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Non-specific amplification compromises environmental DNA metabarcoding with COI

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ABSTRACT

1. Metabarcoding extra-organismal DNA from environmental samples is now a key technique in aquatic biomonitoring and ecosystem health assessment. Of critical consideration when designing experiments, and especially so when developing community standards and legislative frameworks, is the choice of genetic marker and primer set. Mitochondrial cytochrome *c* oxidase subunit I (COI), the standard DNA barcode marker for animals, with its extensive reference library, taxonomic discriminatory power, and predictable sequence variation, is the natural choice for many metabarcoding applications. However, for targeting specific taxonomic groups in environmental samples, the utility of COI has yet to be fully scrutinised.

2. Here, by using a case study of marine and freshwater fishes from the British Isles, we quantify the *in silico* performance of twelve primer pairs from four mitochondrial loci—COI, cytochrome *b*, 12S and 16S—in terms of reference library coverage, taxonomic discriminatory power and primer universality. We subsequently test *in vitro* four primer pairs—three COI and one 12S—for their specificity, reproducibility, and congruence with independent datasets derived from traditional survey methods at five estuarine and coastal sites around the English Channel and North Sea.

3. Our results show that for aqueous extra-organismal DNA at low template concentrations, both metazoan-targeted and fish-targeted COI primers perform poorly in comparison to 12S, exhibiting low levels of reproducibility due to non-specific amplification of prokaryotic and non-target eukaryotic DNAs.

4. An ideal metabarcode would have an extensive reference library upon which custom primers could be designed, either for broad assessments of biodiversity, or taxon specific surveys. Such a database is available for COI, but low primer specificity hinders practical application, while conversely, 12S primers offer high specificity, but lack adequate references. The latter, however, can be mitigated by expanding the concept of DNA barcodes to include whole mitochondrial genomes generated by genome-skimming existing tissue collections.

[Keywords: 12S, COI, eDNA, Environmental DNA, metabarcoding, primer design.]

INTRODUCTION

DNA barcoding and metabarcoding techniques are now established and indispensable tools for the assessment and monitoring of past and present ecosystems (Leray and Knowlton, 2015; Pedersen et al., 2015; Thomsen and Willerslev, 2015; Valentini et al., 2016), and are being increasingly incorporated into policy and management decisions (Kelly et al., 2014b; Rees et al., 2014; Mariani et al., 2015; Hering et al., 2018). A remarkably wide range of biological substrates can now be sequenced to identify presence of a particular species or reconstruct communities, and can include restaurant sushi meals (Vandamme et al., 2016), deep sea sediments (Guardiola et al., 2015), permafrost ice cores (Willerslev et al., 2003), terrestrial insect collections (Ji et al., 2013), animal faeces (Kartzinel et al., 2015) and seawater samples (Thomsen et al., 2012a).

The term "DNA metabarcoding" encompasses two distinct methodologies: (i) bulk sample metabarcoding, which is the direct amplification of a concentrated mixture of organisms, from for example, plankton (Clarke et al., 2017), arthropods (Yu et al., 2012) or gut material (Leray et al., 2013); or (ii) environmental DNA (eDNA) metabarcoding, which is indirect amplification via extra-organismal DNA in water, sediments, or soils (Taberlet et al., 2012). This latter methodology involves first isolating and concentrating DNA using filters, rather than homogenising entire organisms or parts of organisms (Yu et al., 2012; Spens et al., 2017; Macher et al., 2018). The detection of macrobial fauna such as vertebrates and insects using aquatic eDNA has been recognised as a highly sensitive survey technique and a key use-case of metabarcoding (Rees et al., 2014; Valentini et al., 2016). However, DNA from environmental samples such as seawater is likely to be degraded (Collins et al., 2018), and also have a significant quantity of co-extracted microbial DNA that may co-amplify with the targeted metazoan DNA molecules (Stat et al., 2017; Andújar et al., 2018).

Early eDNA metabarcoding studies targeting fishes used the cytochrome *b* gene (Minamoto et al., 2012; Thomsen et al., 2012b,a), but more recent studies have used the 12S ribosomal rRNA locus (Kelly et al., 2014a; Hänfling et al., 2016; Port et al., 2016; Stoeckle et al., 2017; Yamamoto et al., 2017; Ushio et al., 2018), and also 16S rRNA (Shaw et al., 2016; Berry et al., 2017; Bylemans et al., 2018; Jeunen et al., 2018; Stat et al., 2018). Various regions of 12S have been proposed as metabarcoding markers, including a ca. 63 bp fragment (Valentini et al., 2016), a ca. 106 bp fragment (Riaz et al., 2011; Kelly et al., 2014a), and a ca. 171 bp fragment (Miya et al., 2015). Modified versions of some of these primers have been published by Taberlet et al. (2018). Ribosomal genes such as 12S and 16S offer the advantage of conserved priming sites (Deagle et al., 2014; Valentini et al., 2016), and amplification across a broad range of fish taxa (Miya et al., 2015; Bylemans et al., 2018). However, taxonomic resolution can be low (Miya et al., 2015; Hänfling et al., 2016; Andruszkiewicz et al., 2017), with relatively short length ribosomal markers being unable to distinguish commercially important species of the cod family Gadidae, for example (Thomsen et al., 2016). A problem for studies using ribosomal markers are the reference libraries, which are usually poorly populated, and often have to be developed for each project on an *ad hoc* basis (Miya et al., 2015; Thomsen et al., 2016; Stoeckle et al., 2017). Assembling reference libraries for ribosomal genes is further complicated by frequently-used primer sets amplifying different regions, with any two given 12S references taken from GenBank, for example, not necessarily being homologous.

For animals, the primary DNA barcode is the 5′ "Folmer" region of COI, the cytochrome *c* oxidase subunit I gene (Folmer et al., 1994; Hebert et al., 2003). In comparison to ribosomal markers, the advantages of COI are high interspecific variability (Ward, 2009), an extensive reference database (BOLD; Barcode of Life

Database; Ratnasingham and Hebert, 2007), and due to the protein-coding constraints of the gene, more straightforward bioinformatic procedures such as alignment and denoising (Andújar et al., 2018). Inside the 5′ Folmer fragment, multiple primer sets have been developed, targeting shorter regions in the 100–400 bp range. These are more suitable than a full length barcode (ca. 658 bp) for analyses of degraded DNA, or for sequencing on short read platforms such as Illumina (Leray et al., 2013; Shokralla et al., 2015; Elbrecht and Leese, 2017). However, due to its variability, finding conserved priming regions within the Folmer fragment is difficult, and concerns have been raised about the suitability of some COI primers in terms of species-specific primer-template mismatches, which can result in inefficient, biased amplifications that may hinder quantitative analyses (Deagle et al., 2014). Addressing this problem with bias requires incorporating a high degree of degeneracy into COI primers (Leray et al., 2013; Marquina et al., 2019), particularly by the use of multiple inosine sites (Shokralla et al., 2015; Elbrecht and Leese, 2017; Wangensteen et al., 2018). Despite these issues, Andújar et al. (2018) argue that COI should be the standard marker for metabarcoding, and COI markers are increasingly being used for eDNA metabarcoding (Bakker et al., 2017; Kelly et al., 2017; Stat et al., 2017; Jeunen et al., 2018; Macher et al., 2018; Singer et al., 2019). However, studies comparing the efficacy of different primer sets have done so in a bulk-sample metabarcoding context (Clarke et al., 2017; Elbrecht and Leese, 2017), or have compared only ribosomal markers for vertebrate eDNA applications (Bylemans et al., 2018). Therefore, there lacks a clear assessment of how degenerate COI primers compare to 12S and 16S rRNA when used on low-template-concentration environmental samples where non-target DNA molecules are found in abundance.

Given the importance of marker choice in metabarcoding studies (Alberdi et al., 2018), and the need to thoroughly scrutinise the utility of COI in comparison with the widely used ribosomal markers (Deagle et al., 2014; Andújar et al., 2018), we use a case study of fishes from the British Isles—a well studied and important group in terms of ecosystem health and human food security—to ask whether COI primer sets can be used for eDNA metabarcoding of aquatic vertebrates, and how they compare to alternative 12S, 16S and cytochrome *b* markers. We survey a range of published primer sets both *in silico* and *in vitro*, including metazoan-targeted COI primers with high levels of degeneracy, and novel fish-targeted COI primers with reduced degeneracy. Using *in silico* methods we assess a number of factors: (i) the reference database coverage for the individual fragments, i.e. how many species and individuals of each species are represented in public databases; (ii) the taxonomic discrimination of each fragment, i.e. is each unique DNA sequence unambiguously associated with a single species name; and (iii) the universality of the primer set, i.e. are all species of the target taxonomic group predicted to amplify equally well. Then, we test using a series of water samples taken from locations with corresponding data from traditional fish survey methods, three COI primer sets against a best performing alternative set, as based upon the results of the *in silico* analyses. By PCR amplifying and sequencing these water samples we compare: (i) the specificity of the primer set, i.e. the proportion of the reads that came from the target taxonomic group; (ii) the power of the primer set, i.e. the total species richness estimated; (iii) the reproducibility of the primer set, i.e. are the same species consistently represented in replicate water samples and PCRs; and (iv) the congruence of the primer set, i.e. are the same species detected in the traditional surveys as the eDNA surveys.

METHODS

In silico analyses

Reference library construction

A list of fish species recorded from the marine and freshwater environments of the British Isles was compiled from three sources: (i) the Global Biodiversity Information Facility (https://www.gbif.org; *rgbif v1.1.0*; Chamberlain and Boettiger, 2017); (ii) FishBase (https://www.fishbase.org); and (iii) the European Water Framework Directive United Kingdom Technical Advisory Group list of transitional fish species (https://www.wfduk.org/resources/transitional-waters-fish; Annex 1). These species were then cross-referenced for all synonyms using *rfishbase v3.0.0* (Boettiger et al., 2012). The subsequent list of valid species names and all their synonyms was then searched using *rentrez v1.2.1* (Winter, 2017) against NCBI GenBank release 230 (nucleotide database; https://www.ncbi.nlm.nih.gov/nucleotide/) for any of the following terms: "COI, 12S, 16S, rRNA, ribosomal, cytb, CO1, cox1, cytochrome, subunit, COB, CYB, mitochondrial, mitochondrion". The Barcode of Life Database BOLD (http://www.boldsystems.org/) was also searched for the same species using *bold v0.8.6* (Chamberlain, 2018).

Hidden Markov models of the alignments of each primer set were then constructed using *HMMER v3.1b2* (http://hmmer.org/; Eddy, 1998) and the fish mitochondrial genome database (http://mitofish.aori.u-tokyo.ac.jp/; Iwasaki et al., 2013). These profiles were used to extract homologous regions of nucleotides from the total mitochondrial data obtained from the GenBank and BOLD searches. The resulting sequences were then annotated with metadata using *traits v0.4.2* (Chamberlain et al., 2019). A phylogenetic quality control step was then carried out by aligning the sequences in *MAFFT v7.271* (Katoh and Standley, 2013) and constructing a maximum likelihood tree using *RAxML v8.2.12* (Stamatakis et al., 2008). Sequences with putatively spurious annotations—i.e. those indicative of misidentifications—were filtered out if the following criteria were met: (i) individual(s) of species *x* being identical to or nested within a cluster of sequences of species *y*, but with other individuals of species *x* forming an independent cluster; and (ii) the putatively spurious sequences coming from a single study, while the putatively correct sequences of species *x* and *y* coming from multiple studies. Records flagged by NCBI as "unverified" were also omitted.

Primer design

We designed two new COI metabarcoding primers targeting fishes (Table 1): "SeaDNA-short" and "SeaDNA-mid", which share a forward primer, and are internal to the Folmer fragment. The new primer pairs were designed manually in *Geneious v8.8.1* (Kearse et al., 2012) using the same fish mitochondrial genome dataset as described above, with the assistance of *Primer3* (Untergasser et al., 2012) and the sliding window functions in *spider v1.3.0* (Boyer et al., 2012; Brown et al., 2012). The primers were tested on a range of fish tissue extractions from elasmobranchs and actinopterygians, and produced strong clean PCR amplicons of the expected size.

In silico *PCR and taxonomic discrimination*

Primers were evaluated using a subset of 955 unique sequences from 184 species obtained in the British Isles fish reference library construction step, for which full mitochondrial genomes were available. Twelve primer pairs were chosen for the *in silico* PCRs, representing COI, cytochrome *b*, ribosomal 12S and ribosomal 16S (Table 1). *MFEprimer v2.0* (Qu et al., 2012) was used to perform the *in silico* PCR on the untagged primers.

Amplification universality was estimated using the Primer Pair Coverage (PPC) statistic from *MFEprimer*, where $PPC = \frac{Fm}{Fl} \times \frac{Rm}{Rl} \times (1 - CVfr)$, with Fl and Rl the length of the forward and reverse primers, and $CVfr$ the coefficient of variability of matched lengths Fm and Rm to the template. Therefore, a PPC value of 100% indicates complete binding of both primers to a template. The highest PPC value was then selected for each species, and averaged over all species to provide the PPC for each primer set. Predicted non-amplifications with a default 5 bp 3′ binding stability of $> 0\Delta G$ were set to a PPC of 0%. In order for sufficient RAM to be available to complete the analysis of the highly degenerate Leray-XT primer set, the inosine sites were simplified to double-base ambiguities. This was achieved by choosing the most frequent base combination in the mitogenome alignment. None of the altered inosine sites were within 8 bp of the 3′ end of the primer (Table 1).

Taxonomic discrimination (= resolution) was assessed first using all available species from the British Isles fish reference library for each primer set individually, and then secondly on a subset of species for which sequences were present for all of the primer sets. Discrimination as a proportion of the total number of species was calculated following Ficetola et al. (2010): "A taxon unambiguously identified by a primer pair owns a barcode sequence associated to this pair that is not shared by any other taxa".

Primer evaluation *in vitro*

Field sites and traditional fish survey

Five locations in the United Kingdom were surveyed for fishes using eDNA and traditional methods between October and November of 2016. These included: the River Tees, County Durham (54.631327,-1.164447); two sites within the River Esk estuary, North Yorkshire (54.491633,-0.611833; 54.48975,-0.612617); the River Test, Hampshire (50.901563,-1.440836); and Whitsand Bay, Devon (50.329616,-4.243751). The former four are estuarine sites, while the latter is an inshore coastal area, approximately 1 km from shore. Fish sampling in the River Esk estuary was done by duplicate fyke nets (Esk-fyke) and duplicate beach-seine nets (Esk-seine), in different locations. At the River Tees sampling site, duplicate beach-seine netting and two shallow beam trawls were carried out. The River Test site comprised a 24 h fish impingement survey conducted at Marchwood Power Station. Whitsand Bay was surveyed by four otter trawls, as described in McHugh et al. (2011). The variety of fishing techniques used in the different sampling locations are part of the currently ongoing fish monitoring programmes implemented by local collaborating organisations (Environment Agency, PISCES Conservation Ltd., Marine Biological Association). Further details are presented in Supporting Information.

Water processing and DNA extraction

Three 2 L water sample replicates per site were collected immediately prior to the traditional fish survey commencing, using Nalgene HDPE collection bottles pre-sterilised with a 10% bleach solution. Water was pre-strained with a 250 μm nylon mesh filter to remove debris, if required. After collection, the water samples were put into individual sterile plastic bags, and stored in an ice box while being transported back to the laboratory. Within five hours, each 2 L sample was filtered through an 0.22 μm Sterivex-GP PES filter (Merck Millipore) using a 100 mL polypropylene syringe or a peristaltic pump, and cleared of water. When the full 2 L could not be passed due to filter clogging, the volume of water was recorded. After filtration, the filters were stored at −20°C. DNA was extracted from the filters using the DNeasy PowerSoil DNA Isolation Kit

(MoBio/Qiagen), following the manufacturers' protocol, with the addition of an initial 2 h agitation step to promote the release of DNA from the filter, during which the filter membranes were placed in tubes with lysis buffer C1 and garnet beads from the PowerWater Isolation kit and shaken at 65°C. Filtration blank controls were processed in parallel. All processing was carried out in dedicated eDNA extraction laboratories, and equipment and surfaces were regularly cleaned using a 10% bleach solution. The eDNA extraction, pre-PCR preparations and post-PCR procedures were carried out in separate rooms.

PCR and library preparation

Four primer sets were selected to go forward for *in vitro* testing: three COI primer sets (Leray-XT, SeaDNA-short, SeaDNA-mid), and one best-performing primer set from the *in silico* analysis (12S MiFish-U). All PCR amplifications were done in duplicate reactions each with a unique 7/8-mer oligo-tag barcode, differing by at least three bases (Guardiola et al., 2015). In order to increase variability of the amplicon sequences, a variable number (two, three or four) of fully degenerate positions (Ns) were added at the 5′ end of the oligo tags (Wangensteen et al., 2018). For PCR amplification with the newly designed SeaDNA-short and SeaDNA-mid primers, a two-step protocol was used, first using untagged primers, then tagged primers in a second PCR round. The reaction for the first PCR step included 10 μL AmpliTaq Gold 360 Master Mix (Thermofisher), with 1 μL of each 5 μM forward and reverse primer, 0.16 μL of bovine serum albumin and 10 ng of purified DNA in a total volume of 20 μL per sample. Thermocycling profile for the first step included an initial denaturation at 95°C for 10 minutes, then 40 cycles of 94°C for 30 sec, 47°C for 45 sec and 72°C for 30 sec, and then a final extension of 72°C for 5 minutes. The profile for the second PCR step was identical, except for the annealing temperature being 50°C instead of 47°C. Amplifications were assessed by electrophoresis on a 1.5% agarose gel, and the field and laboratory controls were checked for the presence of amplicons. Between the first and second PCR step, amplicons were purified using MinElute PCR purification columns (QIAGEN) and diluted by a factor of ten prior to being used as a template for the second PCR. After the second PCR, all tagged amplicons were pooled by marker, purified again using MinElute columns and eluted into a total volume of 45 μL, in order to concentrate the amplicons approximately 15 times. For 12S MiFish and Leray-XT we used a one-step procedure with tagged PCR primers, with PCR cycling conditions following Miya et al. (2015) and Wangensteen et al. (2018), respectively. Reagents and volumes were the same as for the two-step protocol.

Libraries (one for each primer set) were built using the PCR-free NEXTflex library preparation kit (BIOO Scientific). The libraries were quantified using the NEBNext qPCR quantification kit (New England Biolabs) and spiked with with 1% PhiX (Illumina). The libraries were sequenced on an Illumina MiSeq platform, using V3 chemistry (2×75 bp paired-end) for the SeaDNA-short library, which was run along with two other libraries from unrelated projects. For the MiFish-U and SeaDNA-mid libraries, V2 chemistry (2×150 bp paired-end) was used, and these were sequenced in the same run. The Leray-XT library was run using V2 chemistry (2×250 bp paired-end) along with another library from an unrelated project.

Bioinformatic processing

Raw sequencing data were converted to fastq format using *bcl2fastq v2.20* (https://support.illumina.com/sequencing/sequenc conversion-software.html). The remaining bioinformatic steps were carried out using *cutadapt v2.3* (Martin, 2011) and *dada2 v1.10.1* (Callahan et al., 2016). Because a PCR-free library preparation kit was used,

adapters could have been ligated to either the 5′ or the 3′ end of the amplicon, and in order to take advantage of the Illumina error profiling in the *dada2* denoising step, the sense- and antisense-orientated sequences were first isolated and processed independently. This was achieved using *cutadapt* by filtering the R1 fastq files for reads with the forward PCR primer, and then for those with the reverse PCR primer. The reads were then demultiplexed by tag, followed by primer and adapter trimming. Quality trimming was carried out in *dada2* using default settings, but with read truncation length "truncLen" determined to give an approximate 30 bp overlap between forward and reverse reads. The reads were then denoised, dereplicated, merged, cleaned of chimaeras and reorientated, using the *dada2* workflow. Our reference library sequences for each primer set were used as priors to avoid low abundance but valid sequences being discarded during denoising. A homology filter was then implemented by aligning the ASVs against a hidden Markov model of the expected fragment using *HMMER hmmsearch*, and the non-homologous reads discarded.

Taxonomy assignment of the amplicon sequence variants (ASVs) produced by *dada2* was carried out using a multi-step procedure, incorporating distance-based and phylogenetic methods. First, a preformatted "nt" blast database was downloaded from NCBI (ftp://ftp.ncbi.nlm.nih.gov/blast/db/v5; 21 March 2019). Each ASV sequence was then locally blasted against this database using *blastn v2.9.0* ('-task blastn -evalue 1000 -word_size 11 -max_target_seqs 500'), and the results filtered to obtain a rough taxonomic classification based on the best-scoring blast hit. Next, a more stringent procedure was carried out, with the putative fish sequences extracted from this initial blast result subjected to a second *blastn* search, this time using our curated reference library of British Isles fishes as the blast database (same settings as the "nt" search but with '-word_size 7'). The same reads were then run through the Evolutionary Placement Algorithm (*EPA-ng v0.3.5*, *gappa v0.2.0;* Barbera et al., 2018; Czech and Stamatakis, 2018). Species name(s) were assigned based on either of the following rules: (i) species-level EPA placement same as the best scoring blast hit, with an aligned match length of $\geq 90\%$ of the modal length of the fragment, and an identity of $\geq 97\%$; or (ii) highest likelihood EPA placement same as the best scoring blast hit, with an EPA probability $\geq 90\%$ and blast identity $\geq 90\%$. Rule (i) finds assignments that are congruent between both the EPA and blast methods, but rejects assignments with low similarity and short match lengths. Rule (ii) allows for dissimilar hits, but only ones that have a high phylogenetic probability, and which are usually indicative of low abundance variants with errors. Our prior knowledge of the expected fish fauna of the sites was used to set these cut-off values, with the aim of conservatively minimising false positive assignments. The fish reads were also summarised by OTU clustering using *Swarm v2.2.2* (Mahé et al., 2015), with $d = 1$ and the "fastidious" option enabled. This step permitted an evaluation of possible misassigned and unassigned species.

RESULTS

In silico analyses

A total of 531 species were identified as part of the United Kingdom marine and freshwater fish fauna. Of these, 176 names were flagged as "common" species, having been identified as relatively widespread marine or freshwater taxa that are likely to be encountered during survey work of coastal and inland habitats (Kottelat and Freyhof, 2007; Henderson, 2014). The remainder were mostly highly localised species, deep water offshore species, or rare migrants. The combined reference library for all primer sets, after cleaning, duplicate removal and quality control, comprised 43,366 sequences from 491 total species, and 25,799 sequences from

172 common species.

In terms of reference database coverage for individual primer sets (Table 2), COI primers had the greatest number of reference sequences at 23,911–24,058, covering 91% of species. The "Minamoto-fish" cytochrome *b* set had 15,405 sequences and a species coverage of 65%. Of the ribosomal primer sets, the "Berry-fish" 16S set had the greatest number of sequences at 4,089, with species coverage at 77%. Among the 12S sets, the "Riaz-V5" primers had the greatest number of sequences (2,416; species coverage 69%), while the "Valentini-tele01" set had the fewest sequences (1,699; species coverage 51%). The "MiFish" primers and their variants (MiFish-U/E, Taberlet-tele02, Taberlet-elas02) had 1,904 sequences, and a coverage of 61%. Per species, the average number of reference sequences was greatest for the COI primer sets (mean 49–50; median 24), followed by cytochrome *b* (mean 45; median 7), 16S (mean 9.9; median 4), and then 12S (mean 5.9–6.6; median 2–3). When only the subset of common species was considered, the species coverage increased for all primer sets, as did the average number of sequences per species (Table 2).

In terms of taxonomic discrimination of the fragments obtained from each primer set (Table 2), the proportion of British Isles fish species where all individuals could be unambiguously identified was greatest for the Leray-XT COI fragment at 95%, while the shorter SeaDNA-mid and SeaDNA-short COI fragments resolved 91% and 87% respectively. The cytochrome *b* fragment discriminated 91%. The MiFish fragment had the greatest discrimination among the ribosomal primer sets at 93%, with the Berry-fish 16S, Valentini-tele01, and Riaz-V5 pairs having lower rates (89%, 86%, and 79% respectively). When a standardised dataset of species common to all primer sets ($n = 88$) was used, the overall pattern remained similar (Table 2).

In terms of primer universality as estimated by *in silico* PCR for British Isles fish species with comparable data available for all markers ($n = 184$; Table 2), the 12S primer sets targeting actinopterygians had a higher mean PPC than all other markers, at between 77.1% (Valentini-tele01) and 92.2% (Riaz-V5), compared to between 19.1% (cytochrome *b*) and 50.9% (16S). The best performing COI marker for actinopterygians (SeaDNA-short) had a PPC value of 34.5%. For elasmobranchs, three 12S primer pairs had the highest mean PPC values, with Taberlet-elas02 at 83.6%, Valentini-tele01 at 68.2%, and MiFish-E at 55%. The 12S Riaz-V5 primers, the cytochrome *b* primers, and the 16S primers, had the lowest PPC values (11.2%, 20.4% and 0% respectively), while the COI primers had PPC values between 21.5% (SeaDNA-short) and 39% (simplified Leray-XT). These patterns remained when only common species were compared (Table 2).

In vitro analyses

Total reads from Illumina sequencing (Table 3) varied between 3.4 million (12S MiFish-U) and 14.3 million (COI SeaDNA-mid). After bioinformatic processing, the proportions of reads retained were 46% (COI SeaDNA-short), 54% (COI Leray-XT), 61% (COI SeaDNA-mid) and 63% (12S MiFish-U). Mean cleaned reads recovered per sampling event (triplicate water samples, duplicate PCR tags; $n = 6$) were: 107,458 (SD = 46,924) for Leray-XT; 290,104 (SD = 118,592) for SeaDNA-mid; 135,804 (SD = 44,993) for SeaDNA-short; and 71,912 (SD = 13,682) for 12S MiFish-U. Supporting Figure S1 shows distributions of read depths per sample for each site and primer set. The 12S MiFish-U primers provided the greatest proportion of chordate and fish reads (100% and 76% of cleaned reads, respectively), resulting in more than 1.6 million putative fish reads and 156 fish ASVs. From these fish reads, 96% were assigned to 41 species and 67 *Swarm* OTU clusters. A total of 73,377 fish reads comprising 18 *Swarm* OTUs could not be assigned, and in addition to PCR and sequencing artefacts, these likely represent at least eight species not present in the reference library

(Supporting Table S1). For the COI primer sets, chordate reads comprised between 0.2% (Leray-XT) and 6% (SeaDNA-short) of the total cleaned reads, with between 0.1% and 5% putative fish reads comprising between 22 (Leray-XT) and 29 (SeaDNA-short) assigned species. Between 42% (Leray-XT) and 85% (SeaDNA-short) of the putative fish reads were unassigned to species. The non-chordate reads were inferred from the preliminary blast search to consist of DNA from other metazoans (4–10%) and eukaryotes (41–83%), or bacteria (17–59%).

Per sampling location the 12S MiFish-U primer set detected a consistently greater number of total species across sites than the COI markers, at between 2.2 (River Test) and 2.6 (Whitsand Bay) fold higher (Figure 1). The SeaDNA-short primers detected a greater number of species than both the SeaDNA-mid and Leray-XT primers, except at the River Tees site where SeaDNA-mid detected one more.

In terms of reproducibility (Figure 2), the 12S MiFish-U primer set showed a greater proportion of shared species—the top ten species by read abundance at each location—amplified across water sample and PCR replicates, with a 71% mean reproducibility over all sampling locations. The COI primer sets had mean reproducibility values of 36% (SeaDNA-short), 29% (SeaDNA-mid) and 12% (Leray-XT).

When compared to traditional survey methods—with the freshwater species omitted from the eDNA results as they were not expected to be found on the traditional fish surveys of the estuarine and coastal habitats—the 12S MiFish-U primer set showed the greatest congruence (Figure 3), at between 15% (Whitsand Bay) and 54% (River Test). The COI primers were between 9% (Leray-XT) and 13% (SeaDNA-short) congruent overall. The MiFish-U primer set also amplified a greater number of marine/estuarine species to the traditional survey methods at all locations except for Whitsand Bay (26 versus 23 species). The COI primer sets amplified fewer marine/estuarine species than the traditional surveys in all cases, except for the SeaDNA-short primer set at the River Tees and River Esk sites. For each site survey, reads per species (eDNA survey) and individuals per species (traditional survey) are presented in Supporting Tables S2–S6.

DISCUSSION

A single metabarcoding marker for fishes?

Of arguably the greatest importance in the ability of metabarcoding to answer a particular question, is that of the choice of marker and primer (Deagle et al., 2014; Valentini et al., 2016; Clarke et al., 2017; Elbrecht and Leese, 2017; Alberdi et al., 2018). The ideal genetic marker for eDNA metabarcoding marker should be flexible, allowing different primer sets to target different taxonomic groups, but requiring only a single reference library. Each individual primer set must also be designed with the following qualities: (i) it must be universal, i.e. amplifying a large proportion of the target taxonomic group; (ii) it must be specific, i.e. it must not amplify other taxa at the expense of the target group; (iii) it must be unbiased, i.e. not preferentially amplifying a subset of the target group; (iv) it must be discriminatory, i.e. the DNA fragment recovered should differentiate at the appropriate taxonomic level for the question; and (v) it must be replete, i.e. associated with a reference library enabling identifications within the target taxonomic group. Here, we assess these characteristics for COI, cytochrome *b*, 12S, and 16S primer sets using the example of marine and freshwater fishes from the British Isles.

Which primers have the best reference library?

In terms of reference libraries, the COI primers were substantially better endowed than all other marker genes, with between 1.6 times (cytochrome *b*) and 14 times (Valentini-tele01) more public sequence data available for all species. This was also reflected in the common species coverage, at up to 97% for COI. The 16S (95%), cytochrome *b* (81%), and 12S Riaz-V5 libraries (81%) were also well developed for common species, but coverage for other 12S primer sets was lower, at 56–62%. A reference library with broad taxonomic depth will allow inferences beyond a comparison of anonymous MOTUs, thereby leveraging the wealth of scientific information that a taxonomic name brings with it (Ward et al., 2009). Deep coverage in the COI reference library—i.e. the number and geographic distribution of sequences per species—also has advantages in terms of potential for population level assignments, and for flagging spuriously identified sequences; due to the lesser weight of evidence from the low numbers of sequences, misidentifications were harder to confirm for 12S during the quality control step. Furthermore, in terms of voucher specimen and location data etc, much of the ribosomal data on GenBank are not validated to the same standard as COI data on BOLD are (Ward et al., 2009). However, it is important to remember that despite the success of 15 years of the DNA barcode initiative producing COI coverage spanning the majority of northern European fish species, the BOLD database still remains seriously underdeveloped for many other taxonomic groups such as marine invertebrates (Bucklin et al., 2011; Leray and Knowlton, 2016).

Which primers best discriminate species?

In terms of the discriminatory power for our dataset of British Isles fish species, all primer sets gave a resolution above 90% except for SeaDNA-short (COI), Valentini-tele01 (12S), Riaz-V5 (12S) and Berry-fish (16S). Predictably, the longer COI fragments resolved more species than the shorter ones, at 95% for the 313 bp Leray-XT and 87% for the 55 bp SeaDNA-short fragment. The 12S primers did not show this pattern as clearly, with the shorter Valentini-tele01 fragment having a better taxonomic resolution (86%) than the longer Riaz-V5 fragment (79%); the longest, MiFish-U/E and Taberlet-tele02/elas02 primers, had the greatest species resolution at 93%. While discriminatory power may depend on the range of species in that particular library, the observed patterns held up when a dataset of sequences that were shared for all primer sets was used. Discriminatory power also tended to remain the same or increase when only the common species were considered, most likely because rare but genetically similar congeners were excluded.

Which primers are most universal?

Primer universality as estimated by *in silico* PCR varied greatly. Our results show that the metabarcode primers targeting protein-coding genes—COI and cytochrome *b*—are likely to exhibit a greater degree of species-level primer bias (i.e. lower universality) than ribosomal 12S and 16S, as indicated by the lower mean PPC values; a mean PPC of 96% was estimated for common actinopterygian species amplified with the Riaz-V5 and Taberlet-tele02 primers. Previous studies have also reported or predicted less primer bias with rRNA targets than protein coding ones (Clarke et al., 2014; Deagle et al., 2014; Elbrecht et al., 2016; Marquina et al., 2019). It is also important to note again that due to the high level of degeneracy the Leray-XT primers were simplified to overcome RAM limitations of the analysis, and therefore the value presented is likely to be an underestimate of their true potential, as highly degenerate COI primers have been shown to reduce bias substantially (Marquina et al., 2019).

Regarding higher level taxonomic bias, for the 12S and 16S primers tested here, no set except Valentini-tele01 appeared suited to amplify actinopterygians and elasmobranchs equally. The COI primers were, however, relatively unbiased in regard to higher taxonomic group. The MiFish primers and the Taberlet et al. (2018) variants of the same sets were both published with actinopterygian (MiFish-U) and elasmobranch (MiFish-E) versions, due to a number of mismatches in the conserved regions (Miya et al., 2015). Unsurprisingly, both of these performed substantially better for their respective taxa. The Taberlet et al. (2018) primers were also predicted here to exhibit reduced species-level primer bias compared to the original MiFish versions, for both elasmobranchs and actinopterygians.

Many studies computationally predict primer amplification by the number of mismatches between primer and template (e.g. Riaz et al., 2011), or by the number of mismatches and their type and position (e.g. Elbrecht et al., 2017), but often do not fully consider the thermodynamics of a primer-template reaction. We used the thermodynamics-based PCR simulation implemented in MFEprimer (Qu et al., 2012), but regardless of whether this method is more realistic or accurate than alternative methods, it is important to remember that these are predicted amplifications, and were used here to compare relative performances between primer sets. Therefore, the lower values estimated do not represent amplification failure *per se*, but rather are indicative of increased bias associated with that primer set (Deagle et al., 2014). For example, the standard COI DNA barcode primers for fishes (Ward-barcode) had a very low PPC, but these are tried-and-tested primers for amplifying a wide range of fish taxa in standard PCR for Sanger sequencing (Ward et al., 2005). The use of mock communities is an important step in quality controlling an assay if primer bias is suspected (Piñol et al., 2015; Elbrecht and Leese, 2017; Bista et al., 2018), but *in silico* PCR has been demonstrated to be an effective proxy in its absence (Clarke et al., 2014).

We used the results of our *in silico* analyses to inform our choices for the *in vitro* experiments. All COI primer sets were selected for testing *in vitro* because of the advantages in terms of reference library and taxonomic discrimination. We chose only one 12S set for comparison, and here we chose the MiFish-U primer pair because this pair had better predicted universality for actinopterygians and more reference sequences available than the Valentini-tele01 primers, and greater taxonomic discrimination than the Riaz-V5 primers. Due to the better predicted universality of the Taberlet-tele02 primer set compared to MiFish-U, these would have been chosen had they been publicly available at the time the experiment was implemented. Despite the well developed reference libraries and good taxonomic discrimination, we did not select cytochrome *b* or 16S because of the lower predicted universality of these primers in comparison to 12S.

Which primers are the most specific?

Despite having the fewest total raw reads, the MiFish-U primer set produced the greatest number and proportion of usable fish reads (76% of processed reads, 48% of raw reads), the greatest overall species richness (41 species), and the greatest proportion of fish reads that were assigned to species (96%). The COI primers amplified a very low proportion of chordate and fish reads compared to the overall sequencing depth (maximum 5% of cleaned reads were fishes). The majority of the SeaDNA-short and SeaDNA-mid reads were estimated by preliminary blast search to have come from bacteria or non-metazoan eukaryotes (86–90%).

That the highly degenerate Leray-XT primers produced a low proportion of fish reads is unsurprising given that previous studies on environmental samples using degenerate COI primers have demonstrated that they can amplify widely beyond their target taxa, and can produce large proportions of unassigned reads (Lim

et al., 2016; Stat et al., 2017; Macher et al., 2018; Singer et al., 2019). The proportion of bacterial reads are generally lower when metabarcoding bulk organismal samples, however, with most reads belonging to metazoans (Leray and Knowlton, 2015; Macher et al., 2018; Wangensteen et al., 2018). More surprising was the poor specificity of the SeaDNA-short and SeaDNA-mid primers, which were designed to target fishes, and with minimal degeneracy. These data are, however, consistent with those of an analysis of shark diversity by Bakker et al. (2017), who used COI mini-barcode primers designed on sharks, and reported a similar level of non-specific amplification.

The cause of this non-specific amplification is likely to be the extensive homoplasy (nucleotide convergence) apparent in the mutationally saturated COI gene and its homologs. Siddall et al. (2009) demonstrated that metazoan-targeted COI primers are likely to co-amplify many marine prokaryote groups—gammaproteobacteria being a particularly diverse and abundant lineage (Sunagawa et al., 2015)—thereby compromising the specificity of these primer sets. Optimisation of PCR protocols or library preparation methods may increase specificity of the assay (Siddall et al., 2009), but it is probably unlikely that it can increase to a level that makes the proportion of usable reads viable for eDNA metabarcoding of targeted taxonomic groups. While this phenomenon was first observed in marine prokaryotes, studies on freshwater and soil faunas have shown a similar pattern, also with large numbers of unassigned reads (Yang et al., 2014; Lim et al., 2016).

Which primers give the most reproducible results?

The low number of usable fish reads for the COI primers is reflected in the reproducibility of the assays across water sample and PCR replicates. For the most frequently amplified species at each site, the COI primers were less consistent than 12S MiFish-U overall. Low quantities of template DNA and stochasticity in early PCR cycles is a known factor in causing poor reproducibility (Leray and Knowlton, 2017; Alberdi et al., 2018; Collins et al., 2018), and can be ameliorated by performing multiple PCR technical replicates (Ficetola et al., 2015). We show that this effect is exacerbated when primer specificity is low and non-target organisms are abundant, as is the case in highly diverse environmental samples such as seawater. For many applications repeatability between assays or sampling sites is a requirement, such as the detection of an endangered or invasive species (Grey et al., 2018). Our results, even considering only the top ten common species, show that detectability can vary between sites with the same genetic marker, and that many more than two PCRs will be required if the rare species are to be detected across multiple PCR and water sample replicates (Dopheide et al., 2018).

Species richness estimates at all sampling sites were greatest with 12S MiFish-U, and this was despite the deficiencies in the reference library, at only 61% species coverage. For example, species including the European plaice (*Pleuronectes platessa*) and European flounder (*Platichthys flesus*)—both common fishes present at all sampling locations—were missing from the reference library and therefore not represented when comparing with the traditional fish surveys. Most of the large number of reads that were assigned to American plaice, *Hippoglossoides platessoides* ($n = 198,445$), were likely misassigned and actually belong to European plaice and flounder (Supporting Table S1). The *Swarm* OTU analysis showed a greater number of clusters (67) than assigned species (41), also suggesting that some species missing from the reference library are likely to have been misassigned. While a small number of the 73,377 unassigned 12S fish reads were low abundance sequences derived from artefacts, almost all could be could be inferred by phylogenetic analysis

or by similarity to geographically disjunct congeners, to belong to at least eight species that were known to be missing from the reference library (Supporting Table S1). Despite this major handicap, the 12S MiFish primers remained superior to COI in terms of congruence with the traditional fish surveys, by recovering a greater overlap of species in all cases. The 12S MiFish primers amplified more species than the traditional surveys at all sites, except Whitsand Bay. This was mainly due to the underrepresentation of the fauna of that site in the 12S reference library, with over half of the surveyed species absent from the library, and a higher proportion of elasmobranchs (five species) than the other sites, which the MiFish-U primers fail to amplify. Overall, no species that were recorded in the traditional surveys were missing from the COI reference libraries, but eighteen species were missing from the 12S MiFish library (37%). The low numbers of species recorded by the traditional surveys at the Esk and Tees sites in comparison to the Whitsand Bay and River Test sites, is partly due to the inherently less diverse fauna of these northerly estuaries, as well as a reflection of the survey techniques, with fyke and seine netting likely to detect fewer species than otter trawling (Whitsand Bay) or a 24 h power station impingement (River Test). It should also be noted that there is no *a priori* assumption that the eDNA and traditional survey data will be completely congruent, as most fish survey methods are imperfect, sampling a moving target of diversity and abundance over difficult-to-define spatio-temporal points. For example, eDNA can be transported in or out by tides, while some species are difficult to sample using particular fishing gears, due to effects of size, behaviour or abundance. Therefore, overlap between eDNA and traditional survey data is best interpreted as a relative measure between the primer sets.

CONCLUSIONS

While PCR-free methods are being actively investigated, it is clear that despite the limitations in quantification, the majority of environmental metabarcoding will be based around amplicon sequencing, at least for the medium term (Creer et al., 2016; Stat et al., 2017; Bista et al., 2018; Wilcox et al., 2018). Particularly important for regulatory applications, or where researchers wish to compare results over time or between studies, some degree of standardisation is desirable (Hering et al., 2018). Our results—and those of previous studies using similar primer sets (Yang et al., 2014; Lim et al., 2016; Bakker et al., 2017; Stat et al., 2017; Jeunen et al., 2018; Macher et al., 2018; Singer et al., 2019)—show that environmental metabarcoding for restricted taxonomic groups using degenerate COI primers results in excessive volumes of "wasted" sequencing effort. This co-amplification of prokaryotic and non-target eukaryotic DNAs and subsequent lack of specificity is due to the nature of mutation patterns in COI (Siddall et al., 2009). Therefore, while we fully support the arguments presented by Andújar et al. (2018) regarding the overall advantages of COI as a bulk-sample metabarcoding marker, we find it difficult to recommend for metabarcoding environmental samples with low target template concentrations and high microbial and plankton diversity, such as natural water bodies.

While the use of multiple primer sets and markers are probably required for a comprehensive view of total biodiversity (Drummond et al., 2015; Stat et al., 2017), for specific taxonomic groups such as fishes a single assay should be a feasible proposition. Unfortunately, no single 12S primer set was shown to be optimal for eDNA fish surveys. The MiFish-U primer set—and *in silico*, the Taberlet et al. (2018) modified versions—performed well in terms of specificity, discriminatory power, and reproducibility. Despite this, MiFish-U is not universal for all fishes, because a separate MiFish-E assay is required to amplify elasmobranchs. The MiFish reference library was also inadequate in this case, missing large numbers of common taxa. The

Valentini-tele01 primer set amplifies actinopterygians and elasmobranchs in a single assay, but suffers from an even more poorly populated reference library than MiFish-U, and weaker taxonomic resolution. The Riaz-V5 primers had the most complete reference library of the 12S primer pairs, but also do not amplify elasmobranchs and have the poorest discriminatory power.

Because no single alternative primer set to COI will be optimal for all applications, it is clear that the current DNA barcode reference libraries will need to be augmented with data from multiple mitochondrial regions to enable their wider utility for vertebrate metabarcoding. However, rather than sequencing individual 12S regions on an ad hoc basis, a better solution is to generate whole mitochondrial genomes which can act as an extended or linking barcode if sequenced from the same collection material (Coissac et al., 2016; Collins and Cruickshank, 2014). Low coverage genome skimming techniques now produce high quality mitogenomes, and are compatible with existing—frequently ethanol-based—tissue collections, and therefore will not require the recollection of specimens (Gillett et al., 2014; Linard et al., 2016). Environmental DNA techniques could potentially be the default survey methodology for aquatic ecosystems, but the existing gap between recovered genotypes and their corresponding phenotypic and historical data can only be filled with substantially more comprehensive reference libraries.

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AUTHOR CONTRIBUTIONS

SM, MJG, DWS, and LC conceived the study and obtained funding; RAC, JB, OSW, and AZS carried out field work and lab work; RAC, JB, and OSW performed the analyses; RAC and JB drafted the manuscript. All authors contributed critically to the drafts and gave final approval for publication.

DECLARATION OF INTEREST

The authors declare that they have no competing interests.

DATA ACCESSIBILITY

The full reference library and code to reproduce it can be found at https://doi.org/10.6084/m9.figshare.7464521. Code to reproduce all other analyses in this study can be found at https://doi.org/10.6084/m9.figshare.8291660. Sequence data (fastq format) generated from the *in vitro* analyses can be found at https://doi.org/10.5061/dryad.b8f6s44.

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Table 1. Primer sets assessed in this study. The approximate fragment length is based upon the length of that region in the *Anguilla anguilla* mitochondrial genome (AP007233.1). The asterisks represent the sequences of the Leray-XT primer set that were simplified by changing inosines to double-base ambiguities to allow an *in silico* assessment with *MFEprimer*. The standard DNA barcode marker for fishes (Ward et al., 2005) is presented for reference.

Primer set	Locus	Primer names	Oligonucleotide 5′–3′	Fragment length (bp)	Reference
Leray-XT	COI	mlCOIintF-XT	GGWACWRGWTGRACWITITAYCCYCC	313	Wangensteen et al. (2018)
		mlCOIintF-XT*	GGWACWRGWTGRACWGTYTAYCCYCC		
		jgHCO2198	TAIACYTCIGGRTGICCRAARAAYCA		
		jgHCO2198*	TAKACYTCWGGRTGRCCRAARAAYCA		
SeaDNA-short		coi.175f	GGAGGCTTTGGMAAYTGRYT	55	This study
		coi.226r	GGGGGAAGAARYCARAARCT		
SeaDNA-mid		coi.175f	GGAGGCTTTGGMAAYTGRYT	130	This study
		coi.345r	TAGAGGRGGGTARACWGTYCA		
Ward-barcode		FishF1	TCAACCAACCACAAAGACATTGGCAC	655	Ward et al. (2005)
		FishR1	TAGACTTCTGGGTGGCCAAAGAATCA		
Minamoto-fish	Cytb	L14912-CYB	TTCCTAGCCATACAYTAYAC	235	Minamoto et al. (2012)
		H15149-CYB	GGTGGCKCCTCAGAAGGACATTTGKCCYCA		
MiFish-U	12S	MiFish-U-F	GTCGGTAAAACTCGTGCCAGC	171	Miya et al. (2015)
		MiFish-U-R	CATAGTGGGGTATCTAATCCCAGTTTG		
MiFish-E		MiFish-E-F	GTTGGTAAATCTCGTGCCAGC	171	Miya et al. (2015)
		MiFish-E-R	CATAGTGGGGTATCTAATCCTAGTTTG		
Taberlet-tele02		Tele02-f	AAACTCGTGCCAGCCACC	167	Taberlet et al. (2018)
		Tele02-r	GGGTATCTAATCCCAGTTTG		
Taberlet-elas02		Elas02-f	GTTGGTHAATCTCGTGCCAGC	171	Taberlet et al. (2018)
		Elas02-r	CATAGTAGGGTATCTAATCCTAGTTTG		
Valentini-tele01		L1848	ACACCGCCCGTCACTCT	63	Valentini et al. (2016)
		H1913	CTTCCGGTACACTTACCATG		
Riaz-V5		12S-V5f	ACTGGGATTAGATACCCC	106	Riaz et al. (2011)
		12S-V5r	TAGAACAGGCTCCTCTAG		
Berry-fish	16S	Fish16sF/D	GACCCTATGGAGCTTTAGAC	219	Berry et al. (2017)
		16s2R	CGCTGTTATCCCTADRGTAACT		

Table 2. Statistics for reference library coverage, taxonomic discriminatory power, and primer universality as estimated by *in silico* PCR, for twelve primer sets from COI, cytochrome *b*, 16S and 12S. Library coverage is calculated as the number of species for which at least one sequence was available out of the total ($n = 531$) or common species subset ($n = 176$) of British Isles marine and freshwater fishes (proportion in parentheses). Library sequences per species is the mean (median in parentheses) number of sequences available for each species. Taxonomic discrimination is the proportion of species for which all individuals can be unambiguously identified by a unique DNA sequence, with values in parentheses showing the proportion for the subset of species that are shared over all primer sets ($n = 221$ for all; $n = 88$ for common). Primer universality represents the mean Primer Pair Coverage (PPC) percent statistic from *MFEprimer*, and was calculated using the 184 British Isles fish species for which data were available for all species. The standard DNA barcode marker for fishes (Ward et al., 2005) is presented for reference. The highly degenerate Leray-XT primers were simplified to overcome analytical RAM limitations (see Table 1).

Locus	Primer pair	Species subset	Total number sequences	Library species coverage	Library sequences per species	Fragment taxonomic discrimination	Primer % universality (Actinopterygii)	Primer % universality (Elasmobranchii)
COI	Leray-XT	All	24,058	481 (0.91)	50 (24)	0.95 (0.96)	27.8	39
	SeaDNA-mid		24,045	481 (0.91)	50 (24)	0.91 (0.94)	23.8	23.7
	SeaDNA-short		23,911	481(0.91)	49.7 (24)	0.87 (0.9)	34.5	21.5
	Ward-barcode		23,975	481 (0.91)	49.8 (24)	0.95 (0.97)	6.3	1.2
CYTB	Minamoto-fish		15,405	344 (0.65)	44.8 (6.5)	0.91 (0.91)	19.1	20.4
12S	MiFish-U		1,904	322 (0.61)	5.9 (3)	0.93 (0.91)	86.6	3
	Taberlet-tele02		1,904	322 (0.61)	5.9 (3)	0.93 (0.91)	92.1	8.3
	MiFish-E		1,904	322 (0.61)	5.9 (3)	0.93 (0.91)	0.6	55
	Taberlet-elas02		1,904	322 (0.61)	5.9 (3)	0.93 (0.91)	0.5	83.6
	Valentini-tele01		1,699	273 (0.51)	6.2 (2)	0.86 (0.85)	77.1	68.2
	Riaz-V5		2,416	364 (0.69)	6.6 (2)	0.79 (0.78)	92.2	11.2
16S	Berry-fish		4,089	411 (0.77)	9.9 (4)	0.89 (0.86)	50.9	0
COI	Leray-XT	Common	12,698	170 (0.97)	74.7 (38.5)	0.97 (1)	23.3	49.3
	SeaDNA-mid		12,639	170 (0.97)	74.3 (37.5)	0.93 (1)	17.6	30
	SeaDNA-short		12,553	170 (0.97)	73.8 (37.5)	0.93 (1)	32.8	28.9
	Ward-barcode		12,579	170 (0.97)	74 (37.5)	0.97 (1)	6.3	0
CYTB	Minamoto-fish		10,936	143 (0.81)	76.5 (16)	0.94 (1)	19.2	12.9
12S	MiFish-U		941	109 (0.62)	8.6 (3)	0.94 (0.94)	91.8	0
	Taberlet-tele02		941	109 (0.62)	8.6 (3)	0.94 (0.94)	96.4	0
	MiFish-E		941	109 (0.62)	8.6 (3)	0.94 (0.94)	0	73.3
	Taberlet-elas02		941	109 (0.62)	8.6 (3)	0.94 (0.94)	0	100
	Valentini-tele01		852	99 (0.56)	8.6 (2)	0.93 (0.94)	76.4	68.2
	Riaz-V5		1,398	143 (0.81)	9.8 (3)	0.85 (0.83)	96.4	0
16S	Berry-fish		2,296	167 (0.95)	13.7 (6)	0.87 (0.91)	53.9	0

Table 3. Number of reads remaining after seven bioinformatic steps, as well as the number of estimated reads for taxonomic groups (assignments were carried out on the reads remaining after the homology search step 7). Fish reads (putative) are reads assigned to fishes based on the best scoring *blastn* hit using the NCBI "nt" blast database. Fish reads (assigned) are reads assigned to fish species by the stringent taxonomic identification step using *blastn* and *EPA-ng* on our curated reference library. Fish reads (unassigned) are putative fish reads that could not be assigned to species by the stringent taxonomic identification step.

Filtering step	COI Leray-XT	COI SeaDNA-mid	COI SeaDNA-short	12S MiFish-U
Total passing filter	5,967,313	14,291,168	8,881,088	3,436,278
(1) Detect primers	4,828,799	11,535,904	6,428,030	2,776,073
(2) Demultiplex	4,648,811	10,879,223	5,994,815	2,473,594
(3) Trim primers	4,618,236	10,300,907	5,852,555	2,462,936
(4) Quality filter	4,519,097	10,344,024	5,856,045	2,455,532
(5) Merge	3,395,057	9,658,709	4,804,502	2,383,162
(6) Remove chimaeras	3,225,240	9,404,746	4,416,647	2,271,541
(7) Homology search	3,223,743	8,703,109	4,074,123	2,157,365
Bacteria	1,476,994	1,388,681	2,242,220	4
Eukaryota	1,745,295	7,294,762	1,815,928	2,157,361
Metazoa	321,590	1,161,769	412,871	2,157,361
Chordata	6,351	337,901	250,650	2,157,361
Fish (putative)	2,371	234,219	193,593	1,637,728
Fish (assigned)	1,368	109,486	30,026	1,564,351
Fish (unassigned)	1,003	124,733	163,567	73,377

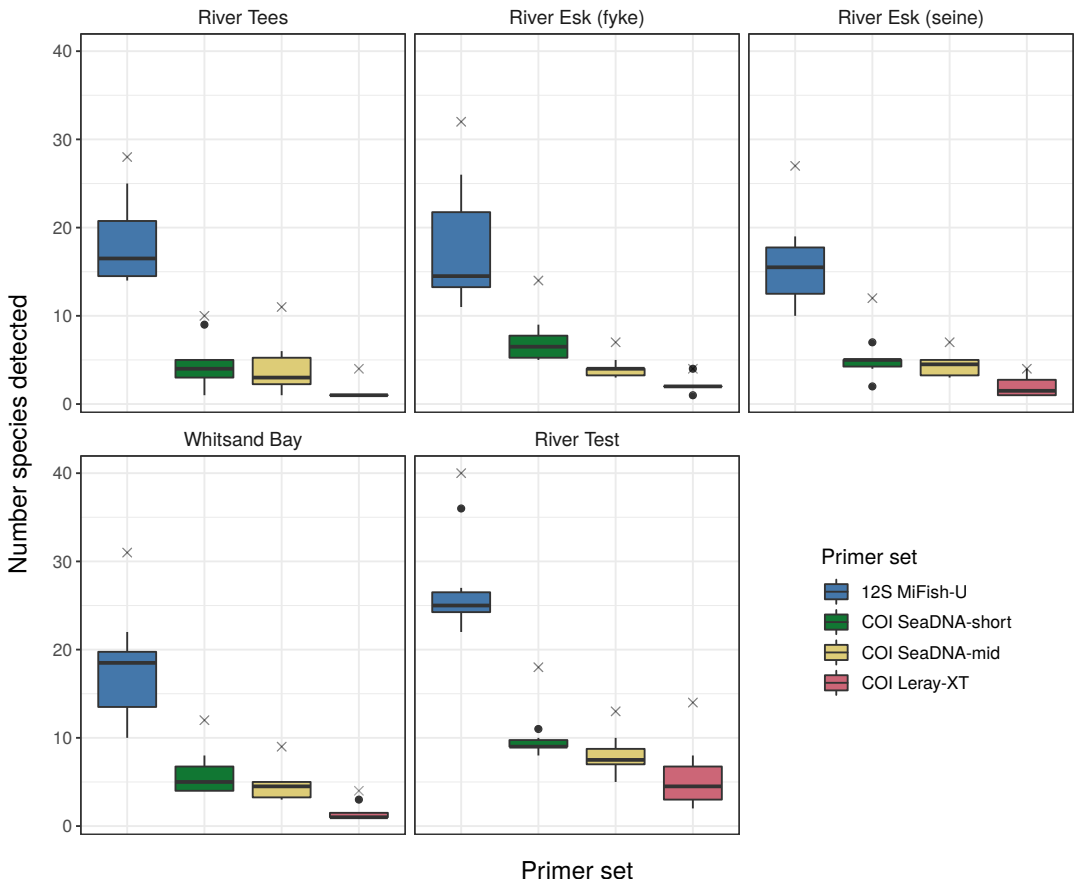

Figure 1. Fish species richness as estimated by four primer pairs at five sampling locations. Per primer-location combination there are three water sample replicates and two uniquely tagged PCR replicates ($n = 6$). The horizontal represents the median value, the boxes represent the 25–75th percentiles, the whiskers represent the values less than 1.5 times the interquartile range, dots represent the outlying data points, and crosses represent the cumulative number of species.

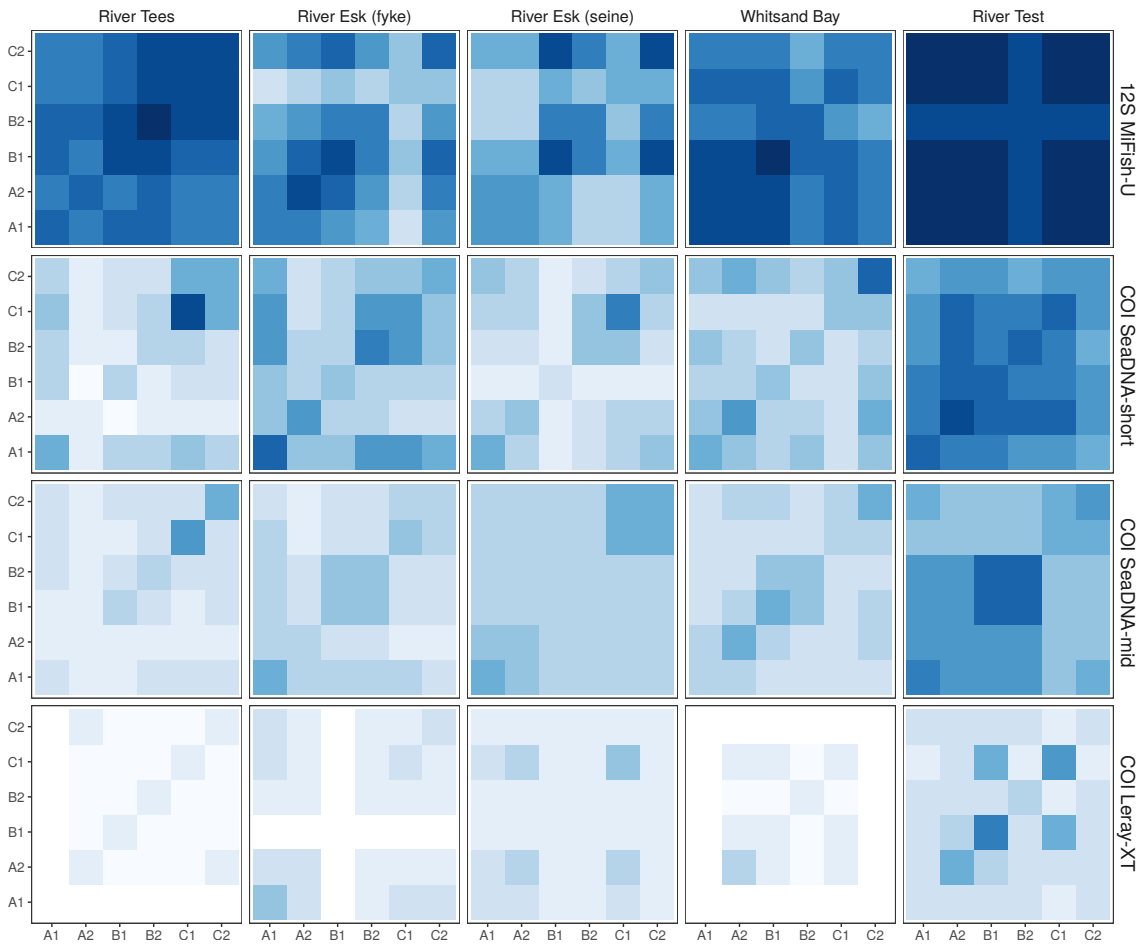

Figure 2. Reproducibility heatmaps of four primer pairs at five sampling locations for the top ten fish species found at each location by read abundance. Letters A, B, and C represent the three water samples taken, while numbers 1 and 2 represent the independent PCR reactions with uniquely tagged primers. There are ten shades showing 10% increments. The darkest shade shows a reproducibility of 100%, i.e. reads from all of the ten species were common to both PCRs. The lightest shade shows 0% reproducibility, i.e. none of the species were present in both of the PCRs. Diagonals show the proportion of the top ten species amplified in that single PCR.

Figure 3. Overlap between fish species found by eDNA metabarcoding (red) and traditional fish surveying (blue). Sizes of circles are proportional only within each primer-location comparison, and not between. Numbers represent number of species in each set. Only marine and estuarine species are shown; freshwater species recorded by the eDNA surveys were removed to allow an equivalent comparison.