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Environmental DNA metabarcoding of lake fish communities reflects long-term data from established survey methods

BERND HÄNFLING,*1 LORI LAWSON HANDLEY,*1 DANIEL S. READ,† CHRISTOPH HAHN,* JIANLONG LI,* PAUL NICHOLS,* ROSETTA C. BLACKMAN,* ANNA OLIVER† and IAN J. WINFIELD‡

*Evolutionary and Environmental Genomics Group (@EvoHull), School of Biological, Biomedical and Environmental Sciences, University of Hull (UoH), Cottingham Road, Hull HU6 7RX, UK, †Centre for Ecology & Hydrology (CEH), Benson Lane, Crowmarsh Gifford, Wallingford, Oxfordshire OX10 8BB, UK, ‡Lake Ecosystems Group, Centre for Ecology & Hydrology (CEH), Lancaster Environment Centre, Library Avenue, Bailrigg, Lancaster LA1 4AP, UK

Abstract

Organisms continuously release DNA into their environments via shed cells, excreta, gametes and decaying material. Analysis of this 'environmental DNA' (eDNA) is revolutionizing biodiversity monitoring. eDNA outperforms many established survey methods for targeted detection of single species, but few studies have investigated how well eDNA reflects whole communities of organisms in natural environments. We investigated whether eDNA can recover accurate qualitative and quantitative information about fish communities in large lakes, by comparison to the most comprehensive long-term gill-net data set available in the UK. Seventy-eight 2L water samples were collected along depth profile transects, gill-net sites and from the shoreline in three large, deep lakes (Windermere, Bassenthwaite Lake and Derwent Water) in the English Lake District. Water samples were assayed by eDNA metabarcoding of the mitochondrial 12S and cytochrome b regions. Fourteen of the 16 species historically recorded in Windermere were detected using eDNA, compared to four species in the most recent gill-net survey, demonstrating eDNA is extremely sensitive for detecting species. A key question for biodiversity monitoring is whether eDNA can accurately estimate abundance. To test this, we used the number of sequence reads per species and the proportion of sampling sites in which a species was detected with eDNA (i.e. site occupancy) as proxies for abundance. eDNA abundance data consistently correlated with rank abundance estimates from established surveys. These results demonstrate that eDNA metabarcoding can describe fish communities in large lakes, both qualitatively and quantitatively, and has great potential as a complementary tool to established monitoring methods.

Keywords: EC Water Framework Directive, eDNA, environmental DNA, fish monitoring, lakes, lentic systems, metabarcoding

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Introduction

Rapid monitoring of changes in biodiversity in response to climate change or other anthropogenic pressures is

Correspondence: Bernd Hänfling, Fax: +44 1482 465458; E-mail: b.haenfling@hull.ac.uk
¹Joint first authors.

imperative, but the time and resources required to generate the necessary data are a major constraint in conservation management and ecological research. This is particularly relevant in large lake ecosystems, where for a number of taxa, established methods currently struggle to deliver the required data to fulfil legislative obligations such as the EC Water Framework (European Communities 2000) and corresponding legislation elsewhere in the

world. This difficulty is particularly marked for fish, for which all established sampling methods have various forms of bias (e.g. Kubečka et al. 2009) and for which biological sampling is typically laborious and destructive (e.g. Argillier et al. 2013). Arguably, the biggest recent development in biodiversity monitoring is the analysis of environmental DNA (eDNA), which refers to DNA released by organisms into their environment, for example, in the form of shed cells, excreta or decaying matter. eDNA has great potential for biodiversity monitoring as it is noninvasive, can detect rare or elusive species that are difficult to detect using established methods, and can distinguish cryptic species or juvenile stages that are difficult to identify taxonomically (as reviewed in Bohmann et al. 2014; Lawson Handley 2015; Rees et al. 2015). Aquatic environments are particularly suited to eDNA analysis as DNA disperses rapidly in the water column and is more homogeneously distributed than in soil or other solid substrates.

The application of eDNA has so far largely focused on targeted detection of one or a few species using standard or quantitative polymerase chain reaction (qPCR). Such targeted eDNA assays have proven highly successful for detecting individual species from a wide range of taxonomic groups in aquatic environments (see Table 1 in Lawson Handley (2015) for a summary). For example, a recent eDNA study targeting great crested newts, Triturus cristatus, demonstrated high repeatability and substantially higher detection rates for eDNA compared to established survey methods (Biggs et al. 2015). The characterization of entire communities is not feasible using such species-specific approaches due to the complexity of most ecosystems. An alternative approach is to simultaneously screen whole communities of organisms using eDNA metabarcoding. Here, community DNA is PCR-amplified using broad-range primers, and sequenced on a high-throughput sequencing (HTS) platform (reviewed by Lawson Handley 2015). Direct metabarcoding of homogenized community samples is revolutionizing our understanding of the diversity of microscopic eukaryotes (Bik et al. 2012) in environments that are notoriously difficult to study, such as soil (Creer et al. 2010) and the deep sea (Fonseca et al. 2010). Metabarcoding of macrobial eDNA is still in its infancy, but the field is moving forward at a fast pace. The first studies focussed on describing fish communities in tanks or aquaria (Evans et al. 2015; Kelly et al. 2014; Mahon et al. 2014; Miya et al. 2015) or on a small scale in natural settings (Thomsen et al. 2012a,b). Recent refinements of the method, including more rigorous testing in aquaria (Miya et al. 2015) and in marine (Miya et al. 2015; Valentini et al. 2016) and freshwater habitats (Valentini et al. 2016), have confirmed the method is extremely sensitive for detecting rare species, and describing their presence/absence.

Table 1 Species previously recorded in the study lakes or recorded with eDNA. Full scientific, common names and three letter codes used in figures are given

Scientific name	Common name	Code	Previously recorded in study lakes
Abramis brama	Common bream	BRE	Yes
Anguilla anguilla	European eel	EEL	Yes
Barbatula barbatula	Stone loach	LOA	Yes
Coregonus albula	Vendace	VEN	Yes
Cottus gobio	Bullhead	BUL	Yes
Cyprinus carpio	Common carp	CAR	No
Esox lucius	Pike	PIK	Yes
Gasterosteus aculeatus	Three-spined stickleback	3SS	Yes
Gymnocephalus cernua (=cernuus)	Ruffe	RUF	Yes
Lampetra fluviatilis	River lamprey	RLA	Yes
Leucaspius deliniatus	Sunbleak	SUN	No
Leuciscus leuciscus	Dace	DAC	Yes
Oncorhynchus mykiss	Rainbow trout	RTR	No
Osmerus eperlanus	Smelt	SME	No
Perca fluviatilis	Perch	PER	Yes
Petromyzon marinus	Sea lamprey	SLA	Yes
Phoxinus phoxinus	Minnow	MIN	Yes
Platichthys flesus	Flounder	FLO	No
Pseudorasbora parva	Topmouth gudgeon	TMG	No
Pungitius pungitius	Nine-spined stickleback	9SS	No
Rutilus rutilus	Roach	ROA	Yes
Salmo salar	Atlantic salmon	SAL	Yes
Salmo trutta	Brown trout	BTR	Yes
Salvelinus alpinus	Arctic charr	CHA	Yes
Scardinius erythrophthalmus	Rudd	RUD	Yes
Squalius cephalus (=Leuciscus cephalus)	Chub	CHU	Yes
Tinca tinca	Tench	TEN	Yes
Umbra pygmaea	Mudminnow	MUD	No

Important questions remain though about the efficacy of eDNA metabarcoding for obtaining accurate estimates of species abundance and biomass. Obtaining quantitative estimates from eDNA is challenging because of the large number of factors that influence DNA dynamics in the environment (reviewed by Barnes et al. 2014; Lawson Handley 2015) and because of the many opportunities for bias during laboratory steps (sampling, DNA extraction, PCR), sequencing and bioinformatics stages (Ficetola et al. 2015; Yu et al. 2012). In metabarcoding studies, in principle, the number of sequences per taxon (or 'operational taxonomic unit') could be taken as an estimator of species biomass, but unfortunately in practice, this relationship is not a simple one. For example, Kelly et al. (2014) demonstrated a

perfect correlation between rank abundance of eDNA sequences representing four fish genera and rank biomass in a large aquarium, but the actual number of sequence reads was not correlated with biomass. Similarly, Evans et al. (2015) found only a modest positive relationship between the number of sequence reads and abundance of eight fish and one amphibian species in mesocosm experiments. A second approach that may be more promising for estimating abundance is to carry out comprehensive spatial and temporal sampling of a given environment and calculate the proportion of sites in which a species is detected with eDNA. Such 'site occupancy' data are often collected in ecological studies and can be used as a proxy for abundance (MacKenzie & Nichols 2004; MacKenzie et al. 2002). Recent studies indicate this approach could be very promising for analysing eDNA data from both targeted assays (Hunter et al. 2015; Pilliod et al. 2013; Schmidt et al. 2013) and metabarcoding data (Valentini et al. 2016).

How well eDNA metabarcoding performs compared to established survey methods for generating both qualitative (presence/absence) and quantitative (abundance/biomass) data remains a key question in the development of the technology for biodiversity monitoring. Here, we addressed this question by comparing eDNA metabarcoding data to the most comprehensive long-term data available for lake fish populations in the UK. We carried out sampling in three large, deep lakes (Windermere, Bassenthwaite Lake and Derwent Water) in the English Lake District, which are the best-studied lakes in the UK in terms of their fish fauna. First, we developed a workflow for lake fish eDNA metabarcoding, which included building an appropriate reference database of mitochondrial 12S and cytochrome b (CytB) genes, testing primer combinations and developing pipelines for eDNA analyses from sampling to bioinformatics. Second, we carried out water sampling along depth profile transects, at gill-net survey sites and at shoreline locations within the lakes. Finally, we compared the qualitative and quantitative results from eDNA metabarcoding with long-term and recent gillnet survey data sets to investigate the performance of eDNA against established methods.

Material and methods

Sampling

Sampling was carried out in three natural lakes (Bassenthwaite Lake, Derwent Water and Windermere) in the English Lake District, UK, that have been intensively studied in terms of their fish populations, physiochemical and other biological properties for many years (Maberly et al. 2011; Fig. 1). Fish populations in these three lakes have been monitored since the early 1990s (Bassenthwaite Lake and Derwent Water, e.g. Winfield et al. 2012a, 2015b) or early 1940s (Windermere, e.g. Winfield et al. 2008a, 2015b). This monitoring has been performed using gill-netting, trapping, hydroacoustics or analysis of recreational anglers' catches and constitutes the best long-term lake fish data sets in the UK. Windermere, England's largest natural lake (surface area 1480 ha, maximum depth 64 m), is composed of two distinct basins with different physical, chemical and ecological characteristics (North Basin: surface area of 810 ha, maximum depth 64 m, mesotrophic; South Basin: surface area 670 ha, maximum depth 44 m, eutrophic). Bassenthwaite Lake (surface area 528 ha, maximum depth 19 m, eutrophic) and Derwent Water (surface area 535 ha, maximum depth 22 m, mesotrophic) are also among the largest lakes in England and are linked by the River Derwent.

In total, 30 offshore samples were collected from each of the two Windermere basins. Additionally, six samples were collected opportunistically from a small area of the shoreline at the Northern end of the South Basin. Water samples were collected from Windermere during 28-30 January 2015. Most offshore samples were collected along three transects with approximately 1 km sampling interval between sites. Transects 1, 2 and 3 ran along the 5 m, 20 m depth contour and the lake midline, respectively (Fig. 1). The sampling depth for transects 1, 2 and 3 was 2 m, 10 m and 20 m, respectively. This sampling scheme covered 7 of the 10 sites that are used for annual gill-net surveys (Winfield et al. 2015b). Water samples were also collected at the 3 remaining gill-net sites (Fig. 1). At the deepest point along the midline transect in both North (approximate depth 64 m) and South Basin (approximate depth 44 m), a depth profile was collected. The North Basin depth transect was collected at 0-10-20-30-40-50-60 m depth and the South Basin depth transect was collected at 0-10-20-30-40 m. (Fig. 1). Water samples were also collected at 5 gill-net sites (Winfield et al. 2015a) and one shore site per lake at both Bassenthwaite Lake and Derwent Water (Fig. 1) on 10 February 2015. The total number of samples (excluding blanks) was therefore N = 78.

Offshore water sampling was carried out by boat using a Friedinger (Windermere) or Ruttner (Bassenthwaite Lake and Derwent Water) sampler (Fig. S1, Supporting information) deployed at a specified depth. For each 2 L water sample, five 400 mL subsamples were collected in proximity of 100 m around the sampling point, and pooled in a sterile plastic bottle (Fig. S1, Supporting information). The GPS location was recorded at the sampling mid-point (Appendices 1 and 2, Supporting information). Between samples, sampling equipment was sterilized by washing in 10% of a

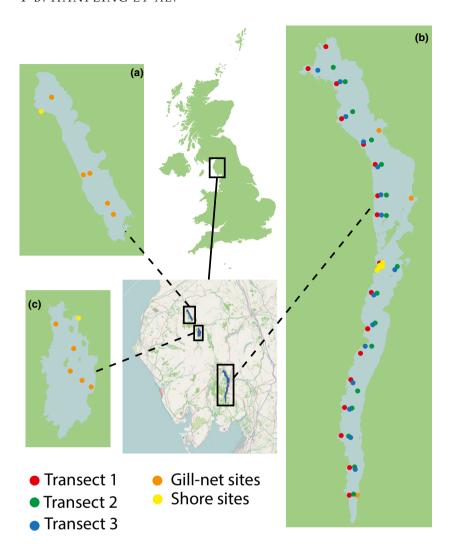


Fig. 1 Sampling sites in the three study lakes (a) Bassenthwaite Lake, (b) Derwent Water and (c) Windermere in the English Lake District (UK). Samples were collected from gill-net sites (orange circles) and single shoreline sites (yellow circles) in Bassenthwaite Lake and Derwent Water. In Windermere, samples were collected along transects following the 5 m (red circles), 20 m (green circles) and midline (blue circles) depth profiles, as well as additional gill-net and shoreline sites.

commercial bleach solution (containing <3% sodium hypochlorite) followed by 10% microsol detergent (Anachem, UK) and rinsed with purified water (Fig. S1, Supporting information). The sampler was then rinsed again in lake water at the next sampling location. Two litres of purified water was rinsed through the sampler following decontamination after every 5 samples, and the water retained as a sampling blank to allow us to check for contamination during sampling. Shoreline samples were collected by immersing a sterile 2-L plastic bottle by hand. For each sample, five 400 mL samples were collected from within a 100 m stretch of shoreline and pooled. All samples were stored in an insulated box at approximately 4 °C until filtration.

eDNA capture, extraction, amplification, library preparation and sequencing

The full 2 L of each sample was filtered through sterile $0.45\text{-}\mu m$ cellulose nitrate membrane filters and pads

(47 mm diameter; Whatman, GE Healthcare, UK) using Nalgene filtration units in combination with a vacuum pump (Fig. S1, Supporting information). Most samples required one filter and filtered in less than an hour. For more turbid and thus slow to filter samples, a second filter was used. Filtration equipment was sterilized in 10% commercial bleach solution for 10 min then rinsed with 10% microsol and purified water after each filtration. Filtration blanks (2 L purified water) were run before the first filtration and then approximately after every sixth sample, to test for contamination at the filtration stage. Windermere samples were filtered within 8 hours of collection in a lakeside laboratory (within the facilities of the Freshwater Biological Association, Windermere) that is not used for handling fish or DNA and was decontaminated before use by bleaching floors and surfaces. Samples from Bassenthwaite Lake and Derwent Water were filtered in a dedicated eDNA facility at the University of Hull within 12 h of collection. Detailed operating procedures are in place in our eDNA laboratory which are aimed at avoiding contamination and access to the laboratory is strictly limited to staff who are familiar with these procedures. DNA was extracted from filters using the PowerWater DNA Isolation Kit (MoBio Laboratories, Inc. Carlsbad, USA) using the manufacturer's instructions.

Full details of the steps involved in reference database construction, in silico and in vitro primer testing, including PCR conditions, are given in the Appendix S6 (Supporting information). Briefly, we compiled custom, phylogenetically curated reference databases (Appendix S6 and Fig. S2, Supporting information) for standard mitochondrial fish DNA barcoding genes (12S and cytochrome b) for 67 freshwater fish species including all those recorded in the UK and additional nonnative species that could potentially be present (Table S1, Supporting information). A number of published primers (Table S2, Supporting information) were evaluated against these databases in silico for conservation of primer binding sites and species resolution of the resulting PCR amplicons (Table S3, Supporting information) using the program EcoPCR (Ficetola et al. 2010). Two previously published primer pairs, which amplify fragments of contrasting length, from two different mtDNA regions, were selected for metabarcoding, as no single primer pair resolved all species (Table S3, Supporting information). The primer pair 12S_F1 and 12S_R1 (Table S2, Supporting information) amplifies a ~106-bp fragment of the mitochondrial 12S gene. These primers were designed and tested in silico (Riaz et al. (2011)) and used in a large marine mesocosm eDNA metabarcoding study of bony fish communities (Kelly et al. 2014). The second selected primer pair, CytB L14841 and CytB H15149 (Table S2, Supporting information), amplifies a 460-bp fragment of the cytochrome b gene (CytB) gene and has been used commonly for standard DNA barcoding of fishes (Kocher et al. 1989). Selected primer pairs were then tested in vitro on 22 species, first in individual reactions (Fig. S3, Supporting information) to check consistency of amplification across taxa and second in 10 mock communities to evaluate whether all species amplified in competitive mixed assemblages. Mock communities were generated from spectrophotometer-quantified DNA extractions of same 22 species (Appendix S6 and Table S4, Supporting information), and community samples were sequenced via metabarcoding as detailed below.

Samples for metabarcoding were PCR-amplified with a one-step library preparation protocol using, for each locus, 8 individually tagged forward primers and 12 individually tagged reverse primers allowing for 96 uniquely dual-indexed combinations (Kozich *et al.* 2013). All collection and extraction blanks were

included in PCRs, and contamination during PCR was evaluated by 'amplifying' all 96 combinations of tagged primers with purified water and checking on ethidium bromide-stained agarose gels. PCRs were replicated three times for each sample, and pooled to minimize bias in individual PCRs (see Appendix S6, Supporting information for full PCR conditions). Each library was normalized to approximately 1-2 ng/µL PCR product per sample using the SequalPrep Normalization Plate Kit (Invitrogen, Life Technologies) and samples subsequently pooled. Libraries were then quantified by qPCR (average of three replicate quantifications) using the KAPA Illumina Library Quantification Kit on a Roche LightCycler Real-Time PCR machine using manufacturer's guidelines. Libraries were run at a 6 pM concentration on an Illumina MiSeq using the 2 × 300 bp V3 chemistry. To improve clustering during the initial sequencing cycles, 10% of PhiX genomic library was added.

Bioinformatics and data analysis

The program Trimmomatic 0.32 (Bolger et al. 2014) was used for quality trimming and removal of adapter sequences from the raw Illumina reads. Average read quality was assessed in 5-bp sliding windows starting from the 3'-end of the read, and reads were clipped until the average quality per window was above phred 30. All reads shorter than a defined minimum read length (12S - 90 bp; CytB - 100 bp) were discarded. Sequence pairs were subsequently merged into single high-quality reads using the program FLASH 1.2.11 (Magoč & Salzberg 2011). The remaining reads were screened for chimeric sequences against the curated reference databases using the 'uchime ref' function implemented in vsearch 1.1 (https://github.com/torognes/ vsearch). To remove redundancy, sequences were clustered at 100% identity using vsearch 1.1 (https:// github.com/torognes/vsearch). Clusters represented by less than 3 sequences were considered sequencing error and were omitted from further analyses. Nonredundant sets of query sequences were then compared to the respective curated nonredundant reference database using BLAST (Zhang et al. 2000). BLAST output was interpreted using a custom python function, which implements a lowest common ancestor (LCA) approach for taxonomic assignment similar to the strategy used by MEGAN (Huson et al. 2007). In brief, after the BLAST search we recorded the most significant matches to the reference database (yielding the top 10% bit scores) for each of the query sequences. If only a single taxon was present in the top 10%, the query was assigned directly to this taxon. If more than one reference taxon was present in the top 10%, the query was assigned to the lowest taxonomic level that was shared by all taxa in the list of most significant hits for this query. Sequences for which the best BLAST hit had a bit score below 80 or had less than 100%/95% identity (12S/CytB) to any sequence in the curated database were considered nontarget sequences. The custom bioinformatics pipeline used for data processing is available on Github (https://github.com/HullUnibioinformatics/metaBEAT). To assure full reproducibility of our analyses, we have deposited the entire workflow in an additional dedicated Github repository (https://github.com/HullUni-bioinformatics/Haenfling et al 2016). In order to obtain a qualitative assessment of the taxonomic diversity, nontarget sequences were pooled across all lake samples and subjected to a separate BLAST search against NCBI's complete nucleotide (nt) database. Taxonomic assignment for nontarget sequences was obtained using MEGAN 5.10.6 (Huson et al. 2007).

Filtered data were summarized in two ways for downstream analyses: (i) the number of sequence reads per species at each site (hereon referred to as read counts) and (ii) the proportion of sampling sites in which a given species was detected (hereon referred to as the site occupancy). To reduce the possibility of false positives, we only regarded a species as present at a given site if its sequence frequency exceeded a certain threshold level (proportion of all sequence reads in the sample). The choice of threshold level was guided by the analysis of sequence data from the mock communities and is explained in full in the Appendix S6 (and corresponding Tables S4 and S5, and Figs. S5 and S6, Supporting information). This analysis revealed that threshold levels of 0.3% and 1% were required for 12S and CytB, respectively, to omit all false positives in the mock communities (hereon referred to as Th100, Tables S4 and S5, and Fig. S5, Supporting information). At Th100, sequences of rare expected species were also lost from the mock community data (Tables S4 and S5, Supporting information) and the lake samples (Fig. S6, Supporting information). We therefore decided to apply slightly less conservative values of 0.1% and 0.2% for 12S and CytB, respectively, at which over 90% of false positives were omitted in the mock communities to the main analysis of lake samples (Th90). We also investigated the potential extent of contamination from tag jumping in our libraries by exploring the distribution of PhiX assigned to target samples (see Appendix S6 and Fig. S7 for full details, Supporting information). The level of PhiX contamination in our samples also indicated that our thresholds were appropriate to eliminate most of false positives created during the sequencing process. In 95% of the 12S and CytB libraries, the proportion of PhiX did not exceed 0.0015 and 0.001, respectively (with a corresponding maximum of 0.0023 and 0.0201).

All downstream analyses were performed in R v.3.1.3. (R Core Team 2015). Before investigating species detection and abundance estimation with eDNA, we first evaluated whether 12S and CytB data sets produced consistent results by calculating the Pearson's product–moment correlation coefficient for both read count and site occupancy in R v.3.1.3. (R Core Team 2015).

A flow chart summarizing our analytical pipeline, from reference database compilation to data analyses, is provided in Appendix 5 of the Supporting information.

Species detection using eDNA

To maintain a balanced sampling design, the Windermere shore sites, which were only collected in a small area of the South Basin, were excluded from all comparisons of species presence and abundance comparisons across basins.

First, we evaluated the performance of eDNA to detect species previously recorded in our four lake basins. Second, we used site occupancy data to investigate the spatial distribution of eDNA records within Windermere. It should be noted that full site occupancy modelling requires temporal replication to estimate the detection probability and the true proportion of occupied sites (MacKenzie et al. 2002). This was not possible during the current study, so our estimates of site occupancy are simply based on the presence/absence, and should be treated as preliminary. We explored whether there were differences in eDNA distribution between transects, between offshore and shoreline samples, along depth profiles, and between Windermere North and South Basins. A persistent difference in species composition between the two Windermere basins has been extensively described by established sampling methods and is linked to their contrasting trophic status (Winfield et al. 2008a, 2012b, 2008b). eDNA records from species with no preference for trophic state are consequently expected to be distributed throughout the lake, whereas eDNA from eutrophic-favouring species will be more predominant in the South than North Basin and eDNA from species that prefer less eutrophic conditions will be more predominant in the North than South Basin. Finally, we used sample-based rarefaction (Gotelli & Colwell 2010) to determine the number of samples needed to detect species present, focussing on Windermere, where sampling was spatially comprehensive. Rarefaction was performed with 499 randomizations in the R package Vegan (Oksanen et al. 2015) for CytB and 12S for the North and South Basins of Windermere combined. Only sequences corresponding to the 16 species previously recorded in Windermere were included in these analyses.

Comparison of data from eDNA and established survey methods

Summaries of fish community composition and abundance were produced for each of the four lake basins using a combination of data collected at six sites in each of our four lake basins in September 2014 using standardized survey gill-netting techniques (described in detail by Winfield et al. 2015a,b). Gill-net survey data alone are not sufficient to describe the whole fish community as this technique under-samples or even fails to record some species, even when they are locally abundant (e.g. those with an extremely shallow distribution such as bullhead, Cottus gobio, or elongate morphology such as eel, Anguilla anguilla). Gill-net data were therefore supplemented with published information (Maberly et al. 2011; Pickering 2001; Winfield et al. 2012a, 1996; Winfield & Durie 2004; Winfield et al. 2010, 2008b) to summarize fish community compositions. This information and IJW's expert opinion developed during 25 years of sampling the four lake basins were then used to assign each recorded species to an abundance rank, with a rank of 1 given to the most abundant species by numbers. The ranking produced in this way is probably very robust for the most abundant species which consistently appeared in the catches of the survey gill nets, but is probably less so for a few species which anglers' catches indicate are present in small numbers in each lake but which are very rarely or never recorded by scientific sampling. This entire expert opinion ranking process was undertaken prior to the eDNA analysis and therefore with no knowledge of the corresponding rankings. Further details of the results from established surveys are provided in Appendix S6 and Table S5 (Supporting information).

A series of correlations was performed to compare the fish abundance data generated from established surveys and eDNA metabarcoding. Specifically, the relationship between eDNA data (read count and site occupancy) and data from established surveys (rank abundance or biomass based on long-term expert opinion or actual numbers from September 2014 gill-net surveys) was investigated by calculating Spearman's rho (for rank correlations) and Pearson's product–moment correlation coefficient (for actual numbers, when data were normally distributed) in R v3.1.3 (R Core team 2015). The analyses were repeated for both loci and all four sampled basins.

Results

The in silico testing of primer pairs showed that both of the chosen 12S and CytB fragments could unambiguously distinguish all species which could potentially occur at the study sites (Tables S1 and S3, Supporting information). However, across the wider reference database a number of taxa could not be identified to species level. Lampetra planeri and L. fluviatilis, which are probably not reproductively isolated, could not be resolved by either fragment. Additionally, 12S did not distinguish species of the genera Salvelinus and Coregonus, three species of non-native Asian carp (Hypophthalmichthys nobilis, H. molitrix, Ctenopharyngodon idella) and two species of the family Percidae (Perca fluviatilis and Sander lucioperca). However, given that Percidae and the genera Coregonus and Salvelinus are represented only a single species each (Perca fluviatilis, Salvelinus alpinus and Coregonus albula, respectively) in the study area, we have attributed sequence counts for the higher taxonomic levels to these individual species for further downstream analysis. This was also confirmed by the CytB data which showed that no other members of these taxonomic groups were present. Both loci amplified consistently well across 22 target species in in vitro testing in single-species amplifications (Fig. S3, Supporting information). All 22 species were detected in the 12S mock communities (Table S4, Fig. S4a, Supporting information), whereas three species were not detected in the CytB mock community data (Table S5, Fig. S4 b, Supporting information and Appendix S6 for full details). Observed and expected number of sequence reads were not significantly different for either locus (12S $\chi^2 = 0.224$, df = 21, P > 0.05; CytB $\chi^2 = 0.367$, df = 21, P > 0.05 Fig. S4, Supporting information). Moreover, there was a significant correlation between the number of sequence reads/ng PCR template DNA for 12S and CytB (Pearson's r = 0.599, df = 20, P = 0.01, Fig. S4c, Supporting information).

Clear PCR bands were obtained for all 78 eDNA samples at both loci. In contrast, no target-sized bands were observed in the PCR negatives, collection or filtration blanks, and we therefore decided not to sequence these. The total sequence read count passing quality control per library, before removal of chimeric sequences, was 6 306 326 for 12S and 4 793 108 for CytB (average read count per sample 71 663 and 54 467, respectively). After chimera removal, the 12S and CytB libraries contained 2 698 144 and 3 161 608 sequences, respectively. This means that 43% of the raw data set was nonchimeric sequences for 12S and 66% for CytB. The final libraries, after removal of redundant sequences, contained 2 562 183 sequences for 12S and 3 012 249 sequences

for CytB, with average read counts per sample of 29 116 and 34 230, respectively. The proportion of target (fish) sequences ranging from 3.4% to 88.3% (average 23.5%) and from 0% to 100% (average 49.0%) for 12S and CytB, respectively. Most of the target sequence assignments in the lake samples were to species level with the exceptions mentioned above. The assignments to higher taxonomic levels were taken into account for calculation of total sequences read number per sample but otherwise not considered for further downstream analysis. For the CytB data of the mock communities, some genus-level sequence assignments were interpreted as belonging to specific species (for full details see Appendix S6 and Table S5, Supporting information). The full sequence count data for each primer pair are available in the Appendices S1 and S2, Supporting information).

High consistency was found between CytB and 12S in terms of both site occupancy (SO) and average read count (RC) (Fig. S8, Supporting information). Data from the two loci were significantly correlated (Pearson's r consistently P < 0.05) for all basins, for both SO and RC (Fig. S8, Supporting information). Consistent significant correlations were also found between SO and RC for each basin and locus (Fig. S9, Supporting information); therefore, only the results for site occupancy are presented in the following main text. All results based on read count data are provided in the Supporting information.

Species detection using eDNA

The gill-net survey of September 2014 detected 25% (4/ 16) of the previously recorded species in Windermere. By contrast, 14 of the 16 previously recorded species (i.e. 88%) were detected using 12S and 75% (12/16) using CytB across the entire lake. Within each Windermere basin, 13 previously recorded species were detected with 12S whereas 12 and 11 species were detected for the North and South Basins, respectively, with CytB (Fig. 2 a, b; Fig. S10, Supporting information). A number of additional species were also detected in Windermere, including C. carpio, Gymnocephalus cernuus, Leucaspius delineatus, O. mykiss, Osmerus eperlanus (12S), Platichthys flesus and Pseudorasbora parva (CytB). Two species that have been recorded in Windermere but are not present in the sequence data are the two lamprey species L. fluviatilis and Petromyzon marinus. In the 12S data set, the majority of potential false positives were found in a single sample from Windermere North Basin which was consequently omitted from all further analysis (sample W14). Gill-net sampling detected 60% (6/10) of the species known to be present in Bassenthwaite Lake whereas 90% (9/10) of species were detected using 12S and 70% (7/10) with CytB (Fig. 2c; Fig. S10, Supporting information). Additional species not previously recorded in Bassenthwaite included *Abramis brama* (CytB), and *Barbatula barbatula*, *G. aculeatus* and *S. erythrophthalmus* (12S, Fig. 2c). In Derwent Water, gill-net sampling in September 2014 detected 77% (7/9) recorded species, whereas 88% (8/9) of species were detected with 12S and 67% (6/9) with CytB (Fig. 2d; Fig. S10, Supporting information). The 12S assay detected an additional four species previously unrecorded, including *B. barbatula*, *G. aculeatus*, *Pungitius pungitius* and *S. erythrophthalmus*.

Sample-based rarefaction analyses on the combined Windermere data set indicated that approximately 10–25 samples captures the majority (~85%) of the taxa present in the entire sample although the number of samples required to achieve the same taxon coverage is higher for CytB (Fig. 3).

Estimating abundance with eDNA

There was a consistent, negative relationship between eDNA site occupancy and long-term rank (where rank abundance decreases from 1 to 16), and this correlation is highly significant for Windermere North and South Basins, for both loci (Fig. 4a, b, e, f). Similar trends were found for Bassenthwaite Lake and Derwent Water but correlations were not significant (Fig. 4c, d, g, h). The number of sequence reads was also significantly correlated with long-term rank in Windermere North and South Basins, for both loci (Fig. S11 a, b, e, f, Supporting information). Again similar trends were seen for Derwent Water and Bassenthwaite Lake, but only the correlation for Derwent Water at 12S was significant (Fig. S11 c, d, Supporting information).

Site occupancy and number of sequence reads were also compared against actual numbers sampled in the September 2014 gill-net surveys for all four basins (Figs. S12 and S13, Supporting information, respectively). There was a consistent positive relationship between abundance data from the recent gill-net surveys and eDNA (both read count and occupancy, and both loci), in spite of the small number of species (4–6) detected in the gill-net surveys and hence low statistical power in the analyses. However, only the correlations for CytB read count were consistently significant in all basins (Fig. S13 e–h, Supporting information), and this result may be driven by the high abundance and read count for *P. fluviatilis*.

Spatial distribution of eDNA records within Windermere

Comparing the distribution of eDNA data by transect indicates a slight trend for more species to be detected

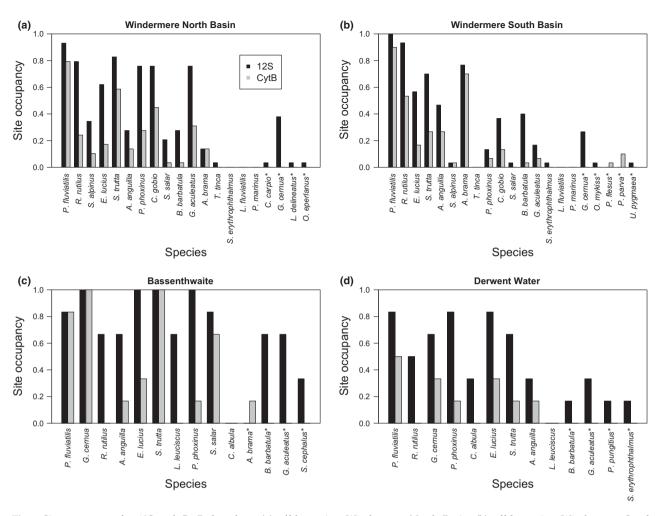


Fig. 2 Site occupancy for 12S and CytB data from (a) offshore sites Windermere North Basin, (b) offshore sites Windermere South Basin, (c) Bassenthwaite Lake and (d) Derwent Water. All species recorded previously are included. Previously recorded species are ordered according to their rank abundance within basin from established survey methods. Species that have not been recorded previously are indicated with an asterisk and are ordered alphabetically. Full species names are given in Table 1.

at inshore versus deeper mid-lake regions (Fig. 5). With 12S, 13 species were detected in samples from the 5-m transect compared to 10 from the midline. Twelve species were detected in the 6 geographically close shore samples. A similar trend was found for CytB, with 11 species detected in both 5-m transect and shore samples, compared to 8 in the midline (Fig. 5). Depth profiles in the North and South Basins revealed that eDNA from the majority of detected species was distributed throughout the water column (Fig. S14, Supporting information). Within the depth profiles, A. anguilla and S. alpinus were only detected in deep water in the North Basin (≥60 m and 30 m, respectively, Fig. S14 a and c, Supporting information). Similarly, in the South Basin depth profile P. phoxinus and S. salar were only detected at the deepest sampling point (40 m) (Fig. S14 b and c, Supporting information).

Site occupancy data based on 12S sequences were used to investigate the spatial distribution of each species recorded at more than two sites around Windermere (Fig. S15, Supporting information). The general pattern emerging from this analysis is that species-specific eDNA was not evenly distributed around the lake. Although some species such as P. fluviatilis, R. rutilus, E. lucius and S. trutta are recorded almost ubiquitously within the lake, eDNA from other species is predominantly found in one of the two basins. S. alpinus, P. phoxinus and G. aculeatus eDNA was common in the North Basin but very rare in the South Basin, whereas A. brama and A. anguilla eDNA was more common in South Basin (Fig. S15, Supporting information). Overall, the relative proportion of sequence read counts for different species across sample sites was significantly different between Windermere North and South Basins

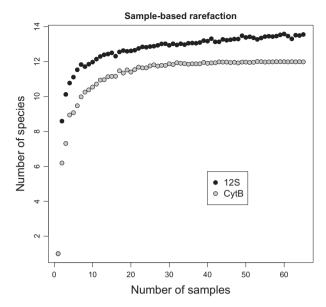


Fig. 3 Sample-based rarefaction analyses for Windermere. Only offshore samples and species recorded previously in Windermere are included in the analyses.

 $(\chi^2 = 47~817; df = 13; P < 0.001 and <math>\chi^2 = 134750;$ df = 11; P < 0.001 for 12S and CytB, respectively, Fig. 6a, b). A similar pattern was observed for the relative proportion of sites occupied ($\chi^2 = 61.43;$ df = 13; P < 0.001 and $\chi^2 = 48.65;$ df = 11; P < 0.001 for 12S and CytB, respectively, Fig. 6c, d). Distribution of eDNA in the two Windermere basins reflected the expected association between species and ecological condition. eDNA from species associated with eutrophic conditions (R. rutilus, T. tinca, S. erythrophthalmus, A. brama and A. anguilla) was more abundant in the South than North Basin, while eDNA from species that prefer less eutrophic conditions (S. salar, S. trutta, S. alpinus, P. phoxinus and C. gobio) was more abundant in the North than South Basin (Fig. 6).

Nonfish sequences

A large proportion of both 12S and CytB sequences could not be assigned to UK freshwater fish from the custom database, and were compared to the NCBI database using BLAST. Nonfish sequences included a wide range of species directly associated with aquatic habitats including mammals such as otter, *Lutra lutra* and birds, including moorhen, *Gallinula chloropus*; cormorant, *Phalacrocorax carbo* and various duck and geese species found within the UK. The list also included many other vertebrate species potentially occurring in the wider catchment area (Table S6, Supporting information) including domesticated farm animals such as cow, *Bos taurus*; sheep, *Ovis aries* and chicken, *Gallus*

gallus domesticus, and wild vertebrates such as red deer, Cervus elaphus; red squirrel, Sciurus vulgaris; red fox, Vulpes vulpes and tawny owl, Strix aluco. Sequences assigned to Homo sapiens were also abundant, likely present as genuine eDNA found in lake water due to the high degree of human interaction with the lakes through water sports, angling and waste water, or present as a laboratory contaminant. The primers appear to be largely vertebrate specific, except for low-level amplification of bacterial 16S detected in the 12S data set. No invertebrate sequences were identified.

Discussion

In this study, we used high-throughput sequencing of eDNA from the mitochondrial 12S and CytB genes to characterize the fish community composition in three large lakes (Lake Windermere, Derwent Water and Bassenthwaite Lake) in the UK. eDNA data were compared to comprehensive long-term data on fish distribution and abundance from established survey methods. eDNA outperformed established methods in terms of species detection. More surprisingly, eDNA data accurately reflected the rank abundance of species within the lake fish community, suggesting eDNA methods may be more quantitative than previously thought.

Comparison of eDNA and established methods for species detection

eDNA metabarcoding was effective in detecting fish species when compared against decades of data from established sampling techniques and other sources (as described most recently by Winfield et al. 2015a,b). In Windermere, 60 offshore (30 for each basin) and 6 shoreline samples were analysed and 14 of the 16 previously recorded species were detected. The two rarest species, river lamprey, L. fluviatilis and sea lamprey, P. marinus, were not detected in the eDNA data, but these species were unlikely to be present in the lakes at the time of sampling and temporally replicated sampling is required to address this issue. Other rare species such as tench, T. tinca and rudd, S. erythropthalmus were detected at low levels with 12S in the North and South Basins, respectively. The results of the rarefaction analysis on the Windermere data indicate that a detection probability of over 85% can be achieved with a substantially lower number of samples, approximately 10 for 12S and 25 for CytB. In contrast, only the four most common species were detected in the gill-net survey from 2014, which is typical of surveys (4–5 species have been typically sampled each year since 2011, Winfield et al. 2012c; Winfield et al. 2013; Winfield et al. 2014).

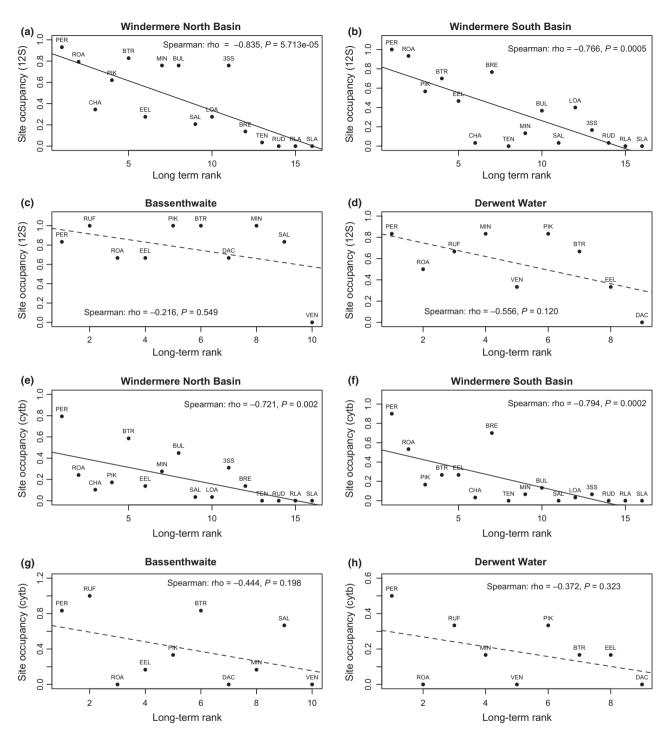


Fig. 4 Correlations between site occupancy data and long-term rank based on established surveys and expert opinion for all four basins and both 12S (a–d) and CytB (e–h), where 1 is the highest and 16 the lowest rank abundance. Species three letter codes are given in Table 1.

The eDNA results from Bassenthwaite Lake and Derwent Water were also remarkably concordant with the fish community based on long-term gill-netting (Thackeray *et al.* 2006) given that only six samples were collected per lake. All but the rarest species were detected

in Derwent Water and Bassenthwaite (dace, *L. leuciscus*, and vendace, *C. albula*, respectively) using 12S. Dace was however detected in Bassenthwaite, and vendace in Derwent Water with 12S, while neither species was detected with CytB. Dace has been recorded

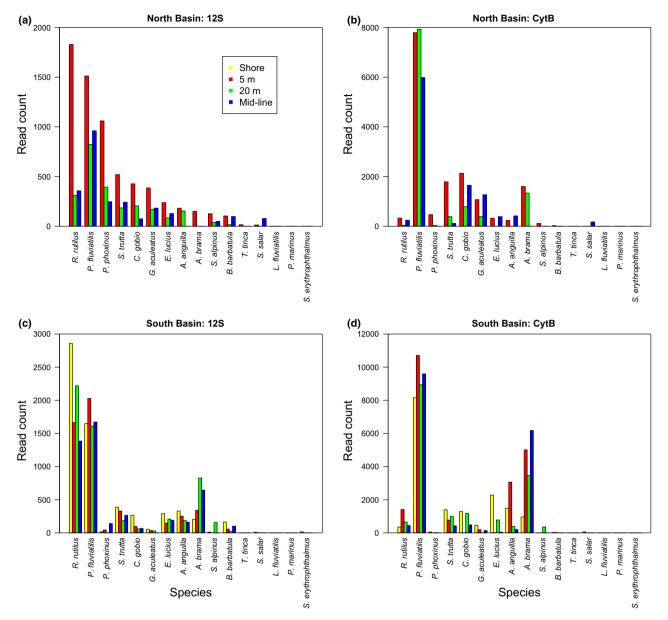


Fig. 5 Average number of sequence reads obtained per transect for Windermere North Basin (a, b) and South Basin (c, d) for both 12S (a, c) and CytB (b, d). Only species that have been recorded previously are included. Species are ordered according to their rank abundance within basin from established survey methods.

intermittently and in low numbers in Derwent Water within the last decade (Thackeray et al. 2006) but was not detected by gill-netting in 2014 (Winfield et al. 2015a). Vendace is known to occur only in a restricted deep area of Bassenthwaite Lake, and only three individuals have been recorded in gill-net surveys since 2000 (Winfield et al. in press). In these cases, DNA concentration might fall below the detection threshold of the PCR assay or those which were set for the bioinformatics analysis in order to reduce the possibility of 'false positives'. Roach, R. rutilus, on the other hand, is a common species in all four basins, but was not

detected with CytB in Bassenthwaite and Derwent Water. This species was also detected in the CytB mock community at lower-than-expected frequency, suggesting that the CytB primers may not amplify this species well in competitive reactions.

Overall, eDNA metabarcoding data produced a more comprehensive species list than gill-net surveys with a similar effort. The under-representation of species in gill-netting surveys is an acknowledged sampling artefact which has a number of causes including fish morphology (e.g. eel species are not susceptible to retention in gill nets), fine-scale spatial distribution (e.g.

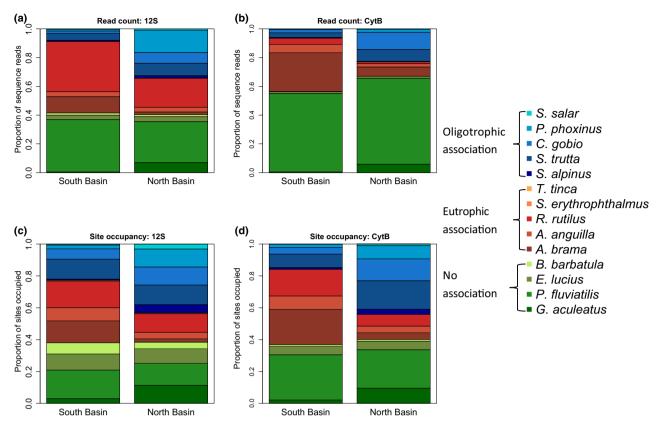


Fig. 6 Relative distribution of fish species and their ecological preferences in Windermere North Basin (mesotrophic) and South Basin (eutrophic) based on the proportion from the total number of sequence reads (a, b) and the relative proportion of sites occupied (c, d) reflecting the trophic status of the two basin.

three-spined stickleback may be limited to the extreme inshore where nets cannot be deployed) or movement patterns (e.g. bullhead may be unlikely to be sampled by gill nets due to their relatively limited movements). This corroborates results from Thomsen *et al.* (2012a) and Valentini *et al.* (2016) who showed that eDNA metabarcoding data detected more species of marine fish than alternative surveying techniques.

Detection of previously unrecorded species with eDNA

Eight previously unrecorded species were detected in Lake Windermere, four in Bassenthwaite Lake and four in Derwent Water. In most cases, these eDNA records were at very low occupancy (1 or 2 sites) and read counts (0.1–1.0%), just above our threshold for accepting a positive record. These records could be either genuine detections of species that have been missed with established methods, false positives from sequencing error (barcode misassignment, Deakin *et al.* 2014; or 'tag jumps' Schnell *et al.* 2015), laboratory or environmental contamination (i.e. the presence of DNA in the environment from, for example, the wider watershed, bird

faeces, waste water or fishing bait). The unexpected records probably originate from a combination of these factors, discussed below.

Only one of the eight previously unrecorded Windermere species, ruffe, G. cernua, was detected at high frequencies with eDNA. 12S sequences were present in 27% of the sites in the South Basin and 38% of the sites in the North Basin although the species was not detected with CytB. This species has been recently introduced to a number of Cumbrian lakes (Winfield et al. 2010) and is present in Rydal Water approximately 3 km upstream of Windermere. It is therefore possible that G. cernua has colonized Windermere and is present at very low abundance (below the detection limits of gill-netting programme) or that eDNA has been transported from the G. cernua populations upstream. Three kilometres is well within the range of eDNA transport distances that have previously been recorded (Deiner and Altermatt 2015). Absence of positive records with the long CytB fragment also suggests that only relatively degraded G. cernua DNA was present in the lake, lending further support to this hypothesis. Although this species was present in the mock communities, the

high frequency of occurrence means it is unlikely that this result can be explained by sequencing errors such as barcode misassignment.

The other seven previously unrecorded Windermere species (common carp, C. carpio; sunbleak, L. delineates; topmouth gudgeon, P. parva; rainbow trout, O. mykiss; smelt, O. eperlanus; flounder, P. flesus and mudminnow, U. pygmea) were detected at very low levels. The actual presence of *U. pygmea*, *L. delineates* and *P. parva* in Windermere seems extremely unlikely as their known distribution does not overlap with the Windermere catchment. Given that all three species were included in the mock communities, these records are most likely explained by low-level laboratory contamination or sequencing barcode misassignment from the mock communities into the samples (Deakin et al. 2014). O. mykiss, O. eperlanus and P. flesus do occur in the catchment, and the former two species are also a very popular dead bait used by pike anglers. As none of these species have been handled in the laboratory and pike anglers were active during the water sampling, it seems that such dead baiting or eDNA transport from other parts of the catchment are likely sources of eDNA for these species in the lake. C. carpio was recorded with both CytB and 12S at one of the shore sites. The fact that both markers were recorded at the same site indicates that common carp DNA and individuals might have been present in the lake water but highly localized and undetected by established sampling techniques. However, this species was also present in the mock communities, and therefore, laboratory contamination or 'tag jumping' cannot be excluded.

Four previously unrecorded species were detected in each of the Bassenthwaite and Derwent Water basins. Again most of these records were based on low sequence reads and site occupancy. The records for some species (common bream, A. brama in Bassenthwaite Lake, nine-spined stickleback, P. pungitius in Derwent Water) are most likely explained by barcode misassignment because they have never been recorded in the catchment but are present in the mock communities. The presence of the remaining species (stone loach, B. barbatula; three-spined stickleback, G. aculeatus; and rudd, S. cephalus) in the lakes or in the catchment cannot be so easily excluded. These records therefore could either represent environmental contamination or indicate that the species are present at low numbers and have not been detected by previous long-term gillnetting (summarized by Winfield et al. 2012a).

We quantified the level of background contamination using sequence information from mock communities and the level of PhiX contamination in target samples, which enabled us to choose a suitable threshold level for filtering the data for false positives without losing more information than necessary. Ultimately though, it is not possible to distinguish between false positives and true positives if they occur at the same frequency, and some rare species are probably lost with a threshold approach. Using consistency of presence across technical replicates as recently used by Port *et al.* (2016) might be a more suitable approach to control for false positives if rare species are of particular interest.

Use of eDNA for assessing relative abundance of lake fish

This study attempted to assess the relative abundance of individual species using their sequence read counts or site occupancy as proxies. Using read count data is a valid approach under the assumption that no significant bias is introduced during sampling, subsequent PCR or sequencing. However, this assumption is unrealistic, and previous studies have demonstrated that the relationship between abundance and read count is complex (e.g. Ficetola et al. 2015; Yu et al. 2012; Evans et al. 2015; Kelly et al. 2014). Site occupancy models have been developed to cope with multiple levels of bias and uncertainty (e.g. imperfect detection, MacKenzie et al. 2002) and are therefore highly promising for eDNA (Schmidt et al. 2013). As discussed in the Methods, full site occupancy modelling requires the estimation of detection probability from temporal sampling, which was beyond the scope of the present study. Our site occupancy estimates should therefore be treated as preliminary. Encouragingly though, read count and site occupancy data were correlated for each basin and each locus, suggesting that both measures of abundance are informative. As we discuss below though, and not surprisingly, site occupancy relies on comprehensive spatial sampling to obtain sufficient power for estimating abundance.

We found a consistent significant relationship between rank abundance and read count or occupancy data for both basins of Lake Windermere. This indicates both read count and occupancy are equally effective at estimating relative abundance under comprehensive spatial sampling. In Derwent Water and Bassenthwaite Lake, correlations with both abundance measures are weak and not significant with one exception (number of 12S sequence reads for Derwent Water). We suggest this is related to low statistical power from analysing only six samples per lake. There was also a consistent trend between eDNA and gill-net data, but the results are less conclusive due to low statistical power from the small number of species sampled in the gill-net survey. Although these results are generally encouraging, further work is critically needed to determine how robust eDNA is for estimating abundance. Increased spatial coverage of Bassenthwaite Lake and Derwent Water, together with temporal sampling to allow estimation of detection probability and site occupancy modelling in all basins, is a critical next step.

Spatial distribution of eDNA in Windermere

We investigated the spatial distribution of eDNA in Lake Windermere by comparing (i) offshore and shoreline samples, (ii) three depth profile transects and (iii) North and South Basins, which differ in their trophic status. First, more species were detected in shallower than in deep water, with 13 species detected along the 5-m contour, compared to 9 in the midline transect. Interestingly, 12 of the 16 previously recorded species were detected in the 6 shore samples, which were collected in close proximity to one another. This suggests eDNA could accumulate on the shoreline and that shoreline sampling could be adequate for the detection of most species. More rigorous sampling along the lake shore is needed to investigate this further. Second, we expected little difference along depth profile transects as our sampling was carried out in the winter, when water stratification has broken down. As predicted, within the depth transects the majority of species were detected throughout the water column but some, including the typically deep-water species Arctic charr, S. alpinus, were only detected at the deepest sampling points, indicating that surface water sampling might be ineffective in deeper lakes. Given the small scale of this experiment, the results regarding vertical sampling should be regarded as preliminary. Third, we hypothesized that eDNA from species associated with less eutrophic (i.e. mesotrophic) conditions would be more abundant in the North Basin, while eDNA from species associated with more eutrophic conditions should be more abundant in the South Basin, and species with no preference should be detected throughout the lake. We observed clear differences in the spatial distribution of eDNA, consistent with this hypothesis. These results are consistent with long-term data sets from trapping, gill-netting and recreational anglers' catches (Winfield et al. 2008a,b, 2011; Craig et al. 2015; Winfield et al. 2015b). For example, established methods have found perch, P. fluviatilis and pike, E. lucius consistently in both basins (Craig et al. 2015; Winfield et al. 2008a; respectively) while S. alpinus is much more abundant in the North than in the South Basin (Winfield et al. 2008b, 2015b) and A. brama, although a relatively minor component of the Windermere fish community is consistently more abundant in the South than in the North Basin (Winfield et al. 2011).

Technical approach and the use of 12S or CytB as a marker

In the present study, we chose to validate the assays by sequencing mock communities, constructed from 22 species of fish, on the same flow cell as the eDNA samples. Although this allows for the success of the assay to be assessed within the same sequencing library as the samples, this approach may cause problems due to the low-level misassignment of sequences from the mock community to the samples. For future studies, we would recommend not including mock communities in the same library, or only including species that have no chance of being found in the eDNA samples and to sequence all negative controls and blanks.

Both markers were generally consistent in terms of the number of read counts and occupancy data generated, although clear advantages and disadvantages were associated with each marker. All species were detected in the mock communities with 12S whereas three were undetected with CytB. In the eDNA samples, site occupancy was higher, and more species were detected with 12S than CytB, as discussed earlier. Differences in amplification success could be due to fragment size (~100 bp for 12S and 460 bp for CytB), mismatches in primer binding sites or both. Given that eDNA degrades rapidly in the environment (Barnes et al. 2014; Rees et al. 2014), the difference in detection is possibly a result of longer persistence of the shorter 12S fragment in lake water. This may allow for dispersion of eDNA across a larger geographical scale, increasing the probability of detection at any site. Consequently, it may be that detection of the longer CytB fragment indicates the species is present closer to where the water sample was taken, while 12S fragments may have originated from some distance away either within the lake or even up its tributaries. Using a longer fragment may be useful for pinpointing the exact location of species, but using a shorter fragment might be more useful for simply detecting the presence of a species anywhere in the water body using a limited number of subsamples. An additional aspect to consider is the persistence of eDNA in sediments, which has been shown to be considerably longer when compared to the water column (Turner et al. 2014). Differential persistence of the different-sized fragments, and resuspension of eDNA during rain events could account for historical eDNA being detected. However, differences in primer specificity and efficiency between the two genes prevent conclusive answers to these issues, and this issue warrants further systematic exploration through experimental approaches and analysing a wider range of eDNA fragment lengths.

Use of eDNA to survey nonfish vertebrates

This study also offers some insights into the feasibility of eDNA techniques for the wider assessment of nonfish vertebrates associated with lakes and their immediate catchments. The majority of the 12S and CytB sequences generated did not match the comprehensive UK fish reference database used and nonfish sequences could be assigned to a wide range of vertebrate species including mammals, birds, amphibians and some marine fish species (known to be used in the lakes as dead bait by anglers) which were not included in our reference database. Moreover, the primers used appear to be largely vertebrate specific as no invertebrate sequences were identified, although many such species are present. Consequently, the eDNA approach employed in this study may have further applications in the qualitative but extensive highlevel survey of nonfish vertebrate taxa occurring in lake catchments.

Conclusions

The present investigation was driven primarily by the need to develop robust and cost-effective lake fish assessments to meet the requirements of the EU Water Framework Directive and other international and national environmental legislation. It is universally agreed that there is no single sampling method that can produce all of the information needed to make such assessments, but even the use of a combination of methods from the range of established techniques still presents an incomplete picture with varying degrees of bias and incomplete coverage (Kubečka et al. 2009). The findings of the present study indicate that eDNA approaches can make a very significant contribution to this challenging task. The results were consistent with our understanding of the fish communities of three large, deep lakes based on long-term monitoring using established techniques. Moreover, this work moved beyond a simple presence/absence analysis to produce indications of the relative abundance of species, which were again consistent with earlier assessments and ecological interpretations. Although the eDNA approach cannot produce information on individual condition or population characteristics such as growth curves, it proved to be very effective at producing robust data at the community level which is undoubtedly the most challenging task for established sampling methods.

eDNA is arguably one of the most rapidly expanding areas of research in molecular ecology, but there is much to learn before methods such as the one described here can be deployed for biological monitoring, particularly under legislative or sensitive

circumstances. Temporal sampling is an essential next step from the current study, to account for imperfect detection and fully test the site occupancy modelling approach, and to investigate the effects of water stratification on the spatial distribution of eDNA. More generally, there is a pressing need to develop and demonstrate the wider applicability of eDNA to a greater range of water bodies (such as those with varied chemical and physical properties) as well as other animal and plant communities.

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B.H., L.L.H. and I.J.W. conceived the study; B.H., L.L.H., I.J.W., J.L. and R.B. carried out the fieldwork.

I.J.W. prepared fish abundance data from established method surveys. P.N., J.L and R.B. carried out all presequencing laboratory work. D.R. assisted in the design of the molecular assays and carried out Illumina sequencing and the initial steps of the raw data analysis; A.O. assisted with the Illumina sequencing; C.H. assembled the bioinformatics pipeline and reference database and wrote the relevant sections of the manuscript. B.H. and L.L.H. performed the statistical analyses. B.H., L.L.H., I.J.W. and D.R. wrote the manuscript; all authors commented on the final draft.

Data accessibility

All de novo sequences generated through Sanger sequencing made available directed through our archived analysis pipeline on Github. Accession numbers and taxon affiliations of all curated sequences are available as electronic Appendices. Raw Illumina read data have been submitted to NCBI (BioProject: PRJNA313432; BioSample Accession nos.: SAMN04530423-SAMN04530510; SRA accessions: SRR3359939-SRR3360124). To assure full reproducibility of our analyses, we have deposited the entire bioinformatics workflow in a dedicated Github repository, which also contains the curated reference databases and further supplementary data, such as taxon specific read counts for each sample as tables (https://github.com/HullUni-bioinformatics/Haenfling_ et al 2016; the repository is permanently archived with Zenodo (DOI 10.5281/zenodo.49823). Our custom data processing pipeline is available on Github (https:// github.com/HullUni-bioinformatics/metaBEAT).

Supporting information

Additional supporting information may be found in the online version of this article.

Fig. S2a Maximum Likelihood tree of all 12S sequences from the reference data base.

Fig. S2b Maximum Likelihood tree of all CytB sequences from the reference data base.

Appendix S1 Sequence count data for the 12S primer data set.

Appendix S2 Sequence count data for the CytB primer data set.

Appendix S3 Genbank accession numbers and taxon affiliations of curated 12S sequences.

Appendix S4 Genbank accession numbers and taxon affiliations of curated CytB sequences.

Appendix S5 Flow chart of steps taken during the method development and full analytical pipeline stages.

Appendix S6 Supplementary text, tables and figures.

Compilation of reference databases

In silico and in vitro testing of metabarcoding primers

Library preparation

Mock communities

Determining a threshold for defining the presence of species at individual sites

Fish abundance and distribution estimated from established method surveys

Table S1 List of species included in reference database.

Table S2 Sequences and references of primers tested in this study.

Table S3 Summary of in silico testing results.

Table S4 Total sequence reads in individual 12S mock communities.

Table S5 Total sequence reads in individual CytB mock communities.

Table S6 Summary of species abundance data from established method survey for the four cumbrian lake basins.

Table S7 Overview of families detected in the 12S and CytB data set respectively.

Fig. S1 Sampling.

Fig. S2 Maximum likelihood phylogenies used to evaluate utility of loci for species resolution in *in silico* testing.

Fig S3 Results of in vitro tests (single species amplifications) of the two chosen primer combinations a) CytB Kocher *et al.* 1989, b) 12S Kelly *et al.* 2014.

Fig. S4 Mock community results.

Fig. S5 Proportion of total false positives retained at different threshold levels for the 12S (a) and CytB (b) mock community data.

Fig. S6 Site occupancy across 29 sites in Windermere based on 12S (a) and CytB (b) using three different thresholds for defining 'presence'.

Fig. S7 Cumulative frequency distribution of relative PhiX read counts in the raw sequence data of the 12S libraries (a) and CytB libraries (b).

Fig. S8 Correlations between 12S and CytB in terms of site occupancy ('SO', a-d) and read count ('RC', e-h) data.

Fig. S9 Correlations between site occupancy and read count data per basin for 12S (a-d) and CytB (e-h)

Fig. S10 a) Read count and b) proportion of sequence reads by lake basin.

Fig. S11 Correlations between read count ('number of sequence reads') and long term rank (with '1' corresponding to the most abundant species).

Fig. S12 Correlations between site occupancy and 2014 gill net numbers.

Fig. S13 Correlations between read count ('number of sequence reads') and 2014 gill net numbers.

Fig. S14 Distribution of eDNA sequence reads in each of the depth profile transects.

Fig. S15 Spatial distribution of species recorded in Windermere North and South Basins at more than 2 sites (site occupancy) using 12S data.