a corrective process based on multiple sequence passes (Circular Consensus Sequences [CCS]) together with rigorous quality filtering, long-amplicon sequencing is emerging as a robust approach for studying environmental diversity.

Here, we used soil eDNA samples to generate broad eukaryote amplicons of about 4,500 bp spanning the 18S rDNA, ITS1, 5.8S, ITS2 and the 28S rDNA regions. We used PacBio-CCS to sequence these long amplicons and applied several filtering steps to retain only high-quality sequences. We then followed a full phylogenetic workflow to accurately annotate long sequences with taxonomy even in the absence of close references. These annotated sequences were combined with available references to infer a well-resolved global eukaryotic phylogeny from a concatenated 18S–28S alignment. Together, this study represents an important step forward to use the full power of phylogenetics to derive the accurate evolutionary origins of known and novel lineages present in the environment, as well as expanding rDNA sequence databases for metabarcoding of eukaryotes.

2 | MATERIALS AND METHODS

All new scripts listed below are available on Github (<https://github.com/Pbdas/long-reads>).

2.1 | Environmental samples and DNA extraction

We used three environmental soil samples for this study: (a) soil from Tibet, China, collected in summer 2011 from alpine meadows and coniferous forests; (b) rape seed rhizosphere samples from Newbald, Nuneaton, York and Morden in the UK, collected in March 2015; and (c) pooled set-aside agricultural soils from Wellesbourne, UK, collected in September 2010 as described by Gosling, van der Gast, and Bending (2017). Rhizosphere samples were collected as follows: loosely adhering soil was removed from the roots leaving no more than 2 mm of rhizosphere soil. Roots were washed sequentially in 4 × 25 ml sterile distilled water to release the rhizosphere soil which was then centrifuged and the excess water drained to leave a pellet of rhizosphere soil. All soil samples were extracted using a PowerSoil DNA Isolation Kit (MoBio Laboratories) following the manufacturer's instructions with the following modifications: (a) rhizosphere soil samples were homogenized in the TissueLyser II (Qiagen) at 20 Hz for 2 × 10 min with a 180° rotation of the plates in-between; and (b) set-aside agricultural soils were processed using a Precellys 24 homogenizer (Bertin Technologies) for the initial mechanical lysis step.

2.2 | PCR and PacBio sequencing

We used two sets of eukaryotic universal primers to amplify a region covering the 18S, ITS1, 5.8S, ITS2 and 28S (Table 1). One 18S internal forward primer, 3NDf (which anneals to the conserved region adjoining the 5' end of the V4 region, *Escherichia coli* position 505), was used in conjunction with two 28S internal reverse primers, 21R (*E. coli* position 1926) and 22R (*E. coli* position 1952), to amplify a region of ~4,500 bp. The forward and reverse primers are described in Cavalier-Smith, Lewis, Chao, Oates, & Bass (2009) and Schwelm, Berney, Dixelius, Bass, and Neuhauser (2016), respectively, and were chosen to maximize the eukaryotic diversity obtained. Taxon coverage of the primers was checked in silico using SILVA TESTPROBE 3.0 (Quast et al., 2013): primers 3NDf, 21R and 22R matched against 91.5%, 88.1% and 87.2% of all eukaryotic sequences in SILVA release 132, respectively. For each sample, two PCRs were carried out (one for each combination of forward and reverse primers), and the PCR products were subsequently pooled.

PCRs were carried out using the Takara PrimeSTAR GXL high-fidelity DNA polymerase, selected for its capacity to amplify long fragments, in 25-μl reactions with 10–20 ng of template DNA. The following cycling conditions were used: denaturation at 98°C for 10 s, primer annealing at 60°C for 15 s and extension at 68°C for 90 s. A final extension time of 60 s was used after 30 cycles. This protocol corresponds to the rapid PCR protocol of Takara GXL

where extension time was shortened by adding twice as much polymerase. PCR products were purified by polyethylene glycol and ethanol precipitation and were pooled and concentrated using Amicon 0.5-ml 50K columns (Merck). Amplicon sizes were checked using TapeStation (Agilent Technologies) before SMRTbell library preparations. Three SMRT cells (one per soil sample) on the PacBio Sequel instrument with v2 chemistry were used for sequencing. Additionally, one RSII SMRT cell was used to sequence a constructed sample with known diversity (see 2.3 below). Sequencing and library preparation were carried out at Uppsala Genome Center, Science for Life Laboratory.

2.3 | Sequencing of a known community

To validate our curation pipeline and to assess error rates, we constructed a small community of three fungal samples: two unidentified isolates of Agaricomycetes species (BOR77 and BOR79) as well as the species *Phaeosphaeria luctuosa*. We amplified the 18S gene using two sets of primers (Table 1): AU2 and AU4 for BOR77 and BOR79 (Vandenkoornhuyse, Baldauf, Leyval, Straczek, & Young, 2002), and 3NDf and 1510R (Amaral-Zettler, McCliment, Ducklow, & Huse, 2009) for *Phaeosphaeria*. All PCRs were conducted in 20-μl final volumes with 1 μl of template DNA and a final concentration of 0.5 μM of each primer, 0.4 mM dNTPs, 2.5 mM of MgCl₂, 0.2 mg bovine serum albumin (BSA), 1× Promega Green Buffer and 0.5 U of Promega GoTaq. Amplicons were sequenced with Sanger sequencing to obtain reference sequences against which the PacBio sequences could be compared. To assess error rates, curated PacBio sequences were searched against the 18S reference sequences with VSEARCH version 2.3.4 (Rognes, Flouri, Nichols, Quince, & Mahé, 2016) using the --usearch_global option with the following settings: --id 0.9 --strand both --maxaccepts 0 --top_hits_only --fulldp --userfields query + target+id + alnlen+mism + gaps. The error rate was calculated as (mismatches + indels)/length of alignment.

2.4 | Sequence curation and clustering pipeline

To address PacBio's high error rate, we used a stringent sequence curation pipeline (Figure 1a; Figure S1a). CCS were generated from raw reads by SMRT LINK version 4.0.0.190159 using a minimum number of two passes and minimum predicted accuracy of 0.99 with all other settings set to default. The latter was shown in Schloss et al. (2016) to be the most important factor in decreasing error rate. At this stage, we pooled sequences from the three samples, resulting in one fastq file. A fasta file was generated using the fastq.info command (pacbio = T) in MOTHUR version 1.39.5 (Schloss et al., 2009). Sequences at this step of the pipeline still include nonspecific PCR amplicons, PCR artefacts such as chimeras and some sequencing errors such as long homopolymer runs. These were filtered out using the trim.seqs command in MOTHUR using the following settings: minlength = 2,500, maxlength = 6,000 (to discard nonspecific and

Primer	Oligonucleotide sequence (5'–3')	Reference
3NDf	GGCAAGTCTGGTGCCAG	(Cavalier-Smith et al., 2009)
21R	GACGAGGCATTTGGCTACCTT	Schwelm et al. (2016)
22R	CCATTCATGCRCGTCACWART	Schwelm et al., (2016)
AU2	TTTCGATGGTAGGATAGDGG	Vandenkoornhuysen et al. (2002)
AU4	RTCTACTAAGCCATTC	Vandenkoornhuysen et al. (2002)
1510R	GGTTACCTTGTACGACTT	Amaral-Zettler et al. (2009)
TAReuk454FWD1 ^a	CCAGCASCYGGGTAATTCC	Stoeck et al. (2010)
TAReukREV3 ^a	ACTTTCGTTCTTGATYRA	Stoeck et al. (2010)

^aThese primers were only used in silico in this study.

TABLE 1 List of primers used in this study

incomplete PCR amplicons), maxhomop = 6 (to stringently discard sequences with a homopolymer run of more six nucleotides), and qwindowsize = 50 and qwindowaverage = 30 (to trim the few sequences with a stretch of low-quality sequence). The remaining nonspecific PCR amplicons were filtered out by using BARRNAP version 0.7 (--reject 0.4 --kingdom euk) (<https://github.com/tseemann/barrnap>), which predicts the presence and location of 18S and 28S genes in the sequences. Reads with unexpected structure (more than one 18S, 28S or 5.8S) or incomplete/nonspecific reads (missing 18S and/or 28S) were discarded. An in-house perl script was used to identify sequences represented by reverse strand (using the BARRNAP output) and subsequently to reverse complement them so that all sequences are in the same direction.

The sequences were then denoised by preclustering as described by Martijn et al. (2019) to curate the remaining sequencing errors that are randomly distributed. Briefly, sequences were clustered at 99% similarity using VSEARCH version 2.3.4 (Rognes et al., 2016) (--cluster_fast --id 0.99). For each resulting precluster with three or more reads, we aligned the reads with MAFFT version 7.271 (--auto) (Katoh & Standley, 2013) and generated a majority-rule consensus sequence using the consensus.seqs (cutoff = 51) option in MOTHR. Gaps were removed to yield final consensus sequences.

The denoised sequences as well as sequences from preclusters of size 1 and 2 were subjected to de novo chimera detection using UCHIME (Edgar, Haas, Clemente, Quince, & Knight, 2011) (as implemented in MOTHR) (chunks = 40, abskew = 1; abundance of the denoised sequences was taken as the number of sequences in their respective preclusters). Our PCR primers amplified a few archaea ribosomal genes, and these were filtered out by removing sequences with BLAST hits (Altschul, Gish, Miller, Myers, & Lipman, 1990) to prokaryotic sequences in the SILVA SSU Ref NR 99 database version 132 (Quast et al., 2013). Finally, we used in-house perl scripts to extract the 18S and 28S sequences from the cleaned reads, and aligned them with MAFFT-AUTO version 7.271 (Katoh & Standley, 2013). Poorly aligned sequences were removed after manual inspection.

We used the canonical 97% similarity threshold for 18S to cluster sequences into OTUs using an average-linkage hierarchical clustering method. This was done by first generating a distance matrix using the dist.seqs (cutoff = 0.2) command in MOTHR and then clustering sequences using the cluster command. We used the get.oturep command (label = 0.03, method = distance) in MOTHR to obtain as the representative sequence of each OTU the sequence with the smallest distance to all other sequences in the cluster, and extracted the same sequences from the 28S sequence set as OTU representatives. From the total set of 1,154 OTUs, we discarded all singletons to be conservative, and obtained a final set of 650 OTUs (hereafter referred to as queries).

2.5 | Taxonomic annotation

Several data sets were constructed for phylogeny-aware taxonomic annotation and accuracy assessment. These are summarized in Table 2 and described below.

2.5.1 | Phylogeny-aware annotation

The 18S gene alone was used for taxonomic annotation because the reference database for this gene is much more comprehensive than its 28S counterpart. The basis for this pipeline is an 18S rDNA tree constructed with both labelled references and the (yet unlabelled) queries. Known reference sequences (RS) were obtained from SILVA SSU Ref NR 99 release 132 (Quast et al., 2013). The RS set comprised two subsets: (a) 504 RS representative of global eukaryotic diversity—these were derived from the 512 taxa data set used by Mahé et al. (2017); and (b) two to five nearest neighbours of each query in the SILVA database. To obtain these, each query sequence was aligned (mafft --auto) with the top 50 BLAST hits against high-quality (pintail > 0) eukaryotic SILVA SSU sequences, and pairwise

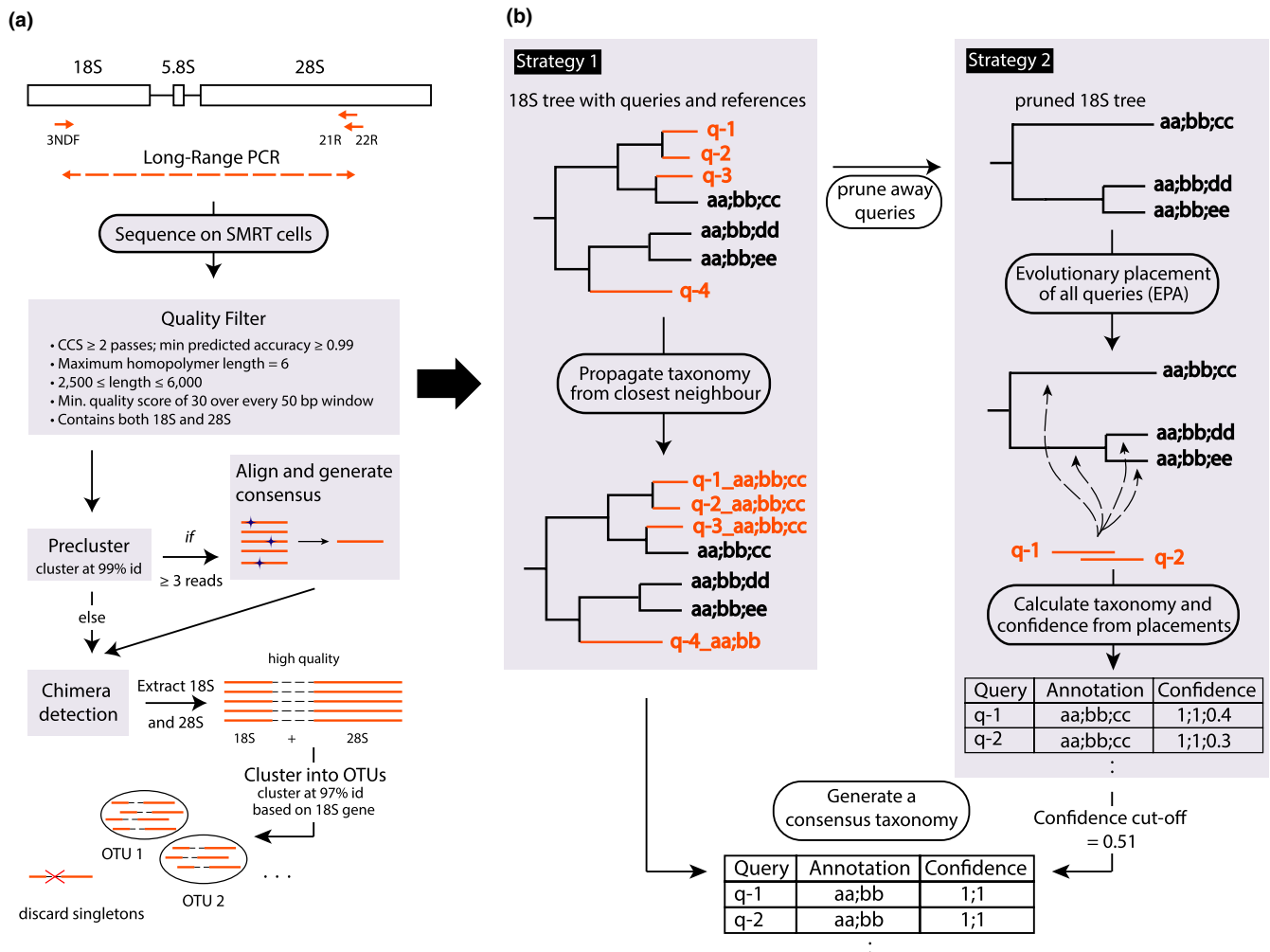


FIGURE 1 Workflow for generating long rDNA amplicons, quality filtering and taxonomic annotation. (a) General eukaryotic primers (3NDF, 21R and 22R) were used to amplify an ~4500-bp fragment of the rDNA operon (including 18S, ITS1, 5.8S, ITS2 and 28S) from environmental soil samples. The amplicons were sequenced on SMRT cells and subject to quality filtering before being preclustered at 99% similarity. Preclusters with at least three reads were denoised further by aligning them and generating majority-rule consensus sequences. These were combined with the remaining reads and subject to de novo chimera detection. The 18S and 28S gene regions extracted from these regions in the final step were physically linked as they originated from the same amplicon. (b) Overview of the taxonomic annotation pipeline. Reference sequences are shown in black with taxonomic ranks separated by ";" (e.g., aa;bb;cc denote three taxonomic rank labels: aa, bb and cc). Queries are shown in orange. See text for details on the pipeline [Colour figure can be viewed at wileyonlinelibrary.com]

maximum-likelihood (ML) distances were computed in RAxML (option -f x) (Stamatakis, 2014) under the GTR + GAMMA model of substitution (Yang, 1994). RS with the lowest pairwise ML distances with the query were selected as the nearest neighbours, resulting in 1,157 RS after removing duplicates. Combining the two subsets resulted in a total of 1,661 RS, which covered all major eukaryotic groups and, when available, included sequences closely related to queries. The final data set thus comprised the 650 queries plus the 1,661 RS (2,311 sequences in total; Table 2). These 2,311 sequences were aligned with MAFFT (--retree 2 --maxiterate 1,000) and trimmed with TRIMAL (-gt 0.3 -st 0.001), resulting in a multiple sequence alignment (referred to as MSA) with 1,589 alignment sites. The best unconstrained ML tree was selected from 20 tree searches run using RAxML-NG (version 0.6.0) (Kozlov, Darriba, Flouri, Morel, & Stamatakis, 2018). We assumed that the SILVA taxonomy is correct

and consistent with the exception of a few cases—preliminary tree searches detected several potentially mislabelled RS, which were re-labelled after careful inspection.

Based on this tree, a consensus taxonomy was derived using a combination of two strategies (Figure 1b). Strategy 1: use a custom program written with the Genesis library (Czech, Barbera, & Stamatakis, 2019) to propagate the taxonomy of the closest related reference to each query. Specifically, the program propagates the taxonomic annotation up the tree (where one exists), solving conflicts at inner nodes by taking the intersection of the taxonomic annotation (i.e., lowest common ancestor). Once complete, it propagates that information down to the nonlabelled taxa (queries). Strategy 2: queries were first removed from the tree before being placed back one at a time using EPA-NG (version 0.2.1-beta; Barbera et al., 2019). The location and likelihood weights of

TABLE 2 Summary of data sets used in this study

	Eukaryote 18S	Eukaryote 18S + 28S	Apicomplexa
Usage	Taxonomic annotation of environmental queries	Phylogenetic reconstruction of environmental samples	Assessing the effect of query length on inferring phylogeny of environmental diversity
Alignment size (no. of taxa: no. of positions in alignment)	(i) 2,311:1,588 ^a (ii) 2,311:RS = 1588, QS = V4 region ^b	1,019:4,304	(i) 107:1,509 (18S) (ii) 107:3,250 (18S + 28S) (iii) 107:RS = 1588, QS = V4 region (18S-V4)

Abbreviations: QS, query sequences; RS, reference sequences.

^aThis data set is referred to as "MSA" in the main text.

^bThis data set is referred to as "MSA-V4" in the main text.

the placements are then used to compute the taxonomic assignment and the confidence associated with each taxonomic rank as in SATIVA (Kozlov, Zhang, Yilmaz, Glöckner, & Stamatakis, 2016). This last step is implemented in the gappa tool "assign" (Czech et al., 2019) (<https://github.com/lczech/gappa>). Finally, the consensus taxonomy for all queries was produced by a perl script that calculates the intersection of taxonomic paths from strategies 1 and 2 (when the SATIVA-derived confidence score for a rank is 0.51 or above). Taxonomic annotations assigned to each query were propagated to their 28S counterparts as they are physically linked on the same molecule.

2.5.2 | Comparison with short reads

We evaluated the effect of query sequence length by running the taxonomic annotation pipeline with short Illumina reads that were generated in silico. We focused on the V4 region (~500 bp) of the SSU gene, which is commonly used in barcoding studies, for example in Mahé et al. (2017). This data set (MSA-V4) was derived from the original MSA by using the V4 flanking primers (Table 1), TAREuk454FWD1 and TAREukREV3 (Stoeck et al., 2010), to trim only the query sequences (median length ~340 nucleotides), leaving the rest of the MSA untouched (Table 2). After running the taxonomic annotation pipeline, we performed the following analyses:

1. The accuracy of placement is crucial for correct taxonomic annotation and we compared that for the long and short queries using two metrics: (a) LWR (likelihood weight ratio) of the most probable placement for each query—this is computed as the ratio of the likelihood of the tree with the query at branch x to the sum over the likelihoods of all other possible placements (Matsen et al., 2010); and (b) EDPL (expected distance between placement locations) shows how far the placements are spread across the tree. It is computed as the sum of the distances between placements along the branches of the tree, weighted by their probability (LWR) (Matsen et al., 2010).
2. We conducted pairwise comparisons of the taxonomic assignments and the confidence for taxonomic ranks given to each query based on MSA and MSA-V4.

2.5.3 | Comparison with sequence similarity-based methods of taxonomic assignment

To assess how our method compares with similarity-based methods, we initially constructed a reference database consisting of high-quality (pintail > 0) eukaryotic sequences in SILVA SSU Ref NR99 release 132 and the 504 RS derived from Mahé et al. (2017). RS were trimmed with the forward primer 3NDf and both queries and RS were trimmed with the reverse primer 1510R to ensure that they spanned the same region. Queries were searched against this reference set using the global pairwise alignment strategy (--usearch_global option) in VSEARCH version 2.3.4 (Rognes et al., 2016) with the following settings: --notrunclabels --userfields query + id1 + target --maxaccepts 0 --maxrejects 32 --top_hits_only --output_no_hits --id 0.5 --iddef 1. Sequences were taxonomically assigned based on the top hit, and in case of multiple top hits, the common ancestor of the hits was computed. For each query, the percentage similarity to the closest reference sequence was recorded and the taxonomic classification to deep-branching lineages was compared to that of the phylogeny-based method.

2.6 | Phylogenetic analyses

Information on the different alignments used for phylogenetic reconstruction can be found in Table 2.

2.6.1 | 18S + 28S global phylogeny

To phylogenetically resolve the biodiversity in our soil samples, we constructed a phylogeny using a concatenated 18S + 28S data set. For each of the two genes, queries were aligned with their respective reference sequences using MAFFT version 7.271 (--retree 2 --maxiterate 1,000). Alignments were filtered with TRIMAL (-gt 0.3 -st 0.001) and a perl script (https://github.com/iirisarri/phylogm/blob/master/concat_fasta.pl) was used to concatenate the SSU and LSU alignments. The phylogeny was inferred with RAXML version 8.2.10 as offered on the Cipres web server (Miller, Pfeiffer, & Schwartz, 2010), with 20 tree searches under the GTR + GAMMA model of

substitution and 300 nonparametric bootstrap replicates. The construction of the reference data set is described below.

Reference sequences were included only when we could easily verify that the 18S and 28S genes originated from the same species or organism. These reference sequences were derived from several public databases, as follows: (a) searched NCBI nt using the following search filters: ((ribosomal RNA) AND 4,000:9000[Sequence Length]) AND Eukaryota[Organism]; (b) blasted whole queries (18S, ITS, 28S) against nt and retained sequences with a minimum HSP of 2,500 bp and 80% similarity; (c) obtained all 18S and 28S sequences from SILVA release 132 possessing the same accession number in the SSU Ref NR 99 and LSU Ref databases; (d) included the 108-taxon data set used in an article studying the eukaryote tree with 18S + 28S genes (Moreira et al., 2007); and (e) used BARRNAP to search all "protist" genomes available in Ensembl Release 92 (Zerbino et al., 2018). This resulted in 3,479 taxa after removing duplicates, from which we manually selected sequences, in a best effort, to assemble the most representative data set possible. Initial tree-building attempts placed certain cercozoan and apicomplexan lineages aberrantly among Excavata and Amoebozoa due to long branch attraction. To mitigate this effect, we sorted taxa by branch length (as in Heiss et al., 2018) and removed the longest 118 (10.4%) branches from subsequent analyses. The final data set contained 589 queries and 430 reference sequences (1,019 in total) with 4,304 alignment sites (Table 2).

2.6.2 | Apicomplexa phylogenies

To investigate the effect of query length on resolving environmental diversity in more detail, we constructed additional phylogenies of the fast-evolving group Apicomplexa (Table 2). Reference sequences were obtained by downloading 40 GenBank 18S accessions of which 28 had 28S sequences also available. We constructed concatenated and full-length 18S genes trees by aligning the references and queries separately using MAFFT version 7.271 (--linsi) and trimming alignments with TRIMAL (-gt 0.3 -st 0.001). Trees were inferred with RAXML, using the substitution model GTR + GAMMA from 20 searches and 100 bootstrap runs. Finally, we constructed an 18S tree from reference sequences alone on which queries shortened to the V4 region (trimmed with universal eukaryotic primers TAReuk454FWD1 and TAReukREV3; Stoeck et al., 2010), were placed with EPA-NG version 0.3.5.

3 | RESULTS

3.1 | Sequence curation

A total of 113,362 long rDNA CCS, all containing two or more passes, were generated with PacBio Sequel. These CCS reads were filtered by a series of stringent quality controls including the removal of nonspecific amplicons and prokaryotic sequences, as well as chimera detection (Figure 1a; Figure S1a). At the end of the curation

pipeline, the amplicons had on average 9.95 CCS passes ($SD = 2.9$) (Figure S1b). The mean error rate was estimated to be 0.17% based on comparisons between CCS reads curated by our pipeline and known Sanger sequences of the same species of fungi (see Materials and Methods). OTUs were generated using a 97% similarity threshold based on the 18S region only, leading to 650 high-quality clusters after removing singletons. These OTUs ranged in length from 2,501 to 5,956 bp (Figure S1c). Most OTUs contained fewer than 10 reads, but some were much larger and probably represented the most abundant organisms in the samples; the largest OTU (6,416 sequences) corresponded to *Brassica napus*, the main crop species cultivated in one of the samples, while the second largest OTU (1,322 sequences) belonged to the gregarines (Apicomplexa), a group of parasites of various invertebrates that has been shown to be particularly abundant in some soil environments (Mahé et al., 2017).

3.2 | Phylogeny-aware taxonomic annotation

In order to annotate the environmental queries with taxonomy, we developed a phylogeny-aware approach that takes advantage of the increased sequence length (Figure 1b). We used only the 18S part of the queries because the taxon sampling of 18S reference sequences is considerably denser than that of 28S sequences. The taxonomic assignments were based on an 18S tree (Figure S2) inferred from the 650 queries together with 1,661 full-length references from the SILVA SSU database (i.e., a total of 2,311 taxa; Table 2). A consensus taxonomy was then derived from two strategies (Figure 1b; Materials and Methods).

Using this approach, we could confidently assign a majority of queries (627/650, or 96.5%) to deep-branching eukaryotic lineages (Figure S3), including queries with similarity to references below 80%. The remaining 23 queries that were not assigned to any of the recognized major lineages were all highly divergent; of these, 18 could be classified with confidence only to higher-rank assemblages that roughly correspond to the so-called supergroups—the most inclusive established groups of eukaryotes (Burki, Roger, Brown, & Simpson, 2019). For the remaining five queries, two were ambiguous even at the level of supergroups, thus possibly representing novel deeply branching lineages and/or sparsely sampled taxonomic groups in the reference database, whilst three proved to be chimeras that had escaped automated filtering. Interestingly, our method also performed well for low-rank taxa, as 226 queries could be reliably annotated down to the genus and species levels (Table S1).

We further investigated the performance of our method by comparing it to a commonly used similarity-based taxonomic annotation tool (VSEARCH, Rognes et al., 2016), which revealed several discrepancies. As expected, the most divergent sequences (i.e., <80% similar to known references) showed the highest level of conflicts in taxonomic assignment (Figure 2); 43.7% of these divergent sequences (21/48 queries) were assigned to different deep-branching eukaryotic lineages, and even to different supergroups in four cases. These conflicting assignments became less pronounced for more

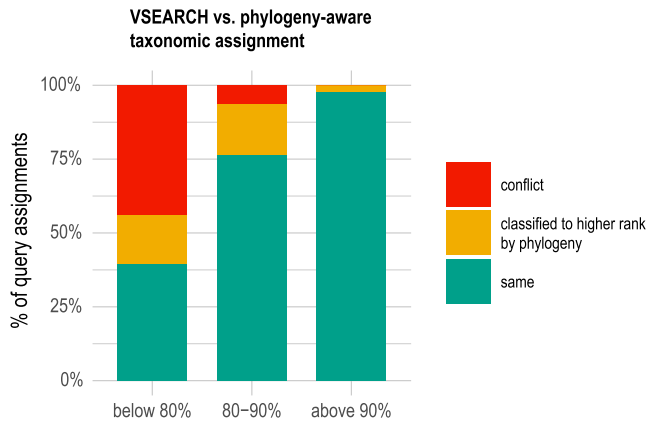


FIGURE 2 Broad comparison of taxonomic classifications by *VSEARCH* and our phylogeny-aware method. Environmental queries are split into three sets depending on the level of similarity to the closest reference in the SILVA database. Stacked bars show percentages of taxonomic assignments to the same deep-branching lineage (green), different lineages (red; deemed as conflicting) and to higher ranks in the same taxonomic path (yellow) [Colour figure can be viewed at wileyonlinelibrary.com]

similar sequences (i.e., between 80% and 90% similarity), where we observed only nine (1.6%) conflicts, while there was no conflict for queries >90% similar to a reference (Figure 2).

To explore the conflicts between the different approaches in more detail, we focused on the 10 most abundant lineages in our data (Figure 3a). For each lineage, taxonomic assignments by *VSEARCH* was used as a reference and compared to the assignment derived from our phylogeny-aware method, using both the full-length sequence and the V4 region. Several differences were observed: false negatives, that is sequences assigned to the lineage by phylogeny but not by similarity; false positives, that is sequences assigned to the lineage by similarity but not by phylogeny; and higher-rank assignments (but not conflicting), that is sequences assigned by phylogeny only to a more inclusive rank in the same taxonomic path. The amount and type of differences were to a large extent group-specific. Three groups contained no conflicting taxonomy (Ciliophora, Phytomyxea and Tubulinea), only a small number of higher-rank assignments by V4 phylogenetic assignment. All other groups, however, showed some levels of conflict. Apicomplexa and Zoopagomycota displayed the highest number of false negatives, with both phylogeny-based approaches classifying more queries to these groups than *VSEARCH* (42.5% and 100% more queries respectively, blue bars in Figure 3a). False positives were relatively more abundant in Colpodellida, where ~40% of queries assigned to this group by *VSEARCH* were assigned to a different group by one or the other phylogenetic method (pink bars in Figure 3a).

We found that a key difference between similarity and phylogeny is that the latter approach is much more flexible in the level of taxonomic resolution without requiring subjective decisions. For instance, ~23% of the queries with <90% similarity to known sequences were conservatively classified to higher taxonomic ranks by our approach compared to *VSEARCH* (Figure 2; Figure S4).

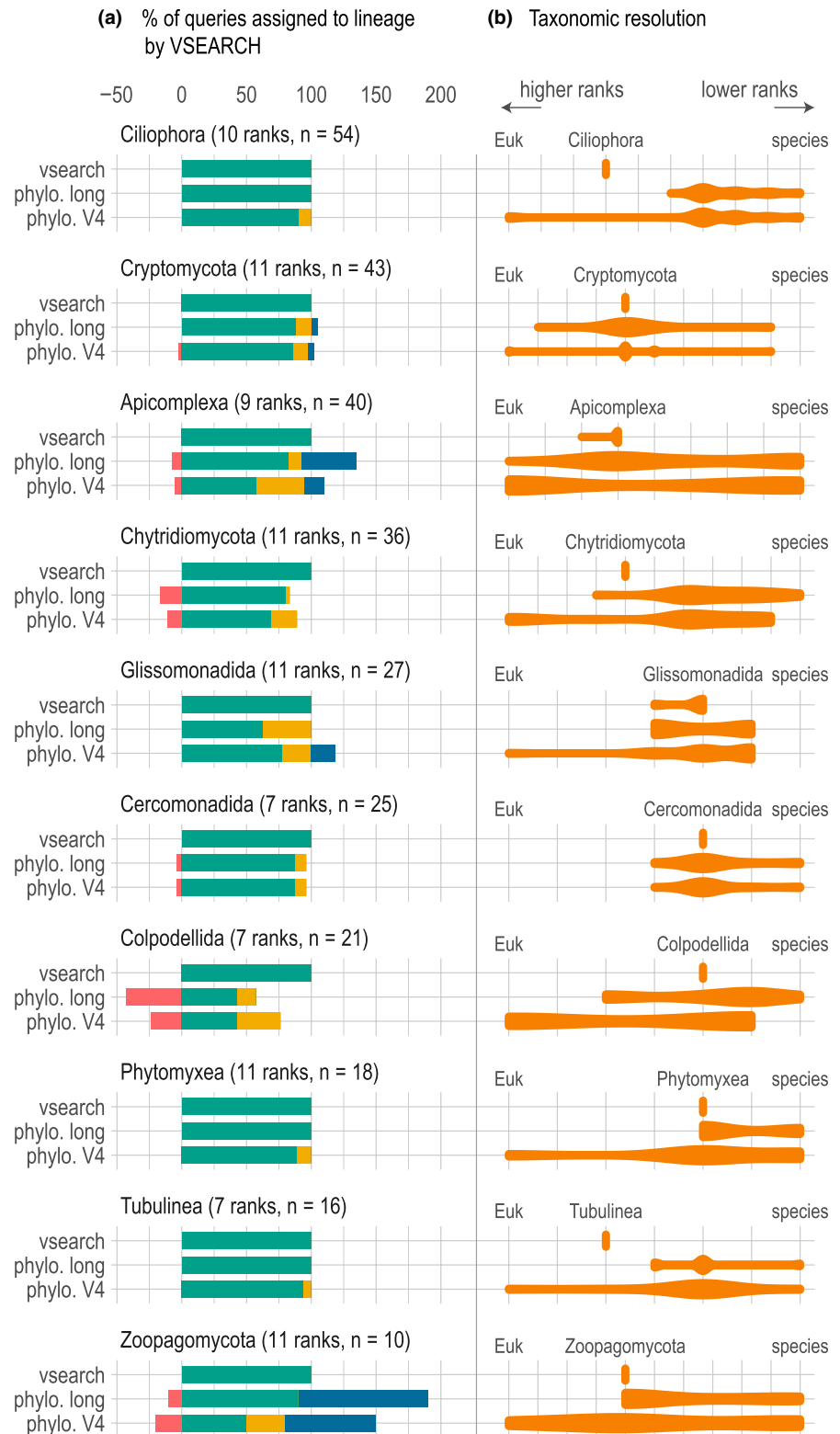
A comparison of the lowest rank assignments by all three methods illustrates this behaviour (Figure 3b). The similarity method always classified queries to the same predetermined rank, here corresponding to one of the 10 most abundant lineages in our data. In sharp contrast, both phylogenetic methods displayed a broader range of taxonomy, from higher to lower ranks (sometimes to the species level), depending on the confidence in the assignment. Interestingly, the added information from the longer sequences (long vs. V4) translated into increased taxonomic resolution, namely more assignments towards lower ranks (e.g., Phytomyxea in Figure 3b; Figure S5). Furthermore, in the absence of closely related references, our method can correctly propose no specific annotation. A good test case for this case was the recently suggested supergroup-level lineage Hemimastigophora (Lax et al., 2018). One query with 85% sequence similarity to Streptophyta in SILVA was labelled as an “unidentified eukaryote” by our method, whilst it was logically annotated as a land plant by *VSEARCH*. When using GenBank instead, this query was 98% similar to a newly added hemimastigote sequence, and thus not a land plant but indeed no specific grouping in the absence of that sequence.

3.3 | Combined 18S + 28S rDNA phylogeny of eDNA

The availability of long queries in principle allows us to better resolve the origin of environmental sequences due to increased phylogenetic signal. We assembled a concatenated 18S + 28S data set including the annotated queries and reference sequences mined from various public databases. The references were selected such that it could be verified that both the 18S and the 28S rDNA sequences originated from the same species (see Materials and Methods). We included representatives of all major eukaryotic lineages where possible. In addition, preliminary tree searches were used to identify long-branching taxa, which were removed in downstream analyses to reduce potential long branch attraction artefacts. This yielded a final data set of 1,019 taxa, of which a majority (589 taxa = 58%) represented new environmental queries. Importantly, because the 18S sequences are physically linked to their 28S counterpart on the CCS reads, the taxonomic annotation inferred with our phylogeny-aware method could be transferred to the combined 18S + 28S reads. This provided a diverse set of taxonomically annotated environmental queries in otherwise sparsely populated reference sequences (Figure 4).

Figure 4 shows an ML tree of the 1,019 taxa data set. The phylogenetic relationships were in general agreement with previous phylogenies based on the 18S and 28S (Moreira et al., 2007; Zhao et al., 2012), even recovering several well-established supergroups that were first proposed based on substantially larger concatenated protein data sets such as Sar (including the subclades Stramenopila, Alveolata and Rhizaria) or Opisthokonta (including the subclade Holomycota and Holozoa) (Baldauf, Roger, Wenk-Siefert, & Doolittle, 2000; Burki et al., 2007). Overall, more than half of the

FIGURE 3 Detailed comparison of taxonomic assignments for the 10 most abundant protist lineages in our data by VSEARCH (similarity-based approach) and the phylogeny-based approach for long (phylo. long) and truncated queries (phylo. V4). (a) Stacked bars indicate the proportion of queries assigned to each lineage by different methods. Green bars: same taxonomic assignment as VSEARCH; yellow bars: queries assigned to higher ranks in the same taxonomic path; blue bars: queries assigned to the lineage by phylogeny but not by similarity (i.e., false negatives); pink bars: sequences assigned to a different lineage by phylogeny (i.e., false positives). The number of taxonomic ranks in SILVA (from “Eukaryota” to species level) for each lineage is indicated in parentheses, followed by the number of queries (n) assigned to the lineage by VSEARCH. (b) Violin plots showing the taxonomic resolution of the assignments to each lineage. For the phylogeny-based approach these correspond to the green, yellow and blue bars [Colour figure can be viewed at wileyonlinelibrary.com]



newly sequenced diversity (345 queries; 53% of all queries) corresponded to microbial taxa other than fungi or animals (Figure 4). Members of Alveolata and Rhizaria accounted for nearly 70% of these protist queries—the most dominant lineages in decreasing number of queries were Ciliophora, Apicomplexa, Cercomonadida, Phytomyxea, Glissomonadida and Vampyrellida. The remaining

sequenced diversity was dominated by fungal lineages, accounting for 203 queries (31% of all queries) that equally represented dikarya (Ascomycota and Basidiomycota) and the so-called early-diverging fungi (EDF). Of these EDF, Cryptomycota and Chytridiomycota were particularly diverse. The remaining 16% of the queries corresponded to various animal lineages as well as land plants.

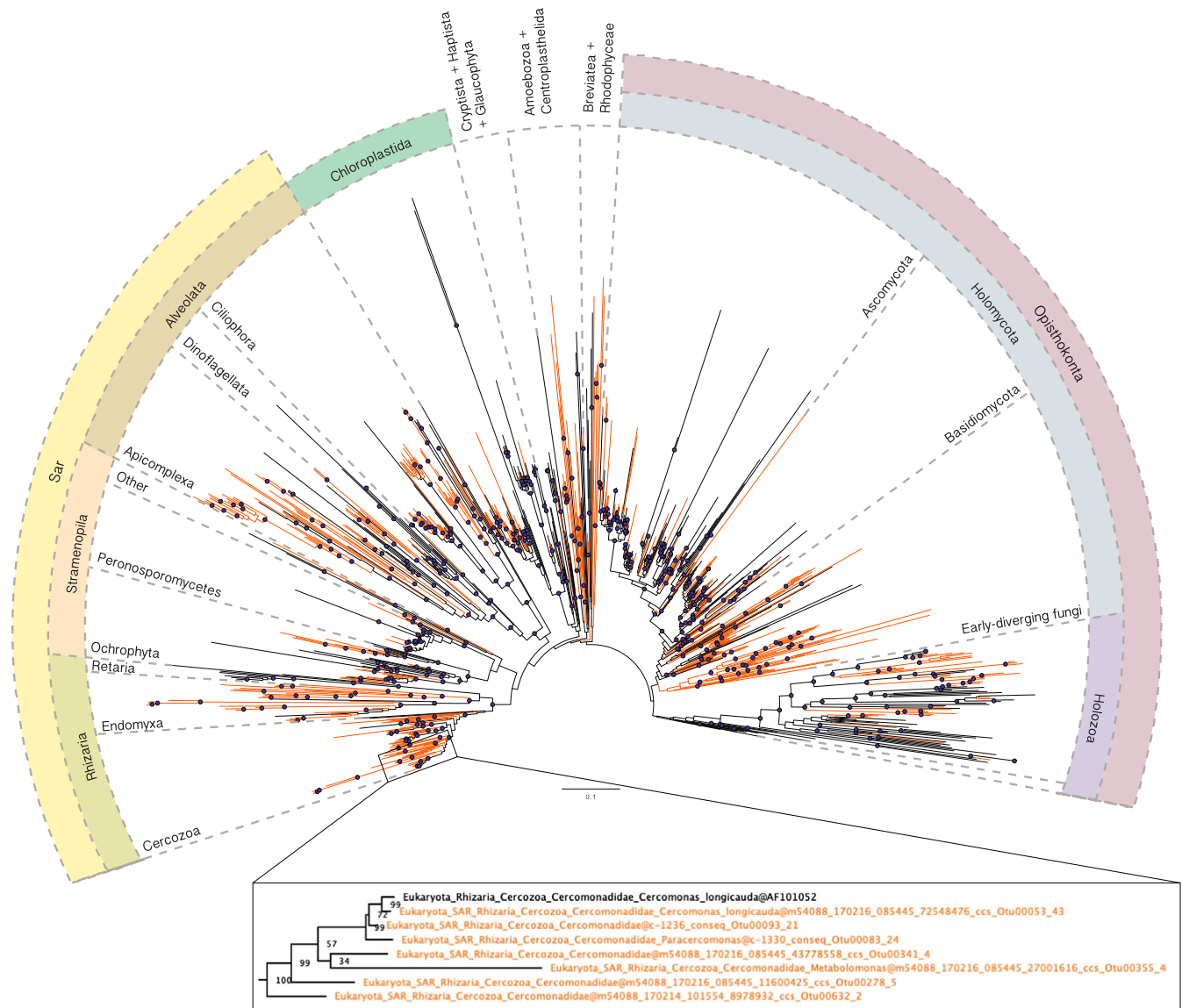


FIGURE 4 Maximum-likelihood (ML) tree showing the phylogenetic distribution of eukaryotic ribosomal diversity from soil samples. The tree corresponds to the best ML tree inferred from a concatenated alignment of 18S and 28S genes using the GTR + Gamma model. Bipartition support (BS) is derived from 300 bootstrap replicates. Purple circles represent BS $\geq 75\%$. The tree includes 589 queries from soil samples (coloured in orange) and 430 reference sequences (coloured in black) and all major eukaryotic supergroups (with the exception of Excavata which were eliminated during removal of long branches; see Materials and Methods for details). For details on taxon sampling, see Materials and Methods. The lower panel shows a subtree in Rhizaria. Zooming in illustrates how transferring the taxonomic annotation to the queries (coloured in orange) makes the phylogeny more informative, especially in clades (such as Rhizaria) where there are few references (labelled in black) with both the 18S and 28S available [Colour figure can be viewed at wileyonlinelibrary.com]

3.4 | Comparison to 18S-only and V4-based phylogenetic classifications of eDNA

The combined 18S + 28S tree described above (Figure 4) provides a new solution for obtaining a taxonomically annotated and well-resolved phylogenetic framework from high-throughput environmental sequencing. To assess the extent to which the added information of the 28S gene improved phylogenetic resolution, we first compared the combined tree to the 18S-only tree constructed for the taxonomic assignment. Interestingly, both trees were largely in agreement, suggesting that the ~1,000-bp fragment sequenced for

the 18S gene combined with the substantially denser reference sampling available for this gene provided sufficient phylogenetic signal to recover many groupings. However, the combined tree received generally higher bootstrap support values: 54.3% of the bipartitions (552/1,016) received $\geq 75\%$ bootstrap support in the combined tree compared to 43.1% bipartitions (994/2,308) in the 18S tree. The combined tree also supported (bootstrap $> 75\%$) more specific phylogenetic positions for a few queries that were taxonomically annotated only to high-rank taxa based on the 18S tree. For example, two queries labelled as Opisthokonta could be assigned more precisely to Aphelidea and as sister to nucleariids in the combined 18S + 28S

tree, respectively; and one deep branching eukaryote in the 18S tree in fact corresponded to a long branch within Ascomycota in the combined tree.

To investigate the benefits of long reads for phylogeny-based resolution of environmental diversity in more detail, we constructed three additional data sets with varying sequence lengths focusing on the Apicomplexa (Table 2). The sequence lengths corresponded to (a) the combined 18S + 28S alignment; (b) the full-length 18S-only alignment; and (c) an alignment of full-length 18S reference sequences but with query sequences shortened to the V4 region. Taxon sampling across the three data sets was identical to facilitate comparison, containing 67 queries and 40 reference sequences. Inspection of the combined and the 18S trees (Figures S6 and S7) revealed no major discrepancies and placed all 56 queries among gregarines. As with the full eukaryotic tree, bootstrap values were globally higher in the combined tree but many relationships remained unsupported. However, we observed several exceptions where the increased resolution of the combined tree allowed for better interpretation. Most importantly, the monophyly of Apicomplexa was statistically supported in the combined tree (83%) whereas it was unsupported in the 18S tree (23%). The same was observed for the monophyly of other established groupings, such as the haematozoa (79% vs. 55% in the combined and 18S trees, respectively) and haematozoa + coccidians (92% vs. 47% in the combined and 18S tree, respectively). Furthermore, the combined tree (Figure S6) resolved the eugregarine superfamily, Actinocephaloidea, into two separate clades with moderate to strong support (99% and 75%, respectively), while they did not form separate clades in the 18S tree as previously noted based on this gene only. Comparison with the phylogenetic placement of the V4 query sequences revealed that EPA-NG successfully placed all apicomplexan queries among gregarines with the exception of one query, which was placed close to *Plasmodium* instead. However, the reference-only 18S tree had a different topology than the full 18S tree, presumably because the short queries were not used for inferring the latter tree (Figures S7 and S8). Furthermore, a close inspection showed that three queries were probably misplaced by EPA-NG on the branch leading to the cephaloidophorids (parasites of marine invertebrates; Figure S8) when instead they formed a robust monophyletic clade putatively representing novel lineages on both the 18S and the combined trees (Figures S6 and S7).

4 | DISCUSSION

In this study, we broadly sequenced the near-complete eukaryotic rDNA operon from environmental soil samples, using PacBio sequencing. To our knowledge this is the first long amplicon environmental sequencing study that uses a full phylogenetic approach to assess the diversity of all eukaryotes. To reduce the inherently high error rate of PacBio, we combined the CCS approach with a series of stringent filtering steps and clustering. The final error rate of the CCS reads has a mean of 0.17%, which is comparable with the error rate of Illumina (0.21%; Schirmer, D'Amore, Ijaz, Hall, & Quince, 2016), or

in other PacBio-based studies (Schloss et al., 2016; Tedersoo et al., 2018; Wagner et al., 2016). Although the curating pipeline discarded the majority of CCS reads, many of which might still be of high quality, the 650 OTUs that passed all filtering steps comprised a large and broad diversity of eukaryotes. Almost all major microbial lineages were sampled, from known abundant taxa in soils such as Ciliophora, Cercozoa, Apicomplexa and fungi, to rarer lineages in soil such as the mainly aquatic Bacillariophyceae (diatoms) and Chlorophyta (green algae) (Bahram et al., 2018; Foissner, 1987; Geisen, Cornelia, Jörg, & Michael, 2014; Geisen et al., 2018, 2015; Mahé et al., 2017). A few main protist groups lacked new OTUs altogether, including Cryptista, Retaria, Rhodophyceae and Glaucophyta, but these are almost exclusively aquatic and thus less likely to be recovered among soil sequences even if present in the environment at very low abundance (Geisen et al., 2018; Lallias et al., 2015; de Vargas et al., 2015). However, not all major groups typically widespread in soils were recovered with a correspondingly high sequence diversity. For example, this was the case for Amoebozoa, Excavata and Centrohelida, whose low diversity might be at least partially explained by primer bias (see Materials and Methods), and/or by the ecological conditions represented by the samples.

The availability of longer environmental sequences opens up the possibility to phylogenetically resolve environmental diversity with improved accuracy. Previous studies using both the 18S and the 28S genes recovered many relationships within and between major eukaryotic groups with greater resolution than that afforded by 18S alone (Moreira et al., 2007; Zhao et al., 2012). The use of both genes was proposed to more robustly derive the origin of environmental sequences, particularly in the case of fast-evolving taxa, but this was based on Sanger sequencing of clone libraries (Marande, López-García, & Moreira, 2009). Near full-length 18S amplicons and even longer fragments including parts of the 28S gene have also recently been sequenced with PacBio for group-specific investigations, demonstrating that long-read high-throughput sequencing is a promising complement to Illumina for investigating the environmental diversity of eukaryotes (Heeger et al., 2018; Orr et al., 2018). Here, we extended the approach to ~4,500 bp of the rDNA operon across the whole phylogenetic diversity of eukaryotes. We built a combined 18S + 28S tree of eukaryotes that is globally well resolved and can serve as a robust phylogenetic framework to describe the environmental diversity in samples (Figure 4). Comparisons to the 18S region alone of the queries (~1,200 bp) provided a similar overall topology to the combined tree, but with lower overall resolution (Figure S2). Furthermore, some key groups in the apicomplexan phylogeny were either missing or not supported by the 18S-only tree, a pattern that was also recovered by previous analyses (Simdyanov et al., 2017, 2018). Together, our phylogenetic comparisons revealed that 18S and 28S together provide increased resolution compared to 18S alone, but the differences between single- and two-gene trees vary across groups.

To assign taxonomy to the long environmental reads, we applied a novel phylogeny-aware approach that enables robust annotation to be derived even in the absence of closely related

references. Most commonly, taxonomic annotation is conducted by similarity comparison to reference databases (e.g., in Mahé et al., 2017; de Vargas et al., 2015). Similarity works well when closely related references are available, although it requires the use of arbitrary similarity cutoffs without biological grounding below which sequences are considered of unknown origin (Bahram et al., 2018; Stoeck et al., 2010). To enable the use of phylogenetics with short environmental reads, methods such as the EPA have been recently developed and successfully applied to microbial diversity (Bass et al., 2018; Mahé et al., 2017). Whilst the need for similarity cutoffs is alleviated, the EPA still requires longer reference sequences to build a stable evolutionary framework and thus does not fully overcome the limitations of short read sequencing when references are lacking. Whilst our method relies partly on the EPA, it makes explicit use of environmental sequences to build a reference tree and computes a confidence score for each taxonomic rank. We show that it provides accurate taxonomic annotation with ranks corresponding to the phylogenetic position of queries in the reference tree—higher ranks correspond to deeper branches in the phylogeny—and that it performs better than similarity-based methods for divergent sequences ($\leq 90\%$ similarity). Comparison with the classical use of EPA with V4 reads revealed that whilst the overall annotations were similar, our approach utilizing long queries led to higher confidence scores. It was also more informative than placing short reads on a reference phylogeny, because the long queries directly contributed to the phylogenetic inference by filling gaps between references. Thus, the relationships between the queries themselves can be determined to reveal whether they cluster around known sequences or form entirely new clades.

One of the main benefits of our approach is that it provides both the 18S and the 28S genes for the *same amplicon*. The 18S gene has long been the reference molecular marker for environmental studies of protist diversity (Diez et al., 2001; López-García et al., 2001; Massana, Balagué, et al., 2004; Massana et al., 2015; Moon-Van Der Staay et al., 2001; de Vargas et al., 2015). With this approach, each 18S sequence should be paired with its 28S counterpart (or ITS). As a result, we rapidly generated a massive increase of 28S sequence diversity for which the attached 18S provides a direct link to the much larger availability of 18S sequences contained in databases such as SILVA, PR2 or GenBank. As a point of comparison, the new sequences produced in this study alone represented the majority (58%) of all broad eukaryote diversity for which we could gather reference sequences for both genes. At lower taxonomic ranks, the increase in sequence diversity can be even more significant. For example, we found a total of only nine species of gregarines (Apicomplexa) that have both 18S and 28S genes in public databases. Here, we obtained 56 new gregarine OTUs, corresponding to a six-fold increase in diversity for this group. Thus, we suggest that the newly generated long environmental sequences can be used in future studies as taxonomically annotated “anchor” sequences to fill phylogenetic gaps in addition to the more traditional Sanger reference sequences.

In conclusion, we demonstrate several advantages of using high-throughput long-sequence metabarcoding for environmental studies of microbial eukaryote diversity. With longer reads comes improved phylogenetic signal, and we show that it is possible to employ a full phylogenetic approach to taxonomically classify sequences and obtain a robust evolutionary framework of environmental diversity. This approach can be adapted for use with other emerging long-read technologies, such as Nanopore sequencing, and may prove particularly powerful in combination with even higher-throughput sequencing technologies such as Illumina. Indeed, it will then be possible to map shorter but more abundant reads on a much more comprehensive reference phylogeny obtained from the same environments. The importance of eDNA studies continues to grow in fields as varied as conservation biology, evolutionary biology and ecology. Long metabarcoding of the eukaryotic rDNA operon will undoubtedly play an increasingly important role in the near future.

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AUTHOR CONTRIBUTIONS

F.B. and D.B. conceived the study. G.B. and S.H. provided soil samples from the UK and performed DNA extraction. R.F. and D.B. performed the wet laboratory experiments. M.J. performed and/or coordinated most of the bioinformatic analyses, in close connection with F.B., P.B., L.C., A.K. and A.S. F.B., D.B. and M.J. wrote the first complete draft of the manuscript, and all authors subsequently contributed to the final version.

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DATA ACCESSIBILITY

Raw PacBio Sequel reads have been submitted to the ENA database under accession number PRJEB25197. Detailed software commands and custom scripts used in the read curation pipeline are available on GitHub (<https://github.com/Pbdas/long-reads>).

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SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section.

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