

Water eDNA metabarcoding is effective in detecting non-native species in marinas, but detection errors still hinder its use for passive monitoring

Marjorie Couton, Laurent Lévêque, Claire Daguin-Thiébaut, Thierry Comtet, Frédérique Viard

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12	Marjorie Couton ^a , Laurent Lévêque ^b , Claire Daguin-Thiébaut ^a , Thierry Comtet ^a , Frédérique
13	Viard ^{a,c*}
14	^a Sorbonne université, CNRS, UMR 7144, Station Biologique de Roscoff, Place Georges
15	Teissier, 29680 Roscoff, France
16	^b Sorbonne université, CNRS, FR 2424, Station Biologique de Roscoff, Place Georges
17	Teissier, 29680 Roscoff, France
18	^c ISEM, Univ Montpellier, CNRS, EPHE, IRD, Montpellier, France
19	* Corresponding author: frederique.viard@umontpellier.fr
20	ORCID ID:
21	Marjorie Couton : 0000-0001-9880-8646
22	Frédérique Viard: 0000-0001-5603-9527
23	Claire Daguin-Thiébaut 0000-0003-4665-6411

24 Abstract

Marinas are high-priority targets for marine non-indigenous species (NIS), where they compose a large portion of the biofouling communities. The practicality of water samples collection makes environmental DNA (eDNA) metabarcoding an interesting tool for routine NIS surveys. Here the effectiveness of water-eDNA-metabarcoding to identify biofouling NIS, in ten marinas from western France, was examined. Morphological identification of specimens collected in quadrats brought out 18 sessile benthic NIS beneath floating pontoons. Water-eDNA-metabarcoding detected two thirds of them, failing to detect important NIS. However, sampling and bioinformatics filtering steps can be optimized to identify more species. In addition, this method allowed the detection of additional NIS from neighboring micro-habitats. Caution should however be taken when reporting putative novel NIS, because of errors in species assignment. This work highlights that water-eDNA-metabarcoding is effective for active (targeted) NIS surveys, and could be significantly improved for its further use in marine NIS passive surveys.

Keywords: biological invasions, environmental DNA, metabarcoding, morphology, ports, quadrats

Introduction

As the number of marine introductions is growing worldwide (Seebens et al. 2017), much effort is directed towards non-indigenous species (NIS) management. The implemented strategies can vary greatly from, limiting the spread of new species by applying regulation on vectors of introduction, to eradication protocols to remove invasive species from a particular area (Robertson et al. 2020). Nevertheless, vectors of introductions are not efficiently regulated (Carney et al. 2017), and measures of eradication are difficult to apply when non-indigenous species (NIS) have become too abundant or widespread (Ojaveer et al. 2015). Monitoring NIS to detect new arrivals or to evaluate the spread of already established species, is thus crucial to support efficient management strategies (Robertson et al. 2020). It is especially true in the primary introduction sites, such as ports or marinas. In such habitats, fouling communities from artificial hard substrates are often targeted since they include many NIS (Dafforn et al. 2009; Leclerc et al. 2020).

Diverse traditional approaches exist to monitor biofouling communities, from Rapid Assessment Surveys or full species inventories (Lehtiniemi et al. 2015) to passive sampling strategies using settlement plates (Dafforn et al. 2009; Outinen et al. 2021). Their inherent issues, however, might prevent their broad use by practitioners and hinder their efficiency. Most of the problems with traditional approaches arise from the morphological identification of organisms. They require taxonomic expertise, and the time constraint of field sessions can lead to erroneous identification for complex taxa (eg the ascidians *Botrylloides* spp., Viard et al. 2019). Furthermore, these approaches do not allow the detection of introductions by (pseudo-)cryptic species, common among NIS, especially in biofouling assemblages (Viard et

al. 2016; Turon et al. 2020). Finally, they are not always sufficient to discriminate species at early-settlement stages. There is thus a particular need for a new method solving these issues.

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Molecular approaches, especially DNA metabarcoding, have been shown to overcome these limitations (Zaiko et al. 2015; Rey et al. 2020). Metabarcoding has been successfully applied to various types of samples in the context of NIS surveys (eg plankton: Couton et al. 2019, sediment: Shaw et al. 2019, water: Borrell et al. 2017, homogenized organisms: Pearman et al. 2021), which provide different results, reflecting the composition of different communities (Koziol et al. 2018; Fernandez et al. 2021). Rey et al. (2020) suggested that, for NIS surveys, all types of samples (or a combination of several of them) should be simultaneously processed (see also Holman et al. 2019). However, this might be hardly feasible for regular monitoring, as it would require expensive sampling tools and expertise that is not always available, and would be very time-consuming. In that context, water environmental DNA (eDNA) would be the sample type of choice as seawater can be easily and rapidly sampled in ports, even by non-experts with a minimal training, and the sampling protocol can be highly standardized (Morisette et al. 2021). Moreover, being non-destructive, water sampling can be conducted as often as required, allowing fine temporal surveys, which is recommended as many NIS are short-lived and show seasonal patterns of abundance (Couton et al. 2019). Finally, water might allow to get a broader picture of NIS present in a given marina, encompassing at the same time species living in the water column or on the bottom (Stat et al. 2017).

Very few studies investigated the reliability of water-eDNA-metabarcoding to study benthic communities, even less when considering hard-bottom, and they provided contrasting results. For example, while Leduc et al. (2019) and Antich et al. (2021) reported many false

negatives (only up to 15% benthic species/MOTUs retrieved in water-eDNA samples), West et al. (2021) recovered 70% of coral taxa. Many authors thus preferentially used bulk samples from scraping of either hard bottoms or settlement plates (Duarte et al. 2021). This approach, however, is not free of bias, and requires processing a large amount of biomass, which is time-consuming and destructive for the community (van der Loos and Nijland 2021). As an example, von Ammon et al. (2018) did not recover 7 NIS out of 48 after metabarcoding of bulk communities from settlement plates. On the other hand, the fact that several authors identified DNA of benthic species in water samples (eg Borrell et al. 2017; Cordier et al. 2019), particularly with species-specific approaches (eg Ardura et al. 2015; Simpson et al. 2017), is encouraging. The practicality of water-eDNA sampling drives the need to determine its reliability to monitor biofouling communities in marinas, to understand why metabarcoding is not always able to detect benthic species and to evaluate how these limitations could be overcome (Trebitz et al. 2017; Blackman et al. 2019; Darling et al. 2020; Makiola et al. 2020).

Some of the limitations inherent to the use of water-eDNA metabarcoding are already well known, and previous works made suggestions to circumvent them. One of these limitations is the degradation of DNA in water over time which can be partially addressed by using smaller marker regions (Taberlet et al. 2012). Others, such as DNA transport and dilution or differential shedding rates between species can also be decreased by sampling as close as possible to the target community (Antich et al. 2021). However, even when all these factors are taken into account to maximize the chances of species detection, the likelihood of false detections or non-detections is still not negligible. It is therefore crucial to evaluate the remaining biases and investigate possible explanations. In addition, the whole process, from

sampling to taxonomic assignment, has rarely been considered at once to determine which steps are the most critical to be optimized.

Here, the capacity of water-eDNA-metabarcoding to detect NIS from biofouling communities in marinas was investigated. The goals were to 1) evaluate the extent of species detection errors, and 2) explore the possible causes for these errors, and propose possible solutions. To this end, water sampling was carried out along floating pontoons in ten marinas and two seasons. This sampling strategy allowed to examine detectability at a regional scale and over time, the latter being particularly important as many benthic sessile NIS display short life cycles. Metabarcoding was performed using three markers (18S, COI, and 16S) to increase the detection capacity. Simultaneously, a list of sessile benthic NIS present in the study marinas was established by identifying morphologically and *in situ* species sampled within quadrats beneath the same floating pontoons. This list was used to evaluate the number of false negatives (species observed at the time of sampling but not recovered with eDNA), and explore various causes for metabarcoding detection failures, such as sampling effort or bioinformatics processing.

Materials and methods

Sampling

Biofouling communities immersed beneath floating pontoons (ie floating structures used as docks for mooring leisure boats) were sampled in ten marinas around Brittany (France; Fig. 1). These locations are known to harbour different fouling communities since they are spread over two ecoregions *sensu* Spalding et al. (2007), namely the Celtic seas and

the North European Atlantic Shelf, characterized by different hydrological and physical parameters and species assemblages, and encompassing a biogeographic transition zone between the Western English Channel and the NE Atlantic (Gallon et al. 2014). Although native species should particularly differ between these two regions, differences were also observed for NIS, as shown by previous Rapid Assessment Surveys (Bishop et al. 2015; FV and LL, unpublished data). Since many biofouling species are short lived and seasonal, sampling was performed in fall (October) 2017 and spring (May-June) 2018. Two pontoons were selected in each marina, to include potential differences in environmental conditions such as water flow, salinity or type of boats docked (long distance travellers *vs.* local boats).

Seawater was collected every 5 m along a 50-m transect per pontoon. At each sampling point, 200 mL of seawater were collected in a plastic jar using a sterile 100-mL syringe fixed to a sampling rod and immersed at 1 m below the surface. A total volume of 2 L was thus collected per pontoon (after pooling all sampling points). To investigate the effect of sampled volume, in spring 2018, an additional 2-L replicate was collected, resulting in a total volume of 4 L per pontoon (the two 2-L replicates were processed separately). The seawater content of each jar was immediately filtered on site on a Millipore® SterivexTM filter unit (0.22 μm) using a Masterflex® L/S® economy drive peristaltic pump. After filtration, 2 mL of preservation buffer (sucrose 0.75 M, Tris 0.05 M pH 8, EDTA 0.04 M) were added before storing the filter at -20°C in a portable freezer, until transfer to the laboratory. All consumables not sold as DNA-free were previously immersed in 12.5 % commercial bleach (ca. 0.65 % hypochlorite) for at least 30 min, rinsed with ultrapure water and placed under UV light for at least 15 min. The sampling rod was immersed in 12.5 % commercial bleach for at least 30 min before sampling in a given marina, and rinsed with tap water between each pontoon. Six controls (ie one in three different marinas for each season) were performed by

sampling, for every control, 10 replicates of 200 mL of ultrapure water every 5 m along a 50-m transect on a pontoon, with the same equipment, and after applying the aforementioned decontamination protocol, thus mimicking the seawater sampling protocol. The 10 replicates from a pontoon were then pooled, as were the real samples, to reach a total volume of 2 L per control.

A list of sessile benthic NIS present, at the time of water collection, in the biofouling community was established. Four taxonomic classes were targeted, namely Gastropoda, Bivalvia, Ascidiacea and Gymnolaemata known to include a substantial number of NIS in the study region (Bishop et al. 2015) (see below for an explanation). Biofoulers were collected beneath pontoons at the same points where water samples were collected, to maximize the probability to sample the same taxa, as shown by Antich et al. (2021). At each of the ten sampling points, all organisms within a 0.25-m² quadrat were scraped by divers and placed in a closing bag. Importantly, the scrapping was performed after the water sampling to avoid any contamination. Specimens (> ca. 5 mm) were morphologically identified on site, to the lowest possible taxonomic level (family, genus or species). The identified organisms were not systematically collected, nor barcoded but, as part of a previous work (Couton et al. 2019), regular barcoding was carried out to get local references on the four taxonomic groups studied here. In order to refine field identification for erected bryozoans from the class Gymnolaemata, specimens were collected at each site for further laboratory identification.

DNA extraction

Caution was taken to avoid any contamination of DNA extracts, as recommended by Goldberg et al. (2016). All equipment and bench surfaces were DNA decontaminated before use and consumables not sold as DNA-free were processed as detailed above. DNA was

extracted from the Sterivex units using the NucleoSnap® Finisher Midi extraction kit (Macherey-Nagel) following the manufacturer's protocol with a few modifications (Supplementary Protocol 1).

Library preparation and sequencing

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All steps were performed following strict rules for preventing contamination, such as UV irradiation of all tips, plates and tubes, and use of filter tips. Libraries were prepared using a dual-barcoded, dual-indexed two-step PCR procedure. Each DNA sample was amplified 15 times with each of the three primer pairs targeting a different marker: i) a 313-bp COI portion amplified using primers designed by Leray et al. (2013), ii) a 350-bp to 450-bp portion of the 18S rDNA V1-V2 region amplified using primers designed by Fonseca et al. (2010), and iii) a 115-bp portion of the 16S rDNA amplified using primers designed by Kelly et al. (2016) (fragment sizes excluding primers). Combining these markers would increase the detection ability, as they differ in their taxonomic coverage and resolution, and conservation of primer binding sites (Couton et al. 2019; Makiola et al. 2020). Detailed PCR protocols are available in Supplementary Protocol 2. Five tagged PCR replicates were performed for each sample and marker (Table S1). After this first PCR, the five tagged replicates from the same DNA extraction sample were pooled according to their intensity on the agarose gel visualisation. A second PCR was performed to complete Illumina® adapters and to tag samples with a unique dual-index combination (Table S2). Sequencing was performed for each marker separately on an Illumina® MiSeq sequencer. The COI and 18S markers were sequenced using a 600 cycles v3 protocol and the 16S marker was sequenced using a 300 cycles v2 protocol, both with two index reads.

Reads processing and filtering

After demultiplexing performed by the sequencing machine based on index combination (defining the samples), reads were assigned to each of the five PCR replicates according to their tag combination. Primers and tags were then removed using CUTADAPT v-2.8. A set of amplicon sequence variants (ASVs) was produced using DADA2 v-1.13.1 (Callahan et al. 2016); see Table S3 for detailed parameter values). When demultiplexing, reads can be falsely assigned to a sample when an index combination is not correctly recognized, which is referred to as 'index-jump' (Taberlet et al. 2018). To assess this phenomenon, 12 index combinations, not used in the described PCR experiments, and chosen among the 96 available, were added to the MiSeq sequencing sample sheet in order to get the corresponding fastq files. For each ASV, the number of reads assigned to one of these internal control index combinations was divided by the total number of reads in the complete dataset for this same ASV. The maximal proportion occurring in an index control was recorded for each marker (Table S3). Any ASV that did not account for more than this maximal proportion in a sample was discarded from this sample. Furthermore, only ASVs found in at least two out of the five PCR replicates per sample were retained.

Taxonomic assignment

A first taxonomic assignment was performed to assess the proportion of reads assigned to metazoans. For that purpose, each dataset was aligned against references retrieved from the GenBank nt database using the ecotag command from OBITOOLS v-1.2.11 (Boyer et al. 2016) with no minimum identity threshold. To perform taxonomic assignment at lower taxonomic levels, assignment was later performed by aligning all ASVs against a restricted reference database using the BLAST® command line tool (Altschul et al. 1990). This database was

composed of sequences retrieved from GenBank and sequences produced locally (Couton et al. 2019), targeting ten metazoan phyla commonly found in marinas: Annelida, Arthropoda (only Crustacea), Bryozoa, Chordata (only Actinopterygii, Chondrichthyes, and Tunicata), Cnidaria, Echinodermata, Mollusca, Nemertea, Platyhelminthes, and Porifera. Only alignments covering 99% of the subject sequence and displaying a minimum identity threshold (18S: 99%; COI: 92%; 16S: 97%) based on barcoding gaps computed from reference sequences (Figs. S1-S3), were considered. If one ASV matched with several references, it was assigned to the one with the highest identity percentage. If two alignments with different references had the same identity, the ASV was assigned to the lowest common taxonomic rank. All assignments to a rank higher than the family were classified as 'unassigned'.

To evaluate the ability of water-eDNA-metabarcoding to retrieve species observed in the field, the metabarcoding dataset was restricted to the four classes targeted in the quadrats survey (ie Ascidiacea, Bivalvia, Gastropoda, and Gymnolaemata). These classes were chosen because they are known to include a substantial number of NIS in marinas of the study region (Bishop et al. 2015), and because the involved team members have a good expertise for their morphological identification. This expertise is further reinforced by a number of previous studies combining morphological and molecular approaches (Bishop et al. 2013; Bishop et al. 2015; Viard et al. 2019; Turon et al. 2020). In addition, for these four classes, local reference data for native species and novel NIS were previously obtained (Couton et al. 2019). This selection thus allowed minimizing issues due to the incompleteness of reference databases, and better assess the effect of the taxonomic assignment step in the metabarcoding process (ie, use of a curated database).

Results

NIS observed in the field

From the 400 quadrat samples collected, 93 taxa were identified, with differences across sites and seasons (Table S4). Most (n=62) were identified to species, 20 to genus, 8 to family, and 3 to higher levels. Forty-seven sessile or poorly mobile taxa belonged to the four targeted classes (ie Ascidiacea, Bivalvia, Gastropoda, and Gymnolaemata). In these, 15 NIS (incl. two pairs pooled in the field: *Botrylloides diegensis* and *B. violaceus*, and *Didemnum vexillum* and *D. pseudovexillum*) and three cryptogenic species were identified (Table S4), all previously reported in the study area. Among the 47 taxa observed, only 43 were considered for assessing the ability of metabarcoding to detect them. *Ruditapes* sp. was removed from the positive control list because it lives buried in soft sediments and is thus likely to be inconsistently recovered in quadrat surveys. *Membraniporella nitida*, *Scruparia ambigua*, and *Amathia gracilis* were excluded because they usually did not fulfil the chosen size criteria of 5 mm, and thus the observed species list might also be biased across locations.

Taxonomic assignment of sequence reads

The MiSeq runs yielded 15,754,233 reads, 13,108,322 reads, and 13,396,162 reads for 18S, COI, and 16S, respectively, resulting in 3972, 4258, and 347 unique ASVs. After filtering, all samples (all tagged PCR replicates combined for each sample) contained a mean of 253,600 reads, 188,853 reads, and 199,910 reads for 18S, COI, and 16S, respectively.

Only a small proportion of reads were assigned to Metazoa for 18S (23%) and COI (24%), contrasting with the results for 16S (98%; Fig. 2). The most abundant metazoan phyla identified across markers were Arthropoda, Chordata, Cnidaria, Annelida, and Mollusca, totalizing 88%, 35% and 76% of the reads for 18S, COI and 16S, respectively, with sharp differences between markers. For instance, bryozoans accounted for 14% of 16S reads but for less than 1% of 18S and COI reads. Differences were also observed when assigning ASVs to the ten targeted metazoan phyla. For instance, almost three times more taxa were identified with 18S, and almost four times more species were identified with COI, as compared to 16S (Table 1). Overall (60 2-L samples), 557 taxa including 409 species were identified by at least one marker, over the ten selected phyla (Table S5-S8). Among the 154 taxa from the four classes compared with morphological data, 114 were identified at the species-level across all markers (Table S5).

Recovery of species known to be present by water-eDNA-metabarcoding

Half of the 43 taxa (n=21) observed in quadrats were unambiguously recovered by water-eDNA-metabarcoding (Fig. 3, column Tot). A higher number were identified with 18S (14) as compared to COI (10) and 16S (5), with 18S being more efficient for Ascidiacea, and the two latter being more suited for Bivalvia and Gymnolaemata (Fig. 3). Four species could not be recovered because reference sequences were lacking for any of the three markers. Five taxa were not identified because of identical reference sequences (ie lack of taxonomic resolution with the markers used). For example, *Ascidiella aspersa* and *A. scabra* were distinguished in the quadrats sampling, based on morphology, when metabarcoding only identified them at the genus level, because of identical 18S references for the two species, and amplification failure for this genus with COI and 16S (Figs. S4-S5).

Focusing on the 18 NIS and cryptogenic taxa known to be present, a large proportion (66.7 %) were detected in eDNA with at least one marker (Table 2). Out of the six taxa not recovered, four (*Botrylloides violaceus*, *Crassostrea gigas*, *Bugulina stolonifera*, and *B. simplex*) had reference sequences identical to sequences of closely related species, resulting in the assignment to the genus. For instance, *Bugulina stolonifera* and *B. simplex* share the same sequence for the only marker able to retrieve them (16S). The other two NIS (*Didemnum vexillum*, and *Tricellaria inopinata*) remained undetected despite the availability of references for at least one marker, their large abundance in some locations, and their presence in almost all localities at both seasons, suggesting PCR amplification biases with the markers used.

The recovery rate of NIS by water-eDNA-metabarcoding was much lower when examining in details the distribution of taxa, across localities and seasons, with only 23.9 % overlap (119 out of 498 cases; Fig. 3). Focusing on NIS or cryptogenic species, out of the 239 occurrences in quadrats, only 73 (30.5 %) were successfully recovered with water-eDNA-metabarcoding (Fig. 3).

Additional detections with water-eDNA-metabarcoding

Within the four targeted classes, 114 species were detected with water-eDNA-metabarcoding. All species identified were classified as 'expected' vs. 'unexpected', based on previous knowledge on their presence/absence in the study area, as reported in the Roscoff biological species inventory (http://abims.sb-roscoff.fr/inventaires/) and in the literature, or as reported before by the team members involved in this study. Eighteen species were 'unexpected' species, and thus might be novel NIS for the study area. Sixteen of them were assigned with more than 99% identity (Table S9). All were, however, identified with only one

marker, except *Mytilus trossulus*, identified with COI and 16S. None of them were seen in the quadrats.

Regarding the distribution of taxa across localities, water-eDNA-metabarcoding identified taxa in localities where they were not reported based on quadrats sampling in eight cases (eg Ascidia mentula in SM and BLO during spring; Fig. 3). Of these eight cases, only three are associated to a NIS or a cryptogenic species (*Crepidula fornicata* in MB, Watersipora subatra in ET, and Bugulina fulva in ET). Moreover, water-eDNA-metabarcoding efficiently identified three Molgula species (Molgula socialis, M. bleizi, and M. complanata) when only the genus could be confidently recovered in the field. Finally, this method allowed the identification of two additional NIS, Mya arenaria and Ruditapes philippinarum, known to be present in the study region but living buried in soft sediments.

Effect of the sampling strategy on taxa detection by water-eDNA-metabarcoding

Less than half of the taxa from the ten metazoan phyla (42.5 %) were shared between the two pontoons sampled in a given locality with water-eDNA-metabarcoding (Fig. 4). The proportion of shared taxa ranged between 22.6 % (CAM, fall) and 50.8 % (PG, fall). Similarly, the total number of taxa was slightly higher when merging the results obtained with the two 2-L replicates (482 taxa) than when using only one of them (436 and 428, for the two series analysed separately), over all marinas, in the spring sampling. This result is consistent at every pontoon, with a significantly higher number of taxa when using 4 L of water than when using one of the two 2-L samples (Wilcoxon tests for paired data: V = 0, P < 0.001 for the two series). When focusing on the four targeted classes, the number of taxa unique to one 2-L replicate ranged from 0 to 8 for 16S and 18S and 0 to 7 for COI (Fig. 5). The proportion of reads assigned to these unique taxa was very low for 16S (ranged from 0 to 24 %; mean

value: 1.6 %), but was substantially higher for 18S (ranged from 0 to 66.1 %; mean value: 6.8 %), and even higher for COI (ranged from 0 to 100 %; mean value: 29.5 %). This proportion was largely influenced by the total number of reads associated to all species from the four target groups within each replicate for the different markers. These numbers ranged from 5,439 to 381,749 reads with 16S (mean: 116,582.1), 109 to 37,921 reads with 18S (mean: 8,370.6), and 1 to 6,191 reads with COI (mean: 1,764.7). For example, in the replicates of the pontoon 2 from AW, only two species from the four groups were identified with COI, one different in each replicate associated with 13 and 51 reads respectively (Fig. S6), leading to the proportion of 100% (Fig. 5).

Discussion

The growing 'hardening' of marine coastal areas offers a wide range of new artificial substrates for fouling organisms, many being non-indigenous species (NIS). Regular monitoring of ports and marinas is thus critical for implementing efficient management strategies (Trebitz et al. 2017). In that sense, species detection through water-eDNA-metabarcoding is an appealing approach promising fast and easy sampling, and reduced costs. However, its reliability and the extent of its biases when focusing on benthic NIS are still unclear (Holman et al. 2019; Antich et al. 2021). In particular, one need to assess the risks of both false negatives (which species are not detected and why; eg Hansen et al. 2018) and false positives (detecting species which are not truly present; Darling et al. 2021). Here, the effectiveness of water-eDNA-metabarcoding to identify the presence of biofouling animals (particularly NIS) over the study area was shown, as it correctly recovered half of the 43 taxa, and two third of the 18 NIS or cryptogenic species observed in the field. In addition, as

expected, water-eDNA-metabarcoding retrieved species from adjacent and non-targeted micro-habitats, refined the assignment at the species level for 2 taxa identified at higher taxonomic level in the field, and detected one previously unreported species, putatively a novel NIS, in the study area. However, others species remained undetected in eDNA, with important discrepancies locally. These limitations, and solutions to circumvent them, are discussed below.

Improving the quality of reference data and targeting several markers is needed to increase the detection effectiveness

Half of the species observed in quadrats, and one-third of NIS, were not recovered with water-eDNA-metabarcoding, including abundant ones. For instance, the solitary tunicate *Ascidiella aspersa*, native in the study area, and invasive in other coastal areas in the world, was found in every quadrat sample, reaching up to 384 individuals m⁻² in MB (Table S4), but remained unassigned with metabarcoding. The detection failure was due to both a lack of resolution with 18S for this genus (same sequence as *A. scabra*), and amplification failures for COI and 16S (Figs. S4-S5). Six other abundant NIS (tunicates: *Botrylloides violaceus*, *Didemnum vexillum*; bryozoans: *Tricellaria inopinata*, *Bugulina simplex*, and *Bugulina stolonifera*; bivalves: *Crassostrea gigas*) also remained undetected in the molecular dataset. At a first glance, these results contrast with those by Antich et al. (2021) in which abundant benthic MOTUs were more easily detected in water-eDNA than rare ones. However, in their study, the benthic data were obtained from bulk metabarcoding, and abundant benthic species might have also been missed because of amplification biases or low marker resolution.

The substantial proportion of false negatives observed highlights flaws that may still impede the routine use of water-eDNA-metabarcoding. Those include species-specific DNA

shedding rate and decay (Andruszkiewicz et al. 2021; Holman et al. 2021), differential primer binding efficiency between taxa (Piñol et al. 2018), and issues related to taxonomic assignment. As several studies targeting benthic species showed the occurrence of their DNA in water using species-specific methods (eg Simpson et al. 2017) their shedding rate must be sufficient for detection. This is why more effort should be dedicated to design new primers targeting markers with high resolution, where the sensitivity of each primer should be reported to identify easily species that cannot be detected with them (Xia et al. 2021). For instance, although abundant in quadrats from every location, the invasive colonial ascidian Didemnum vexillum, was not recovered in water-eDNA, whereas Simpson et al. (2017) detected it in water samples from Australia using species-specific primers. Similarly, the invasive tunicate B. violaceus remained undetected because of identical sequences with congeneric species, although it can be distinguished from the introduced B. diegensis and native B. leachii using genus-specific primers (Couton et al. 2021). Issues related to taxonomic resolution/accuracy were here (partly) addressed by combining several markers, as suggested by Zhang et al. (2018), and by producing reference sequences for at least one marker for each NIS observed within quadrats.

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The availability of reference sequences is another well-known problem, which depends on the marker, COI being the most populated in public databases (Porter and Hajibabaei 2018), or on the taxonomic group. For instance, a barcode search in BOLD revealed that only 14% and 5% of the 2,925 and 5,651 accepted species, for the classes Ascidiacea and Gymnolaemata, respectively, were associated to a COI sequence (WoRMS and BOLD search in August 2020), a proportion that could even be lower when considering quality (Paz and Rinkevich 2021). This proportion seems higher for NIS: Briski et al. (2016) reported proportions of 59% and 72% of 45 and 36 NIS, for the same two classes and marker,

respectively. These higher values might be explained by the long-lasting and still increasing use of DNA barcoding in the study of marine NIS (Duarte et al. 2020). Such an imbalance in the proportion of sequences available for native and non-indigenous species might, in turn, increase the risk of false detections of new introductions, as will be discussed below.

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Improving sampling strategy, data production and data processing to overcome limitations

Increasing the sampling effort and optimizing data processing can substantially decrease the detection failures of water-eDNA-metabarcoding. Sampling water at two pontoons in each marina allowed increasing (sometimes doubling) the number of metazoan taxa recovered. For example, in the three marinas where it was detected with water-eDNAmetabarcoding in spring (BL, CAM, and SM), Asterocarpa humilis was identified from one pontoon only. Furthermore, doubling the volume of water filtered improved the detection of rare species (ie low number of reads), which might be particularly useful for NIS early detection. Although the sampling effort (volume and number of replicates) is within the range of that recommended and/or mostly used in other studies (eg Grey et al. 2018; Holman et al. 2019; Duarte et al. 2021), it was not sufficient to recover all the species observed in the field. The sampling standards (eg water volume, sampling points) should thus be increased substantially when targeting specific benthic species, such as particular NIS, because their DNA might be scarce in the water, especially those lacking pelagic stages (Leduc et al. 2019). This would still keep reduced labor- and time-costs in the field as compared to traditional methods (sampling and analyzing 400 quadrats in the field required eight people over two weeks per season). The results also underlined the importance of seasonal sampling, a much less addressed issue. Because many biofoulers, such as tunicates or erected bryozoans, are short-lived, their abundance can vary greatly depending on the season. Their DNA shedding and decay rates can also differ throughout the year (eg reproduction period; Salter 2018;

Wacker et al. 2019). The tunicate *Ciona robusta*, for example, would not have been identified from spring samples only, which is congruent with what is known about its settlement period (mostly in late summer/autumn) in the study area (Bouchemousse et al. 2017). In this regard, sampling effort should be maximized, with an emphasis on spatial and temporal replicates.

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Data filtering steps are crucial in any metabarcoding analysis, notably to avoid false positives, but it may decrease detection sensitivity. Here, all taxa identified were retained after index-jump and replicate filtering in the whole dataset, with three exceptions (Musculus sp., Bugulina stolonifera, and B. simplex; Figs. S7-S8) but the detection capacity was much lower at the unit level (ie 'species-site-date' combination). Before any corrections, water-eDNAmetabarcoding recovered a given taxon in 41% of all units. This number dropped to 22% after corrections, with 106 units of water-eDNA-metabarcoding detection disappearing. Considering NIS or cryptogenic species, this proportion decreased from 56% to 30% (Figs. S7-S8). PCR replicate filtering, used to limit false positives due to PCR errors (Alberdi et al. 2018; Tsuji et al. 2020), resulted in 32% loss of detection (Figs. S4-S5), while most losses (68%) occurred after index-jump filtering. Choosing filtering parameters is thus critical and depends on the study aim. For example, von Ammon et al. (2018) kept all OTUs, even singletons, to increase their chance of detecting rare species, advocating that the cost of missing even one rare NIS would be higher than that of a false positive. This option might, however, lead to a dramatic increase in the number of false detections due to index-jump. Indeed, in the dataset, all index-jump controls exhibited reads belonging to several ASVs. Index-jump is an important issue very delicate to account for in metabarcoding studies (Taberlet et al. 2012), and more work needs to be done to find the right compromise between the proportion of false positives and false negatives in the context of benthic NIS detection. Another way to circumvent filtering issues would be to increase the proportion of reads associated to the target species. Most filtering procedures indeed rely on the fact that errors are in low abundance. In most studies, ours included, using universal primers (Cowart et al. 2015; Grey et al. 2018; Collins et al. 2019), the vast majority of either 18S or COI reads were not assigned to metazoans (in this study, 77% and 76%, respectively). As a consequence, the DNA from many biofoulers was likely not amplified in the first place. For example, each species detected within the four targeted classes represented less than 1% of read abundances, except for *Ciona intestinalis* with 18S (2.9%; Table S9). More effort should thus be directed toward the development of new primer sets for more restricted groups of organisms that could be multiplexed to increase the taxonomic range (Zhang et al. 2018; Corse et al. 2019; West et al. 2020).

Erroneous species assignment challenges the opportunity to detect novel introduced species

One expected outcome of water-eDNA-metabarcoding is the detection of species new to a given area, either introduced or expanding their current distribution range, and not actively looked for (ie passive surveillance). Over the four targeted classes, 18 species yet unreported in the study area were recovered (Table S9). Most were, however, likely the result of erroneous assignments. For example, the ascidian *Clavelina meridionalis*, known from the southern hemisphere and yet unreported in Europe, was identified by 18S only, with 99.16 % identity. No reference sequence, however, was available for the closely-related native *C. lepadiformis*, observed in the field and identified with COI (100% identity) that reliably distinguishes these species (Pérez-Portela and Turon 2008). In total, eight species were identified with 18S only, including six showing identity below 100%. Considering the low taxonomic resolution of 18S, and that most of them have native congeners in the study area, these eight species were not considered as novel introduced species, in the absence of other evidence. Three other species were assigned with 100% identity with COI, and might thus be

true identifications, yet still not necessarily novel introduced species. Indeed, the tunicate *Polycarpa tenera* might have been previously misreported in the study area as the Mediterranean species *P. gracilis* (Monniot 1974; Vazquez et al. 1995). Similarly, the gastropod *Cerithiopsis petanii*, described from the Croatian coast (Prkić and Mariottini 2009), belongs to the *C. tubercularis* species complex (Modica et al. 2013), reported in the study area. It could have thus been overlooked by previous surveys. The only likely new record would then be *Haminoea orteai*, reported in the Mediterranean Sea and the southern part of the Northeast Atlantic (Garabedian et al. 2017).

Knowing the evolutionary history of the targeted species is also important to avoid false positives. In this study's dataset, ASVs from both mitochondrial markers were assigned to *Mytilus trossulus* although very similar to *M. edulis* references (only one gap). The reference sequences for *M. trossulus* corresponded to individuals from the Baltic Sea, a well-known hybrid zone between *M. trossulus* and *M. edulis*, where the mitochondrial genome of *M. trossulus* has been fully introgressed by the mitochondrial genome of *M. edulis* (Kijewski et al. 2006). Aligning the COI ASVs with sequences of *M. trossulus* from Scotland, which have retained their ancestral mitochondrial genome (Zbawicka et al. 2010), led to 83.4% identity only, confirming they were not *M. trossulus* but *M. edulis*.

These case studies demonstrate that passive detection of novel NIS by using water-eDNA-metabarcoding could lead to false positives, as pointed by Darling et al. (2020). Therefore, results need to be confirmed using alternative methods, either traditional and/or more sensitive and targeted molecular approaches such as individual barcoding, qPCR or digital PCR (Borrell et al. 2018; Harper et al. 2018; Wood et al. 2019; Ardura et al. 2021).

Benefitting from metabarcoding of water samples to monitor a wider range of habitats

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While water-eDNA-metabarcoding did not recover all species observed, it detected many more taxa by encompassing a wider range of environments. Marinas are composed of diverse micro-habitats, such as pillars, seawalls, floating pontoons, and soft-sediment bottoms, displaying different faunal assemblages and NIS contributions (Leclerc et al. 2020), which all contribute to the local pool of eDNA. Traditional standardized surveillance protocols (eg Hewitt and Martin 2001; HELCOM and OSPAR 2013) usually target water, sediment, and various hard substrates to encompass the whole diversity of marinas, but with important time and labor costs. This study showed that water-eDNA-metabarcoding can provide data over a wide range of habitats. Considering the 96 expected species identified with water-eDNA-metabarcoding over the four target classes (Table S9), only 35 were biofoulers, four might eventually live fixed (occasionally producing a byssus; Aequipecten opercularis, Limaria hians, Mimachlamys varia, and Venerupis corrugata), 28 were softbottom species, and 30 were motile gastropods. Focusing on NIS, the bivalves Mya arenaria and Ruditapes philippinarum, known from the study area, were recovered by water-eDNAmetabarcoding, while neither targeted nor expected in quadrats because of their burrowing lifestyle.

A wide variety of management strategies exists, depending on the targeted species but also on the phase of the introduction process this species has reached (Robertson et al. 2020). These include measures on transport vectors to limit new arrivals, containment measures to avoid the spread of already established NIS, or suppression or eradication measures to limit the impact of introduced NIS. Whatever the management strategy implemented, however, supporting activities (*sensu* Robertson et al. 2020) are required. In particular there is a dire need to identify NIS and monitor them over time and space, either to report new arrivals or to

investigate their spread and putative impact of reported ones. According to the results, watereDNA-metabarcoding could be a perfect tool for monitoring species in this context. The wide variety of targeted habitats and the ease of sampling allow both for the investigation of NIS spread and putative impact on the communities, as well as allowing for regular sampling.

As a conclusion, the fact that two thirds of NIS present at the time of water sampling were effectively recovered by water-eDNA-metabarcoding supports that this technique is a valuable fast and cost-efficient monitoring method, although not yet ready to be used as a unique tool for marine NIS passive surveillance programs. This results also show that species detection errors (false positives or false negatives) can be limited thanks to the use of multiple markers, the design of suitable primer sets (as performed in Westfall et al. 2020), an increased sampling effort, and the building of reliable curated reference databases, including databases specific to NIS (Darling et al. 2017).

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Disclosure statement

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The authors report that there are no competing interests to declare.

Data availability statement

Raw sequences were deposited on NCBI SRA: BioProject PRJNA744352. The metabarcoding and analyses scripts used in this study are available at https://github.com/joarwrie/benthic_NIS_detection

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Tables

Table 1 Number and proportion (in parentheses) of ASVs and reads assigned to one of the reference sequences from the restricted database targeting ten metazoan phyla, for each marker (18S, COI, and 16S). The number indicated for 'Taxa' includes assignments to species-, genus- and family-level. The last three columns indicate the number of species, genera and families detected.

	ASVs	Reads	Taxa	Species	Genus	Family
18S	389 (9.8%)	1,881,029 (15.8%)	291	203	233	159
CO	943 (22.1%)	706,023 (7.0%)	286	235	221	177
I						
16S	173 (49.9%)	7,311,390 (54.6%)	104	65	78	70

Table 2 List of non-indigenous (NIS) and cryptogenic (Crypto) species within the four target classes, known from the study area, and either observed in quadrats or identified in water-eDNA samples. Stars indicate that, even if the species was not identified in the dataset, sequences were assigned either to the genus because of a lack of resolution, or to another species of the same genus. When applicable, the absence of reference sequence is indicated.

Class	Species	Status	Observed	Detected in eDNA samples		
			in quadrats	18S	COI	16S
Ascidiacea	Asterocarpa humilis	NIS	yes	yes	no	no
Ascidiacea	Botrylloides diegensis	NIS	yes	yes	yes	no
Ascidiacea	Botrylloides violaceus	NIS	yes	no*	no*	no
Ascidiacea	Botryllus schlosseri	Crypto	yes	yes	no	no
Ascidiacea	Ciona robusta	NIS	yes	yes	yes	no
Ascidiacea	Corella eumyota	NIS	yes	yes	no	no
Ascidiacea	Didemnum vexillum	NIS	yes	no	no	no
Ascidiacea	Diplosoma listerianum	Crypto	yes	no	yes	no
Ascidiacea	Perophora japonica	NIS	yes	yes	no	no
Ascidiacea	Styela clava	NIS	yes	yes	no	no
Bivalvia	Crassostrea gigas	NIS	yes	no*	no	no
Bivalvia	Mya arenaria	NIS	no	yes	no	no*
Bivalvia	Ruditapes philippinarum	NIS	no	yes	yes	yes
Gastropoda	Crepidula fornicata	NIS	yes	no	yes	no
Gymnolaemata	Bugula neritina	NIS	yes	yes	yes	yes
Gymnolaemata	Bugulina fulva	Crypto	yes	no	yes	yes
Gymnolaemata	Bugulina simplex	NIS	yes	no	no*	no*
Gymnolaemata	Bugulina stolonifera	NIS	yes	no	no*	no*
Gymnolaemata	Tricellaria inopinata	NIS	yes	no reference	no	no reference
Gymnolaemata	Watersipora subatra	NIS	yes	yes	yes	no reference *

818 Figures with caption

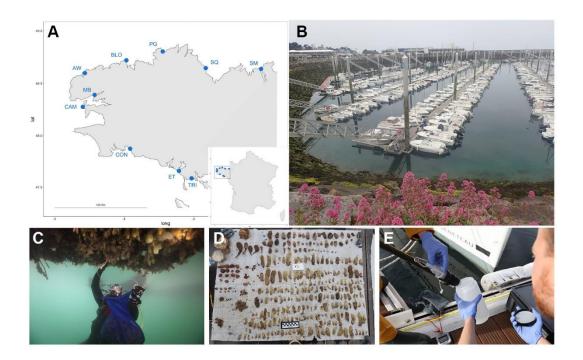


Figure 1 A. Location of the sampling sites. SM = Saint-Malo, SQ = Saint-Quay-Portrieux, PG = Perros-Guirec, BLO = Bloscon-Roscoff, AW = L'Aber Wrac'h, MB = Moulin Blanc-Brest, CAM = Camaret-sur-Mer, CON = Concarneau, ET = Étel, TRI = La Trinité-sur-Mer. **B.** Picture of one of the study marina (SQ), showing sampled pontoons. **C.** Sampling method (ie, diving under the pontoons) for morphological identification. **D.** Picture showing all the specimens scraped in a quadrat after sorting on site. **E.** Seawater sampling method, with seawater collected in a plastic jar using a sterile 100-mL syringe fixed to a sampling rod. Photo credits: Frédérique Viard (B), Wilfried Thomas (C), Laurent Lévêque (D), and Yann Fontana (E).

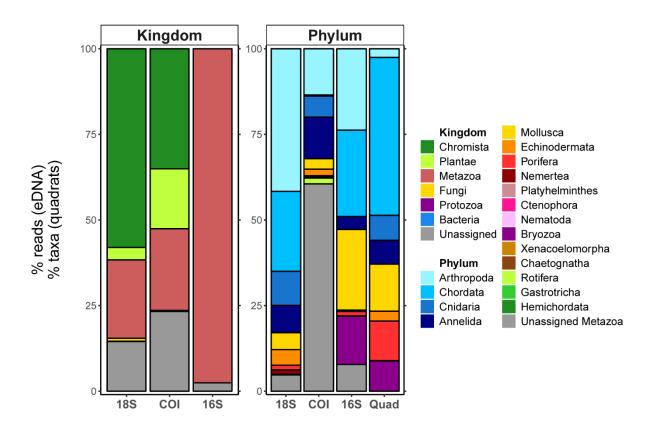


Figure 2 Percentage of reads assigned to different kingdoms (left) or metazoan phyla (right), over all samples, for each of the three markers, as well as for quadrats data (Quad).

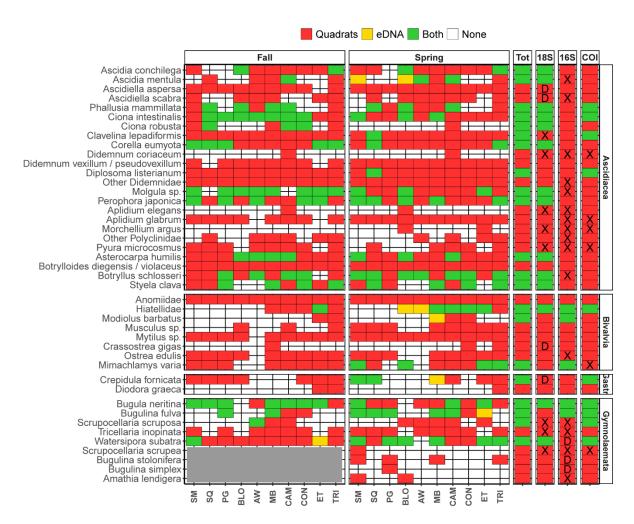


Figure 3 Comparison between water-eDNA-metabarcoding and morphology-based detection. Only the taxa observed within quadrats and belonging to the four targeted classes are considered. Taxa found with both methods within the same marina are displayed in green, those found only in quadrats are in red, and those only found in water-eDNA are in yellow. The first two panels correspond to samples from fall 2017 and spring 2018, respectively. Metabarcoding detection is recorded across all markers. The column Tot represents the presence of each taxon in the whole dataset. The last three columns represent the presence of a taxon in the whole dataset (all marinas and dates combined) for each marker separately. Black crosses indicate the lack of reference sequences for a particular taxon and D indicates that this marker was not able to discriminate a particular taxon. The grey area represents species that were not isolated during morphological identification in fall 2017. They were all pooled in a unique category called 'erected bryozoans'. See Fig. 1 for location codes.

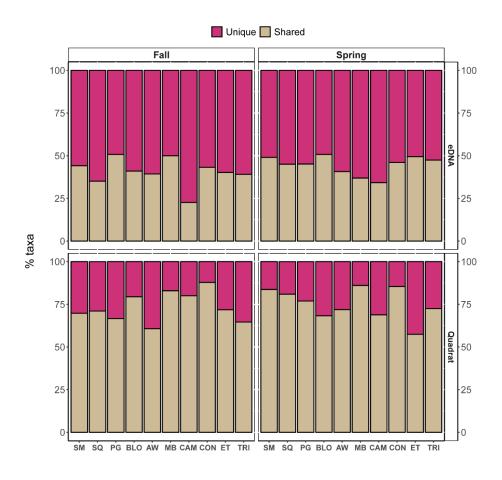


Figure 4 Proportion of taxa recovered with water-eDNA-metabarcoding across all markers or within quadrats, in samples from the two pontoons of a marina for a given season (shared) or in samples from only one pontoon (unique). See Fig. 1 for location codes.

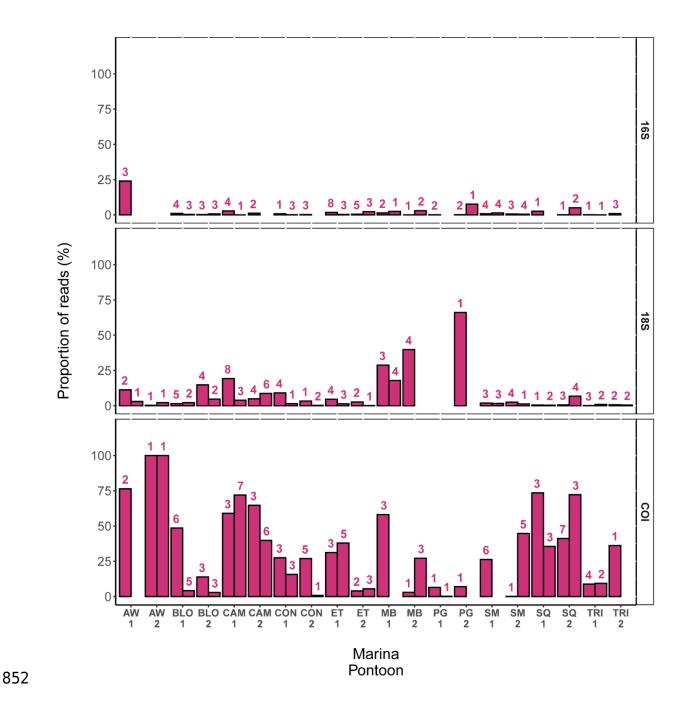


Figure 5 Proportion of reads associated to taxa found in only one of the two 2-L replicates (unique taxa) for each pontoon of each marina. This proportion is calculated by dividing the sum of reads for unique taxa by the sum of all reads for the four target groups in each replicate. The results for each replicate are shown separately (one bar per replicate). The number of unique taxa is indicated at the top of each bar. See Fig. 1 for location codes. Results are presented separately for each marker. Detailed and additional data (values per taxa unique to one replicate and for taxa shared between pontoons) are provided in Fig. S6.