

### **Data Paper**

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# 18S rRNA amplicon sequence data (V1–V3) of the Bronx river estuary, New York

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## **Abstract**

Characterising and monitoring biological diversity to foster sustainable ecosystems is highly recommended as urban centres rapidly expand. However, much of New York City's biodiversity remains undescribed, including in the historically degraded, but recovering Bronx River Estuary. In a pilot study to identify organisms and characterise biodiversity patterns there, 18S rRNA gene amplicons (V1–V3 region), obtained from river sediments and surface waters of Hunts Point Riverside and Soundview Parks, were sequenced. Across 48 environmental samples collected over three seasons in 2015 and 2016, following quality control and contaminant removal, 2,763 Amplicon Sequence Variants (ASVs) were identified from 1,918,463 sequences. Rarefaction analysis showed sufficient sampling depth, and community composition varied over time and by substrate at the study sites over the sampling period. Protists, plants, fungi and animals, including organisms of management concern, such as Eastern oysters (*Crassostrea virginica*), wildlife pathogens and groups related to Harmful Algal Blooms, were detected. The most common taxa identified in river sediments were annelid worms, nematodes and diatoms. In the water column, the most commonly observed organisms were diatoms, algae of the phylum Cryptophyceae, ciliates and dinoflagellates. The presented dataset demonstrates the reach of 18S rRNA metabarcoding for characterising biodiversity in an urban estuary.

# **Key Words**

eDNA, metabarcoding, New York City, next-generation sequencing, QIIME2, urban ecology

## Introduction

With the extensive modification of ecosystems by people and increasing urbanisation, the disrupted ecology of cities is garnering substantial research interest (Alberti 2008). Understanding how urban life forms are responding to habitat alteration, invasion by non-native species, pollution, human population growth, overexploitation, disease, climate change and/or interactions amongst these threats are key areas of urban ecology research. New York City is one of the world's great metropolises, yet much of its biodiversity remains to be identified and described, especially the microfauna and microflora of its rivers and estuaries.

The once highly polluted, 23 mile-long (ca. 37 km), Bronx River is currently considered "impaired", with pollutants including faecal coliforms, garbage, refuse, polychlorinated biphenyls (PCBs) and other toxins, coming from a combination of urban and stormwater runoff, Combined Sewer Overflow (CSO) outfalls, contaminated sediments, and other sources (BRA 2021; NYSDEC 2020). These waters have been impacted by invasive species, such as green (*Carcinus maenas*) and Asian shore (*Hemigrapsus sanguineus*) crabs. They are also affected by harmful algal blooms (HABs), including *Gymnodinium* dinoflagellates (Fuss and O'Neill 2015). Eukaryotic pathogens include *Cryptosporidium parvum* 



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(an intestinal parasite that can affect humans), and the oyster pathogens *Perkinsus marinus* and MSX (*Haplosporidium nelsoni*). However, the recovering area now hosts several clean-up and revitalisation programmes, such as targeted restoration of American eels (*Anguilla rostrata*), river herring (*Alosa pseudoharengus*, *A. aestivalis*) and eastern oysters (*Crassostrea virginica*). In the Bronx River Estuary (Fig. 1), Hunts Point Riverside Park used to be an illegal garbage disposal site, but is now an integral part of the Bronx River Greenway (Kimmelman 2012). Soundview Park waters contain a successfully-restored oyster reef identified as a key research site (Grizzle et al. 2012).

Although the first step in any ecological study is the correct identification of organisms in the focal system, there are numerous challenges in conducting biodiversity inventories (Bik et al. 2012; Bohmann et al. 2014; Taberlet et al. 2018; Deiner et al. 2021). Traditional surveys involving morphological classification and other techniques provide essential data, including abundance information. However, many taxa remain difficult to detect and describe due to a combination of their microscopic or cryptic natures, collection logistics, lack of taxonomic expertise and labour-intensive morphological assessments, leading to uncertain species identifications and biodiversity estimates. Rapid, effective and standardised approaches are needed to guide more detailed investments and cost-effectively complement morphological data for comprehensive management through improved biodiversity information (Bik et al. 2012; Bohmann et al. 2014; Rees et al. 2014; Taberlet et al. 2018; Deiner et al. 2021).

Recently developed, non-invasive and transformative environmental DNA (eDNA) metabarcoding technology provides volumes of data for biodiversity assessment via next-generation sequencing. DNA barcoding was the first worldwide effort to document life and identify species using genetic sequences from a standard DNA segment (Hebert et al. 2003). Approaches such as metabarcoding, a high-throughput extension of DNA barcoding, offer orders of magnitude more data than traditional morphological or DNA barcoding research. Multiple organisms can be identified simultaneously from genetic material extracted from environmental samples (e.g. water, air and sediment) by sequencing and analysing specific marker genes using primers that target their conserved flanking regions (Taberlet et al. 2018; Deiner et al. 2021).

Environmental DNA has numerous advantages, offers high-throughput presence, absence and relative abundance data, and can improve representation of microscopic or cryptic taxa (Bik et al. 2012; Bohmann et al. 2014; Taberlet et al. 2018; Deiner et al. 2021). Environmental DNA metabarcoding is low-impact, efficient, cost-effective, rapid and replicable. The method has been effectively used in estuaries (Chariton et al. 2010, 2015; Leray and Knowlton 2015; Taberlet et al. 2018; Afzali et al. 2021; Carraro et al. 2021; García-Machado et al. 2021) and has found previously undetected or poorly characterised organisms, in particular bacterial, protist and invertebrate

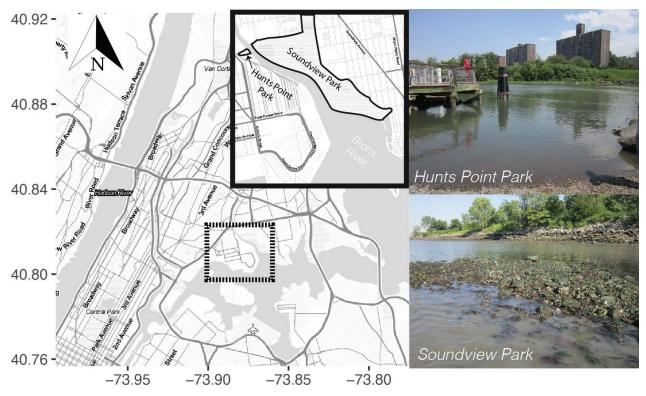
taxa (Bik et al. 2012; Bohmann et al. 2014; Goldberg et al. 2015; Taberlet et al. 2018; Deiner et al. 2021; Leese et al. 2021). Laboratory, computational and data storage limitations exist and reference data for taxonomic assignment of many groups are lacking, but the methods are continuously improving (Valentini et al. 2016; Taberlet et al. 2018; Deiner et al. 2021).

One advancement that facilitates biodiversity assessment and monitoring is state-of-the-art bioinformatics pipeline development to perform quality-control and large-volume data analysis (Taberlet et al. 2018), such as for the 18S rRNA amplicon datasets presented here. Some of the advantages offered by 18S rRNA metabarcoding are broad amplification across eukaryotic kingdoms, a rapidly growing reference database due to wide marker use, as well as conservation within and divergence amongst species genetic profiles (Leray and Knowlton 2016; Taberlet et al. 2018). This marker was selected here to provide a broad overview of estuarine eukaryotic biodiversity, including microorganisms, other algae and invertebrates, that would mirror our prior 16S rRNA metabarcoding work on prokaryotes (Naro-Maciel et al. 2020). Within the 18S rRNA gene, several markers for metabarcoding are being used. For protist taxa, the V4 and the V9 regions are utilised especially often (Stoeck et al. 2009; Dunthorn et al. 2012; de Vargas et al. 2015; Boenigk et al. 2018). Here, the V1-V3 region was targeted due to the high phylogenetic resolution available using hypervariable segments V2-V4, previously demonstrated in dinoflagellates (Ki 2012) and copepods (Wu et al. 2015). Initial checks against published databases and preliminary laboratory tests supported our choice of the V1-V3 region for common taxa or those of management concern, including Eastern oysters (Crassostrea virginica) and HAB-related taxa. Thus, the 18S rRNA dataset, presented here, was used to identify organisms, explore biodiversity patterns and establish a baseline for future work in the Bronx River Estuary.

## Methods

#### Study sites and sampling

The Bronx River Estuary was sampled from August 2015 to September 2016, monthly from May to October during low tide (Fig. 1). Samples were collected from Reach 1 (NYCParks 2021) at Hunts Point (HP, 40.82°N, 73.88°W,  $n_{\text{sediment}} = 9$ ;  $n_{\text{water}} = 8$ ) and Soundview (SVP, 40.81°N, 73.87°W). At Soundview, samples were obtained along a restored oyster reef (SVP-BRO:  $n_{\text{sediment}} = 8$ ;  $n_{\text{water}} = 7$ ) and at another estuarine site about one tenth of a mile (ca. 0.16 km) away where wild oysters were observed (SVP-BRC:  $n_{\text{sediment}} = 8$ ;  $n_{\text{water}} = 8$ ). To investigate two key habitats of estuarine organisms and complement ongoing conventional surveys (NYCParks 2021; Fitzgerald 2013), surface waters and benthic sediments were sampled. The former were sampled by dipping a 1-litre autoclaved jar



**Figure 1.** Sampling site map depicting greater New York City waterways. Inset: Detail map of sample area showing Soundview Park (SVP) and Hunts Point Riverside Park (HP) in bold outline. Right: site photos of the SVP and HP estuaries on the Bronx River. Map data 2019 (C) Google.

horizontally into the river before sediment core collection, in order to avoid contamination (Naro-Maciel et al. 2020). A polyvinyl chloride (PVC) pipe (6-inch length, 2-inch diameter) and a pallet shovel were used to sample river sediments (about 100 g) from the surface sediment layer following standard procedures, including use of disposable gloves and individual, sterilised material for each sample (Fitzgerald 2013). The samples were stored in a cooler and then moved to a laboratory refrigerator (Naro-Maciel et al. 2020).

#### **Environmental DNA analysis**

All materials were processed within 24 hours of sampling (Naro-Maciel et al. 2020). The water samples (n = 23)were divided equally between two funnels (500 ml) and filtered with 0.45 µm Whatman Cellulose Nitrate Sterile filters (Cytiva, USA) using a standard laboratory vacuum pump (Airtech, USA; Type L-250D-G1). Filters were placed into PowerWater DNA Isolation Kit bead tubes (Qiagen, USA) and DNA was extracted following the manufacturer's instructions. A randomly selected 0.25 g soil subsample of each sediment sample (n = 25) to be used for extraction was placed in 2 ml collection tubes and centrifuged. The sediment was then transferred into PowerSoil bead tubes and extracted as instructed (Qiagen, USA). Following DNA extraction and quantification using a NanoDrop 2000 Spectrophotometer or a Qubit Fluorometer (Thermo Scientific, USA), the samples were stored frozen at -20 °C (Naro-Maciel et al. 2020). No extraction blanks or positive controls were included in this pilot study and the turtle-focused molecular biodiversity research lab was not PCR-free. However, contaminant prevention, disinfection, decontamination and sterilisation procedures, standard for a university molecular lab, were assiduously used (e.g. bleach or alcohol disinfection and surface sterilisation, single-use molecular-grade disposable material utilisation, autoclaving, UV-irradiation of supplies etc.) and state-of-the-art *in silico* quality control including contaminant identification and removal was later carried out as discussed below.

All remaining lab work (amplification, purification and sequencing) was conducted in a commercial laboratory (MRDNA, Molecular Research LP, Shallowater, TX, USA) using previously described industry-standard procedures and controls (MRDNA 2021; Dowd et al. 2008; Naro-Maciel et al. 2020). Preliminary runs with small sample sizes were conducted first to confirm primer amplification efficiency, followed by the full sample sets. Through polymerase chain reaction (PCR), about 563 bp of the 18S rRNA gene V1-V3 region was amplified using primers Euk7F (Medlin et al. 1988) and Euk570R (Weekers et al. 1994) (Euk7F: AACCTGGTT-GATCCTGCCAGT + a unique 8 bp identifier barcode; Euk570R: GCTATTGGAGCTGGAATTAC). PCRs were run in 20 µl volumes using the Qiagen HotStarTaq Plus Master Mix Kit (Qiagen, USA) as follows: 94 °C for 3 minutes, 28 cycles of 94 °C for 30 seconds, 53 °C for 40 seconds and 72 °C for 1 minute and a final elongation step at 72 °C for 5 minutes (MRDNA 2021, Dowd et al. 2008). The number of cycles was determined by initial testing to optimise product detection versus errors from over-amplification.

After 2% agarose gel checks, the uniquely barcoded PCR samples were pooled in equal proportions, based on a combination of electrophoresis-based size and density estimations and DNA concentrations. The pooled samples were then purified with calibrated Ampure XP beads (Agencourt Bioscience, USA); the ratio of beads to PCR products used for purification was  $0.75 \times$  as per Illumina manufacturer guidelines. Next, an Illumina DNA library was created from these purified and pooled PCR products ligated to Illumina adapters using the Illumina TruSeq DNA library preparation protocol. Finally, sequencing was performed on an Illumina MiSeq according to manufacturer's instructions, using paired-end  $2 \times 300$  bp v.3 chemistry (MRDNA 2021, Dowd et al. 2008).

#### Amplicon sequence variants (ASVs)

The FASTQ Processor was used to extract barcodes and sort forward and reverse reads into distinct files (MRDNA 2021). Raw reads were then processed using the QIIME2 v. 2019.1 pipeline (Bolyen et al. 2018) (Suppl. material 1: Document S1). The demultiplexed reads were not merged due to insufficient overlap. As quality statistics were high for forward reads, but more variable for reverse, only the forward reads were analysed as single-end. The DADA2 plug-in was then used in QIIME2 to de-noise and quality-filter the forward sequences, assign amplicon sequence variants (ASVs or features), include only Eukaryotes and generate a feature table of ASVs and metadata (Callahan et al. 2016). Primers and low-quality base calls (5–10 bp) were trimmed at the ends of each single-end sequence during the DADA2 step and reads were truncated at 260 bp, based on examination of quality scores, to account for typically observed end-of-sequence decline in quality. All other parameters, including culling short and otherwise low-quality sequences, identifying and deleting chimeras etc. were run as default (QIIME2 2021). After DADA2 filtering, the average percentage of sequences retained was 79%, with a median of 39,758 sequences kept per sample (Table 1). To taxonomically classify the 18S reads, the q2-Naïve Bayesian classifier, as implemented in QIIME2, was employed, using the SILVA 138 reference database (Quast et al. 2013) trained on the entirety of the 18S rRNA gene (Bokulich et al. 2018; Karst et al. 2018). This reference database was selected because it is a comprehensive, frequently updated and quality-curated resource for identifying eukaryotic 18S rRNA gene sequences. ASV taxonomy was manually inspected to ensure adequate taxonomic resolution was achieved.

Once taxonomic annotation was complete, R Studio v. 1.2.1335 (R Core Team 2008) was used to perform statistical analysis (Suppl. material 1: Document S2). The 'DECONTAM' programme v. 1.8.0 was run on the feature table to remove potential contaminants (Davis et al. 2018). The programme's frequency method option works

**Table 1.** Summary of sample data, including sample ID and statistics on the recovery of reads per sample after filtering, de-noising and chimeric sequence removal.

Sample Input Filtered passed filter % input passed filter De-possed chimeric passed chimeric Non-possed chimeric possed chimeric % of input passed chimeric   S.B.BRC 61653 51307 83.22 49865 49355 80.05   S.B.BRO 57043 46675 81.82 45055 43716 76.64   S.B.HP 41172 34193 83.05 32911 31537 76.60
S.B.BRC 61653 51307 83.22 49865 49355 80.05   S.B.BRO 57043 46675 81.82 45055 43716 76.64
S.B.BRC 61653 51307 83.22 49865 49355 80.05   S.B.BRO 57043 46675 81.82 45055 43716 76.64
<b>S.B.BRO</b> 57043 46675 81.82 45055 43716 76.64
<b>CRUP</b> 41172 24102 92.05 22011 21527 76.60
<b>S.C.BRC</b> 65637 54030 82.32 52074 50699 77.24
<b>S.C.BRO</b> 53923 43911 81.43 42387 41716 77.36
<b>S.C.HP</b> 48441 40413 83.43 38983 37581 77.58
<b>S.D.BRC</b> 37670 31600 83.89 29892 29647 78.70
<b>S.D.BRO</b> 39984 32324 80.84 30829 30329 75.85
<b>S.D.HP</b> 38926 32241 82.83 30923 29822 76.61
<b>S.E.BRC16</b> 81853 67172 82.06 65692 64195 78.43
<b>S.E.BRO16</b> 64498 52646 81.62 51301 49170 76.23
<b>S.E.HP16</b> 50365 41017 81.44 39681 38571 76.58
<b>S.F.BRC16</b> 74188 62748 84.58 61411 60144 81.07
<b>S.F.BRO16</b> 72355 59701 82.51 57927 57472 79.43
<b>S.F.HP16</b> 56688 47187 83.24 46094 44870 79.15
<b>S.G.BRC16</b> 60173 50378 83.72 48470 47934 79.66
<b>S.G.BRO16</b> 59639 50520 84.71 49155 45773 76.75
S.G.HP16 62125 49689 79.98 48429 45891 73.87
S.H.BRC16 84518 62017 73.38 60776 60186 71.21
<b>S.H.BRO16</b> 64609 53593 82.95 51590 50524 78.20
<b>S.H.HP16</b> 48136 40675 84.5 39217 37346 77.58
<b>S.I.BRC16</b> 63314 53030 83.76 51647 51166 80.81
<b>S.I.BRO16</b> 54232 44896 82.79 43223 42157 77.73
<b>S.I.HP16</b> 51530 40901 79.37 39682 37770 73.30
<b>S.J.HP16</b> 55575 44475 80.03 43004 41364 74.43
<b>W.B.BRC</b> 66549 57336 86.16 49282 46759 70.26
<b>W.B.BRO</b> 66937 57216 85.48 49679 47022 70.25
<b>W.B.HP</b> 50568 43407 85.84 37591 35988 71.17
<b>W.D.BRC</b> 17978 15045 83.69 14346 14173 78.84
<b>W.D.BRO</b> 33475 27576 82.38 26316 25681 76.72
<b>W.D.HP</b> 23573 19891 84.38 17931 17902 75.94
W.E.BRC16 39716 34005 85.62 32661 31003 78.06
W.E.BRO16 37689 32271 85.62 30832 29645 78.66
<b>W.E.HP16</b> 39206 33090 84.4 31243 30725 78.37
<b>W.F.BRC16</b> 37895 32172 84.9 30931 30462 80.39
W.F.BRO16 43923 36032 82.03 34993 34027 77.47
<b>W.F.HP16</b> 70651 57008 80.69 56284 52129 73.78
<b>W.G.BRC16</b> 50138 42646 85.06 41774 38491 76.77
<b>W.G.BRO16</b> 47758 40433 84.66 39051 36563 76.56
<b>W.G.HP16</b> 54025 45735 84.66 44776 40909 75.72
W.H.BRC16 41440 35408 85.44 34245 32717 78.95
W.H.BRO16 42214 34834 82.52 33777 32871 77.87
<b>W.H.HP16</b> 55457 46005 82.96 44750 41631 75.07
W.I.BRC16 47134 39762 84.36 37335 35006 74.27
W.I.BRO16 48683 40787 83.78 38592 35516 72.95
W.I.HP16 46537 38277 82.25 37216 35306 75.87
<b>W.J.HP16</b> 41706 34926 83.74 33496 32281 77.40
W.J.SVP16 57638 48829 84.72 47289 42721 74.12
Totals 2509137 1918463

by inferring potential contaminants using a simple inverse linear correlation between initial sample DNA concentration and the frequency of each ASV. Contaminants should behave such that their relative proportion increases as sample concentration decreases (Davis et al. 2018). Using a threshold of p < 0.10, the programme filtered ASVs meeting this criterion from the dataset. In total 28 contaminants were removed from the feature table, representing just 1% of the dataset (Suppl. material 2: Table S1).

The PHYLOSEQ v. 1.28.0 package was then used for basic data manipulation and, visualisation and community-level statistical analyses were performed using tools available in the VEGAN v. 2.5.5 package (McMurdie and Holmes 2013; Oksanen 2019). Observed ASV richness and the Shannon Diversity Index (Shannon

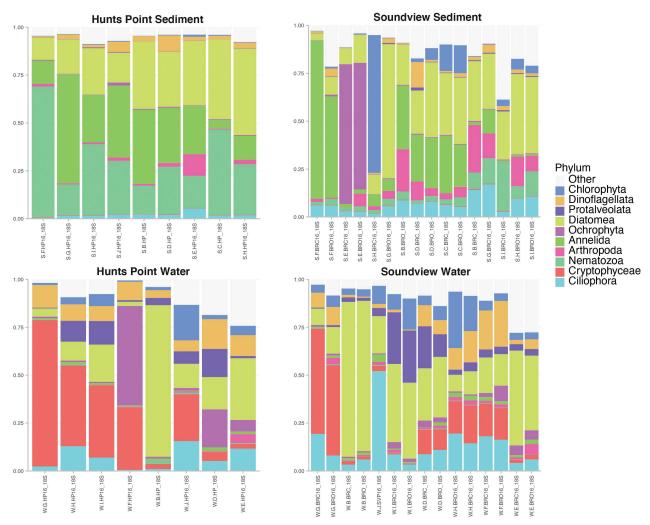


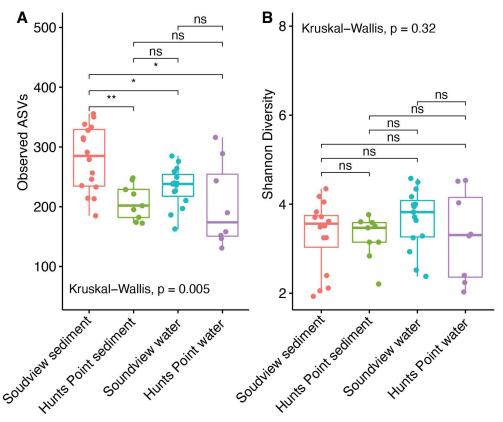
Figure 2. 18S rRNA community profiles of Amplicon Sequence Variants (ASVs) in sediment and water samples from Hunts Point Riverside and Soundview Parks shown at the level of phylum. Bar heights show relative abundance of sequences from each taxon.

1948) were computed to summarise alpha diversity of the eukaryotic communities. Differences in alpha diversity amongst sediments and water from both sites were assessed using a Kruskal-Wallis test with non-parametric pairwise comparisons. For beta diversity comparisons, the data were normalised using a Hellinger Transformation, which takes the square root of each ASV's relative abundance and bounds it between 0 and 1 (Legendre and Gallagher 2001; Lahti et al. 2017). Beta diversity, or turnover between sites, was summarised using the Bray-Curtis Index. To visualise differences in beta diversity, non-metric multidimensional scaling (NMDS) ordination (stress = 0.13), based on Bray-Curtis Dissimilarity was carried out using several random starts and stress assessment through the metaMDS command (Oksanen 2015) and ellipses were drawn using the stat ellipse function.

# Results and discussion

This pilot study identified key organisms, explored biodiversity patterns and established a baseline for future work in the area, but the data must be interpreted with caution considering methodological issues. In total 48 environmental samples were successfully collected and sequenced for the 18S rRNA gene (n<sub>water</sub> = 23; n<sub>sediment</sub> = 25). Within these samples, protists, plants, fungi and animals encompassing 2,763 ASVs were recovered from a total of 1,918,463 post-quality-control sequences (Suppl. material 2: Table S2). Species accumulation curves of each sample reached an asymptote, indicating that the communities were surveyed with sufficient depth to detect robust differences in community structure and composition (Suppl. material 2: Fig. S1). At the study sites over the sampling period, community composition varied over time and by substrate (Fig. 2).

Several organisms of known occurrence, including taxa of management concern, were detected. Commonly observed species identified in this survey included softshell clams (*Mya arenaria*) and blue mussels (*Mytilus edulis*) (Suppl. material 2: Table S1). Oyster DNA (*Crassostrea virginica*) was detected in both Soundview waters and sediments, but not in Hunts Point waters or sediment. Similarly, oysters have been observed at Soundview, but not at Hunts Point; the oyster parasite genus *Perkinsus* was detected only in the water at Soundview Park. The



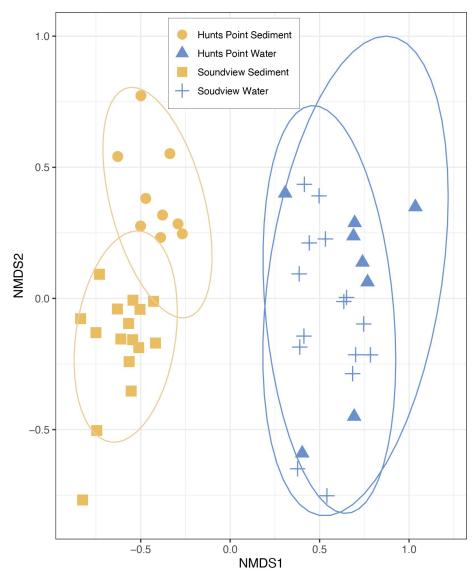
**Figure 3.** Alpha diversity comparison between sediment and water samples from Hunts Point Riverside and Soundview Parks computed using A) Observed ASVs and B) Shannon Diversity. Pairwise comparisons are indicated as follows: \* = p < 0.05, \*\* = p < 0.001, ns = non-significant.

oyster pathogen MSX (Haplosporidium nelsoni) was not identified at any site, but the crustacean parasite Haliphthoros was found in Hunts Point sediments. Dinophyceae dinoflagellates (Gymnodinium, Heterocapsa, Karlodinium) and Raphidophyceae (Heterosigma akashiwo), which may cause Harmful Algal Blooms (Hara and Chihara 1987; Faust and Gulledge 2002; Millette et al. 2015; Lin et al. 2018), were sequenced from the Bronx River Estuary. Macroinvertebrate taxa considered to be indicators of estuary pollution (Pelletier et al. 2010; Smith et al. 2015) were not commonly found, except for the various small aquatic worms (Nematoda), some of which are consistent with poor water quality (Fig. 2). American eels (Anguilla rostrata) and herring (Alosa pseudoharengus and A. aestivalis), key organisms being restored and monitored in the Bronx River, were not detected.

In terms of alpha diversity, Soundview Park sediments were significantly higher in the observed number of ASVs compared with all other sites (Kruskal-Wallis, p=0.05, Fig. 3A). While Soundview water also trended towards higher biodiversity compared with Hunts Point water, the difference was non-significant. However, none of the sites differed significantly in Shannon Diversity (p>0.05, Fig. 3B). Sediment communities between Hunts Point and Soundview were differentiated by the presence of several key taxa missing or less proportionally abundant at Hunts Point. For example, Soundview sediments had higher proportions of arthropod DNA detected than

those at Hunts Point (Figs 2 and 3). In agreement with our results on overall alpha diversity metrics, Soundview Park sediments were more taxonomically diverse when compared with Hunts Point. Water samples from both sites were not apparently different in taxonomic composition (Figs 2-4). However, there were clear differences between the taxonomic make-up of sediments and water column samples, driven mostly by the more frequent detection of annelid worms and nematodes in sediments and larger proportions of diatoms, dinoflagellates and Protalveolata in water samples (Fig. 2). The community turnover (i.e. beta diversity) of eDNA from water samples was significantly different from that of sediment ( $r^2 = 0.169$ , P = 0.001; Fig. 4). Water samples were homogeneous amongst sites ( $r^2 = 0.069$ , P = 0.834). In contrast, sediment samples from Soundview Park were distinct from those at Hunts Point ( $r^2 = 0.245$ , P = 0.001; Fig. 4).

Future metabarcoding work in the area would benefit from lessons learned during and resources developed since this pilot study. The high quality, comprehensive protocols now available to standardise and ensure eDNA metabarcoding excellence should be carefully followed (Taberlet et al. 2018; Minamoto et al. 2021). To better incorporate unicellular organisms and viruses, specialised methods, including use of filters with finer pore size, should be employed. While state-of-the art bioinformatics work conservatively identified and removed errors and contaminants, ongoing research would additionally



**Figure 4.** Community comparisons amongst substrates (sediment and water) and sites (Hunts Point Riverside and Soundview Parks), based on Amplicon Sequence Variants (ASVs) and using Non-metric multidimensional scaling analysis (NMDS).

benefit from stringent laboratory checks. These include the use of extraction blanks in a PCR-free lab, positive controls to assess amplification efficiency and negative controls to identify contaminants directly. Technical replicates address contamination, errors including rare taxa detection, and PCR amplification and sequencing variation. Further, although the V1 - V3 segment did capture known organisms and those of management interest, reference data bases for the increasingly used V4 or V9 regions may be more complete, thus resulting in more identifications. Finally, organismal abundance does not necessarily correlate with sequence abundance given amplification biases and errors such as primer-template fidelity and suboptimal annealing temperatures. Thus even inferences about relative abundance should be interpreted with caution (Fonseca 2018; Taberlet et al. 2018). Comparing metabarcoding results to conventional survey data will continue to be essential for ground-truthing and optimising both methods (Fediajevaite et al. 2021).

In conclusion, the 18S rRNA V1 – V3 dataset, presented here, complements our prior study, "16S rRNA Amplicon Sequencing of Urban Prokaryotic Communities in the South Bronx River Estuary". Future work will comparatively analyse information from these two genetic regions and new data from Cytochrome Oxidase I, the standard locus for animal barcoding (Hebert et al. 2003). Despite its advantages, the 18S rRNA marker alone is insufficient to fully characterise biodiversity and a suite of markers would provide a more complete profile (Leray and Knowlton 2016; Taberlet et al. 2018) to further describe life in a complex urban estuary.

# Data availability

All 18S rRNA amplicon gene sequences from this study are posted on the NCBI Sequence Read Archive (SRA) under BioProject PRJNA606795 accession numbers

SAMN19729835–SAMN19729882 (Table 1). All DNA extracts are stored at the American Museum of Natural History. Bioinformatics and statistical scripts are available as a supplement to this article (Suppl. material 1: Documents S1, S2).

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# **Conflicts of Interest**

The authors declare no competing interests.

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#### Supplementary material 1 Scripts used for metabarcoding analysis

Author: Melissa Ingala

Data type: QIIME and R Scripts (in zip. archive)

Explanation note: **Document S1.** QIIME2 workflow. **Document S2.** R script.

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# Supplementary material 2 Figure S1, Tables S1, S2

Author: Melissa Ingala

Data type: jpg. file, docx. file and xslx. file (in zip. archive)

Explanation note: **Figure S1.** Sample-based species accumulation curves of total 18S rRNA diversity by substrate type (sediment, water) for each site (Hunts Point Riverside and Soundview Parks), calculated using the Vegan 2.4-3 package for Amplicon Sequence Variants (ASVs). **Table S1.** Taxonomies of contaminating ASVs removed by "decontam" analysis. **Table S2.** Taxonomic Assignment including ASV ID to Domain (d), Phylum (p), Class (c), Order (o), Family (f), Genus (g) and Species (s), as applicable.

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