

1 **A workflow for accurate metabarcoding using nanopore MinION sequencing**

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15 Amplification

16

17 **Abstract**

18

19 Metabarcoding has become a common approach to the rapid identification of the species
20 composition in a mixed sample. The majority of studies use established short-read high-throughput
21 sequencing platforms. The Oxford Nanopore MinION™, a portable sequencing platform, represents a
22 low-cost alternative allowing researchers to generate sequence data in the field. However, a major
23 drawback is the high raw read error rate that can range from 10% to 22%.

24 To test if the MinION™ represents a viable alternative to other sequencing platforms we used
25 rolling circle amplification (RCA) to generate full-length consensus DNA barcodes (658bp of
26 cytochrome oxidase I - COI) for a bulk mock sample of 50 aquatic invertebrate species. By applying
27 two different laboratory protocols, we generated two MinION™ runs that were used to build
28 consensus sequences. We also developed a novel Python pipeline, ASHURE, for processing,
29 consensus building, clustering, and taxonomic assignment of the resulting reads.

30 We were able to show that it is possible to reduce error rates to a median accuracy of up to 99.3%
31 for long RCA fragments (>45 barcodes). Our pipeline successfully identified all 50 species in the
32 mock community and exhibited comparable sensitivity and accuracy to MiSeq. The use of RCA was
33 integral for increasing consensus accuracy, but it was also the most time-consuming step during the
34 laboratory workflow and most RCA reads were skewed towards a shorter read length range with a
35 median RCA fragment length of up to 1262bp. Our study demonstrates that Nanopore sequencing can
36 be used for metabarcoding but we recommend the exploration of other isothermal amplification
37 procedures to improve consensus length.

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39

40 **Introduction**

41

42 DNA metabarcoding uses high-throughput sequencing (HTS) of DNA barcodes to quantify the
43 species composition of a heterogeneous bulk sample. It has gained importance in fields such as
44 evolutionary ecology (Lim et al. 2016), food safety (Staats et al. 2016), disease surveillance (Batovska
45 et al. 2018), and pest identification (Sow et al. 2019). Most metabarcoding studies to date have used
46 short-read platforms such as the Illumina MiSeq (Piper et al. 2019). New long-read instruments such
47 as the Pacific Biosciences Sequel platform could improve taxonomic resolution (Tedersoo et al. 2017;
48 Heeger et al. 2018) through long high-fidelity DNA barcodes. Long read nanopore devices are
49 becoming increasingly popular because these devices are low-cost and portable (Menegon et al. 2017).

50 Nanopore sequencing is based on the readout of ion current changes occurring when single-stranded
51 DNA passes through a protein pore such as alpha-hemolysin (Deamer et al. 2016). Each nucleotide
52 restricts ion flow through the pore by a different amount, enabling base-calling via time series analysis

53 of the voltage across a nanopore. (Clarke et al. 2009). The first commercially available instrument,
54 Oxford Nanopore Technologies' MinION™, is a portable, low-cost sequencing platform that can
55 produce long reads (10 kb to 2 Mb reported; Nicholls et al. 2019). The low capital investment costs
56 (starting at \$1,000 US) have made this device increasingly popular among scientists working on
57 molecular species identification (Parker et al. 2017, Kafetzopoulou et al. 2018, Loit et al. 2019),
58 disease surveillance (Quick et al. 2016), and whole-genome reconstruction (Loman et al. 2015).
59 However, a major drawback is the high raw read error rate which reportedly ranges from 10-22% (Jain
60 et al. 2015, Sović et al. 2016, Jain et al. 2018, Kono and Arakawa, 2019, Krehenwinkel et al. 2019), a
61 concern when investigating the within-species diversity or the diversity of closely related species.

62 However, with consensus sequencing strategies, nanopore instruments can also generate high
63 fidelity reads for shorter amplicons (Simpson et al. 2017, Pomerantz et al. 2018, Rang et al. 2018).
64 Clustering of corresponding reads is accomplished by using a priori information such as reference
65 genomes (Vaser et al. 2017), primer indices marking each sample (Srivathsan et al. 2018), or spatially
66 related sequence information, which can be encoded using DNA amplification protocols such as loop-
67 mediated isothermal amplification (LAMP) (Mori & Notomi, 2009) or rolling circle amplification
68 (RCA) (McNaughton et al. 2019). RCA is based on the circular replication of single-stranded DNA
69 molecules. A series of such replicated sequences can be used to build consensus sequences with an
70 accuracy of up to 99.5% (Li et al. 2016, Calus et al. 2017, Volden et al. 2018).

71 The combination of metabarcoding and nanopore sequencing could allow researchers to generate
72 barcode sequence data for community samples in the field, without the need to transport or ship
73 samples to a laboratory. So far only a small number of studies have demonstrated the suitability of
74 MinION™ for metabarcoding using samples of very low complexity, e.g., comprising of three
75 (Batovska et al. 2018), 6 -11 (Voorhuijzen-Harink et al. 2019), or nine species (Krehenwinkel et al.
76 2019).

77 For this study we used a modified RCA protocol (Li et al. 2016) for nanopore consensus sequencing
78 of full-length DNA barcodes (658bp of cytochrome oxidase I - COI) from a bulk sample of 50 aquatic
79 invertebrate species to explore the feasibility of nanopore sequencing for metabarcoding. We also
80 developed a new Python pipeline to explore error profiles of nanopore consensus sequences, mapping
81 accuracy, and overall community representation of a complex bulk sample.

82

83 **Methods**

84

85 Mock community preparation

86 We constructed a mock community of 50 freshwater invertebrate specimens collected with kick-nets in
87 Southern Ontario and Germany. Collection details are recorded in the public dataset DS-NP50M on
88 Barcode of Life Data Systems (BOLD, <http://www.boldsystems.org>, see Ratnasingham & Hebert

89 2007). A small piece of tissue was subsampled from each specimen (Arthropoda: a leg or a section of a
90 leg; Annelida: a small section of the body; Mollusca: a piece of the mantle) and the DNA was
91 extracted in 96-well plates using membrane-based protocols (Ivanova et al. 2006, Ivanova et al. 2008).
92 The 658 bp barcode region of COI was amplified using the following thermal conditions: initial
93 denaturation at 94°C for 2 min followed by 5 cycles of denaturation for 40 s at 94°C, annealing for 40
94 s at 45°C and extension for 1 min at 72°C; then 35 cycles of denaturation for 40 s at 94°C with
95 annealing for 40 s at 51°C and extension for 1 min at 72°C; and a final extension for 5 min at 72°C
96 (Ivanova et al. 2006). The 12.5 µl PCR reaction mixes included 6.25 µl of 10% trehalose, 2.00 µl of
97 ultrapure water, 1.25 µl 10X PCR buffer [200 mM Tris-HCl (pH 8.4), 500 mM KCl], 0.625 µl MgCl
98 (50 mM), 0.125 µl of each primer cocktail (0.01 mM, C_LepFolF/C_LepFolR (Hernández-Triana et al.
99 2014) and for Mollusca C_GasF1_t1/GasR1_t1 (Steinke et al. 2016)), 0.062 µl of each dNTP (10
100 mM), 0.060 µl of Platinum® Taq Polymerase (Invitrogen), and 2.0 µl of DNA template. PCR
101 amplicons were visualized on a 1.2% agarose gel E-Gel® (Invitrogen) and bidirectionally sequenced
102 using sequencing primers M13F or M13R and the BigDye® Terminator v.3.1 Cycle Sequencing Kit
103 (Applied Biosystems, Inc.) on an ABI 3730xl capillary sequencer following manufacturer's
104 instructions. Bi-directional sequences were assembled and edited using Geneious 11 (Biomatters). For
105 specimens without a species-level identification, we employed the Barcode Index Number (BIN)
106 system that assigns each specimen to a species proxy using the patterns of sequence variation at COI
107 (Ratnasingham & Hebert, 2013). With this approach, we selected a total of 50 OTUs with 15% or
108 more K2P COI distance (Kimura, 1980) from other sequences for the mock sample. A complete list of
109 specimens, including taxonomy, collection details, sequences, BOLD accession numbers, and Nearest
110 Neighbour distances are provided in Supplementary Table S1.

111

112 Bulk DNA extraction

113 The remaining tissue of the mock community specimens was dried overnight, pooled, and
114 subsequently placed in sterile 20mL tubes containing 10 steel beads (5mm diameter) to be
115 homogenized by grinding at 4000 rpm for 30-90 min in an IKA ULTRA TURRAX Tube Drive
116 Control System (IKA Works, Burlington, ON, Canada). A total of 22.1 mg of homogenized tissue was
117 used for DNA extraction with the Qiagen DNeasy Blood and Tissue kit (Qiagen, Toronto, ON,
118 Canada) following the manufacturer's instructions. DNA extraction success was verified on a 1%
119 agarose gel (100 V, 30 min) and DNA concentration was quantified using the Qubit HS DNA Kit
120 (Thermo Fisher Scientific, Burlington, ON, Canada).

121

122 Metabarcoding using Illumina Sequencing

123 For reference, we used a common metabarcoding approach with a fusion primer-based two-step PCR
124 protocol (Elbrecht & Steinke 2019). During the first PCR step, a 421 bp region of the Cytochrome c

125 oxidase subunit I (COI) was amplified using the BF2/BR2 primer set (Elbrecht & Leese 2017). PCR
126 reactions were carried out in a 25 μ L reaction volume, with 0.5 μ L DNA, 0.2 μ M of each primer, 12.5
127 μ L PCR Multiplex Plus buffer (Qiagen, Hilden, Germany). The PCR was carried out in a Veriti
128 thermocycler (Thermo Fisher Scientific, MA, USA) using the following cycling conditions: initial
129 denaturation at 95 °C for 5 min; 25 cycles of: 30 sec at 95 °C, 30 sec at 50 °C and 50 sec at 72 °C; and
130 a final extension of 5 min at 72 °C. One μ L of PCR product was used as the template for the second
131 PCR, where Illumina sequencing adapters were added using individually tagged fusion primers
132 (Elbrecht & Steinke 2019). For the second PCR, the reaction volume was increased to 35 μ L, the cycle
133 number reduced to 20, and extension times increased to 2 minutes per cycle. PCR products were
134 purified and normalized using SequalPrep Normalization Plates (Thermo Fisher Scientific, MA, USA,
135 Harris et al. 2010) according to manufacturer protocols. Ten μ L of each normalized sample was
136 pooled, and the final library cleaned using left-sided size selection with 0.76x SPRIselect (Beckman
137 Coulter, CA, USA). Sequencing was carried out by the Advances Analysis Facility at the University of
138 Guelph using a 600 cycle Illumina MiSeq Reagent Kit v3 and 5% PhiX spike in. The forward read was
139 sequenced for an additional 16 cycles (316 bp read).

140 The resulting sequence data were processed using the JAMP pipeline v0.67
141 (github.com/VascoElbrecht/JAMP). Sequences were demultiplexed, paired-end reads merged using
142 Usearch v11.0.667 with `fastq_pctid=75` (Edgar 2010), reads below the read length threshold (414bp)
143 were filtered and primer sequences trimmed both by using Cutadapt v1.18 with default settings
144 (Martin 2011). Sequences with poor quality were removed using an expected error value of 1 (Edgar &
145 Flyvbjerg 2015) as implemented in Usearch. MiSeq reads, including singletons, were clustered using
146 `cd-hit-est` (Li & Godzik, 2006) with parameters: `-b 100 -c 0.95 -n 10`. Clusters were subsequently
147 mapped against the mock community data as well as against the BOLD COI reference library.

148

149 Metabarcoding using Nanopore sequencing

150 We used a modified intramolecular-ligated Nanopore Consensus Sequencing (INC-Seq) approach (Li
151 et al. 2016) that employs rolling circle amplification (RCA) of circularized templates to generate linear
152 tandem copies of the template to be sequenced on the nanopore platform. An initial PCR was prepared
153 in 50 μ L reaction volume with 25 μ L 2 \times Multiplex PCR Master Mix Plus (Qiagen, Hilden, Germany),
154 10pmol of each primer (for 658 bp COI barcode fragment – Supplementary Table S2), 19 μ L molecular
155 grade water and 4 μ L DNA. We used a Veriti thermocycler (Thermo Fisher Scientific, MA, USA) and
156 the following cycling conditions: initial denaturation at 98°C for 30 secs, 35 cycles of (98°C for 30
157 secs, 59°C for 30 secs, 72°C for 30 secs), and a final extension at 72°C for 2 min. Amplicons were
158 purified using SpriSelect (Beckman Coulter, CA, USA) with a sample to volume ratio of 0.6x and
159 quantified. Purified amplicons were self-ligated to form plasmid like structures using Blunt/TA Ligase
160 Master Mix (NEB, Whitby, ON, Canada) following manufacturer's instructions. Products were

161 subsequently treated with the Plasmid-Safe™ ATP-dependent DNase kit (Lucigen Corp, Middleton,
162 WI, USA) to remove remaining linear molecules. Final products were again purified with SpriSelect at
163 a 0.6x ratio and quantified using the High Sensitivity dsDNA Kit on a Qubit fluorometer (Thermo
164 Fisher Scientific, MA, USA). Rolling Circle Amplification (RCA) was performed for six 2.5 µL
165 aliquots of circularized DNA plus negative controls (water) using the TruePrime™ RCA kit
166 (Expedeon Corp, San Diego, CA, USA) following manufacturer's instructions. After initial
167 denaturation at 95°C for three minutes, RCA products were incubated for 2.5 to 6 hours at 30°C. The
168 DNA concentration was measured after every hour. RCA was stopped once 60-70 ng/ul of double-
169 stranded DNA was reached. Subsequently, RCA products were incubated for 10 min at 65°C to
170 inactivate the enzyme. We performed two experiments under varying RCA conditions (Protocol A and
171 B, detailed in Table 1), such as RCA duration (influences number of RCA fragments), fragmentation
172 duration, and fragmentation methods.

173 Protocol A followed Li et al. (2016) by incubating 65µL of pooled RCA product with 2µL (20 units) of
174 T7 Endonuclease I (NEB, M0302S, VWR Canada, Mississauga, ON, Canada) at room temperature for
175 10 min of enzymatic debranching, followed by mechanical shearing using a Covaris g-TUBE™ (D-
176 Mark Biosciences, Toronto, ON, Canada) at 4200 rpm for 1 min on each side of the tube or until the
177 entire reaction mix passed through the fragmentation hole. Protocol B is a more modified approach to
178 counteract the overaccumulation of smaller DNA fragments. Here we did only 2 min of enzymatic
179 debranching with no subsequent mechanical fragmentation. To verify the size of fragments after
180 shearing, sheared products for both protocols were run on a 1% agarose gel at 100 V for 1 hour. DNA
181 damage was repaired by incubating 53.5µL of the product with 6.5µL of FFPE DNA Repair Buffer
182 and 2µL of NEBNext FFPE Repair mix (VWR Canada, Mississauga, ON, Canada) at 20°C for 15. The
183 final product was purified using SpriSelect at a 0.45x ratio and quantified using a Qubit fluorometer.

184 For sequencing library preparation, we used the Nanopore Genomic Sequencing Kit SQK-LSK308
185 (Oxford Nanopore, UK). First, the NEBNext Ultra II End Repair/dA Tailing kit (NEB, Whitby, ON,
186 Canada) was used to end repair 1000 ng of sheared genomic DNA (1 microgram of DNA in 50µl
187 nuclease-free water, 7µl of Ultra II End-Prep Buffer, 3µl Ultra II End-Prep Enzyme Mix in a total
188 volume of 60µl). The reaction was incubated at 20°C for 5 min and heat-inactivated at 65°C for
189 another 5 min. Resulting DNA was purified using SpriSelect at a 1:1 ratio according to the SQK-
190 LSK308 protocol. Then it was eluted in 25µl of nuclease-free water and quantified with a recovery aim
191 of >70 ng/µl. Blunt/TA Ligase Master Mix (NEB, Whitby, ON, Canada) was used to ligate native
192 barcode adapters to 22.5µl of 500 ng end-prepared DNA at room temperature (10 min). DNA was
193 purified using a 1:1 volume of SpriSelect beads and eluted in 46µl nuclease-free water before the
194 second adapter ligation. For each step, the DNA concentration was measured. The library was purified
195 with ABB buffer provided in the SQK-LSK308 kit (Oxford Nanopore, Oxford Science Park, UK).

196 The final library was then loaded onto a MinION flow cell FLO-MIN107.1 (R9.5) and sequenced

197 using the corresponding workflow on MinKNOW™. Base-calling was performed using Guppy 3.2.2
198 in CPU mode with the dna_r9.5_450bps_1d2_raw.cfg model.
199 We designed a new Python (v3.7.6) pipeline, termed ASHURE (A safe heuristic under Random
200 Events) to process RCA reads and to build consensus sequences (Suppl Fig 1). Detailed information is
201 available on GitHub: <https://github.com/BBaloglu/ASHURE>. The pipeline uses the OPTICS algorithm
202 (Ankerst et al. 1999) for clustering and t-distributed stochastic neighbor embedding (Maaten & Hinton,
203 2014) for dimensionality reduction and visualization. Sequence alignments were conducted using
204 minimap2 (Li, 2018) and SPOA (Vaser et al. 2017). Correlation coefficients were determined through
205 ASHURE using both the Numpy (van der Walt et al. 2011) and the Pandas package (McKinney 2010).
206 The Pipeline also includes comparisons of consensus error to several parameters, such as RCA length,
207 UMI error, and cluster center error as well as accuracy determination. The error was calculated by
208 dividing edit distance to the length of the shorter sequence that was compared.
209 We also calculated median accuracy and number of detected species using the R2C2 (Rolling Circle
210 Amplification to Concatemeric Consensus) post-processing pipeline C3POa (Concatemeric Consensus
211 Caller using partial order alignments) for consensus calling (Volden et al. 2018). C3POa generates two
212 kinds of output reads: 1) Consensus reads if the raw read is sufficiently long to cover an insert
213 sequence more than once and 2) Regular “1D” reads if no splint sequence could be detected in the raw
214 read (Adams et al. 2019). We only used consensus reads for downstream analysis. Unlike ASHURE,
215 C3POa does not report information on the RCA fragment length, hence we were not able to make
216 direct comparisons for different thresholds.

217

218 **Results**

219 Mock community

220 Many collected specimens could not be readily identified to species level. Consequently, we employed
221 the Barcode Index Number (BIN) system which examines patterns of sequence variation at COI to
222 assign each specimen to a species proxy (Ratnasingham & Hebert, 2013). We retrieved 50 BINs
223 showing >15% COI sequence divergence from their nearest neighbor under the Kimura 2 \square parameter
224 model (Kimura, 1980). The resulting freshwater macrozoobenthos mock community included
225 representatives of 3 phyla, 12 orders, and 27 families. COI sequences have been deposited on NCBI
226 Genbank under the Accession Numbers MT324068-MT324117. Further specimen details can be found
227 in the public dataset DS-NP50M (dx.doi.org/10.5883/DS-NP50M) on BOLD.

228

229 Metabarcoding using Illumina Sequencing

230 All samples showed good DNA quality. Illumina MiSeq sequencing generated an average of 204 797
231 paired-end reads per primer combination. Raw sequence data are available under the NCBI SRA
232 accession number SRR9207930. We recovered 49 of 50 OTUs present in our mock community (Fig.

233 1D). We obtained a total of 845 OTUs (OTU table including sequences, read counts, and assigned
234 taxonomy is available as Supplementary Table S3) mostly contaminants that were in part also obtained
235 with nanopore sequencing.

236

237 Metabarcoding using Nanopore sequencing

238 Nanopore sequencing with the MinION delivered 746,153/2,756 and 499,453/1,874 1D/1D² reads for
239 Protocols A and B (SRA PRJNA627498), respectively. The 1D approach only sequences one template
240 DNA strand, whereas with the 1D² method both complementary strands are sequenced, and the
241 combined information is used to create a higher quality consensus read (Cornelis et al. 2019). Because
242 of the low read output for 1D² reads, our analyses focused on 1D data. Most reads were skewed
243 towards a shorter read length range (Figure 2) with a median RCA fragment length of 1262bp for
244 Protocol A and 908 bp for Protocol B.

245

246 With flexible filtering (number of targets per RCA fragment = 1 or more), ASHURE results provided a
247 median accuracy of 92.16% for Protocol A and 92.87% for Protocol B (see Table 2, Figures 1A-B).
248 Using ASHURE, we observed a negative, non-significant correlation between consensus median error
249 and the number of RCA fragments (Pearson's r for Protocol A: -0.247, Protocol B: -0.225). For both
250 protocols, we found a positive, non-significant correlation between consensus median error and primer
251 error (Pearson's r for Protocol A: 0.228, Protocol B: 0.375) and between consensus median error and
252 cluster center error (see Figures 3B-C; Pearson's r for Protocol A: 0.770, Protocol B: 0.274). We
253 obtained median accuracy values of >95% for 1/5th of the OTUs in Protocol A and half of the OTUs in
254 Protocol B for flexible filtering. Increasing the number of RCA fragments to 15 or more came with the
255 trade-off of detecting fewer OTUs (from 50 to 36 for Protocol A and 50 to 38 for Protocol B). At the
256 same time, median accuracy values increased to 97.4% and 97.6% for Protocol A and B, respectively.
257 With more stringent filtering (number of targets per RCA fragment = 45 or more), median accuracy
258 improved up to 99.3% for both Protocol A and B but with the trade-off of an overall reduced read
259 output and a reduced number of species recovered (Table 2).

260

261 We mapped the 845 OTUs found in the MiSeq dataset to the Nanopore reads and removed
262 contaminants, (69,911 for Protocol A and 31,045 reads for Protocol B) using ASHURE. With MiSeq,
263 we were able to detect 49 out of 50 of the mock species, whereas all 50 mock community species were
264 detected in both nanopore sequencing protocols A and B. Using the MiSeq dataset, we also removed
265 contaminants from the consensus reads obtained with C3POa (8,843 for Protocol A and 4,222 reads
266 for Protocol B). Using C3POa, we retained a lower number of consensus reads than with ASHURE for
267 Protocol B (see Table 2), but the median consensus accuracy using flexible filtering was similar (94.5-
268 94.7% Protocol A and B). The median accuracy when including all consensus reads was higher for

269 C3POa than ASHURE in both Protocol A and B. Overall the two pipelines showed similar
270 performance in consensus read error profile (Supplementary Figures 2A-D, Supplementary Figure 3).
271 As for Protocol B, ASHURE detected a higher number of mock community species (see Table 2).

272

273 The read error of all consensus reads (Figures 1A-B) spanned a wide range (0-10% error). Running
274 OPTICS, a density-based clustering algorithm, on the consensus reads enabled us to identify cluster
275 centers (Fig. 1C), which possessed comparable accuracy to MiSeq (Fig. 1D). Figures 3A-C show
276 comparisons of consensus error with RCA length, UMI error, and cluster center error. We found that
277 cluster center error correlated better with consensus error, particularly for Protocol A (Pearson's r :
278 0.770), (see Figure 3C). To visualize why OPTICS can identify high fidelity cluster centers, five OTUs
279 were randomly selected and clustered at different RCA fragment lengths (Figure 4). T-distributed
280 stochastic neighbor embedding (t-SNE) was used to visualize the co-similarity relationship of this
281 collection of sequences in two dimensions (Figures 4B-F). Closely related sequences clustered
282 together and corresponded to the OTUs obtained by OPTICS. Clustering of raw reads resulted in less
283 informative clusters, where OTUs were not well separated and cluster membership did not match that
284 of the true species (Fig. 4C). The clustering of reads with increasing RCA length cut-off resulted in
285 clusters that had more distinct boundaries (Figures 4D-F). These clusters corresponded to the true
286 haplotype sequences (Fig. 4F) and contained the de novo cluster centers and true OTU sequences at
287 their centroids. The OPTICS algorithm successfully extracted the OTU structure embedded in a co-
288 similarity matrix, flagged low fidelity reads that were in the periphery of each cluster, and ordered
289 high fidelity reads to the center of the clusters (Fig. 4B).

290

291 **Discussion**

292 This study introduces a workflow for DNA metabarcoding of freshwater organisms using the
293 Nanopore MinIONTM sequencing platform. We were able to show that it is possible to mitigate the
294 high error rates associated with nanopore-based long-read single-molecule sequencing by using rolling
295 circle amplification with a subsequent assembly of consensus sequences leading to a median accuracy
296 of up to 99.3% for long RCA fragments (>45 barcodes).

297

298 We were able to retrieve all OTUs of the mock community assembled for this study. Our mock sample
299 species had at least 15% genetic distance to each other and with ASHURE we were able to retrieve
300 them both under relaxed and strict filtering conditions. This will likely change if a sample includes
301 species that are more closely related with average distances of 2-3%. Although both of our
302 experimental protocols were successful, we observed a higher number of consensus reads, detected
303 species overall and median accuracy for Protocol B which used a higher number of RCA replicates as
304 input DNA, had no mechanical fragmentation step, and a reduced duration of enzymatic debranching

305 (Table 2). We recommend adopting our Protocol B workflow and using strict filtering in the ASHURE
306 pipeline, e.g. a minimum of 15 barcodes per RCA fragment. We used the Illumina MiSeq platform to
307 identify by-products or contaminants as well as for comparison with nanopore sequencing. In terms of
308 accuracy the MiSeq platform performs slightly better (Figure 1C and D). However, the improved error
309 rates clearly make the MinIONTM a more cost-effective and mobile alternative.

310

311 Consensus sequence building is the critical step for achieving high accuracy with MinIONTM reads.
312 Raw outputs of Nanopore sequencing are improving (Volden et al. 2018) and as read accuracy further
313 improves, so will the quality of consensus sequences. We show that RCA is integral for increasing
314 consensus accuracy, but it is also the most time-consuming step during the laboratory workflow, e.g.
315 with 60-70 ng/ul of input DNA 5-6 hours of RCA were necessary to achieve reasonable results. Our
316 results display a trade-off between median consensus accuracy and the detection of species,
317 particularly due to not having enough long reads (see Table 2, Fig. 2). However, despite most reads
318 being relatively short, we observed an inverse correlation between RCA length and the consensus error
319 rate (Fig. 3A). For further improvement of consensus sequence accuracy, the proportion of longer
320 reads needs to be maximized. For more time-sensitive studies on metabarcoding with Nanopore
321 sequencing, e.g. field-based studies, we suggest modifying the RCA duration based on the complexity
322 of the sample. However, given some of the RCA weaknesses, we recommend the exploration of other
323 isothermal amplification procedures such as LAMP (Imai et al. 2017), multiple displacement
324 amplification, (MDA) (Hansen et al. 2018), or recombinase polymerase amplification, (RPA) (Donoso
325 & Valenzuela, 2018).

326

327 Previous studies using circular consensus approaches to Nanopore sequencing, such as INC-seq (Li et
328 al. 2016) and R2C2 (Volden et al. 2018) have already shown improvements in read accuracy. We
329 compared our pipeline ASHURE with C3POa, the post-processing pipeline for R2C2 with a reported
330 median accuracy of 94% (Volden et al. 2018). C3POa data processing includes the detection of DNA
331 splint sequences and the removal of short (<1,000 kb) and low-quality (Q < 9) reads (Volden et al.
332 2018). With C3POa, a raw read is only used for consensus calling if one or more specifically designed
333 splint sequences are detected within it (Volden et al. 2018). Instead of splint sequences we used primer
334 sequences to identify reads for further consensus assembly. Both C3POa and ASHURE showed
335 similar accuracy for our datasets, but C3POa detected fewer species in our Protocol B experiment.
336 Using ASHURE, we were only able to detect 43.4% and 7% of the reads with both primers attached in
337 Protocol A and B, respectively. This points to some issues with the RCA approach and might explain
338 why C3POa generated fewer numbers of consensus reads in Protocol B, as the number of detected
339 sequences was very low. Initially we assumed that increasing the unique molecular identifier (UMI)
340 length for our primers would be useful not only for consensus calling but also for identifying,

341 quantifying, and filtering erroneous consensus reads. However, within the small percentage of reads
342 with both primers attached, we did not find a strong correlation between the UMI error and the
343 consensus read error (Figure 3B).

344

345 Several MinIONTM studies have implemented a reference-free approach for consensus calling,
346 however, these studies are limited to tagged amplicon sequencing that allows for sequence-to-
347 specimen association (Srivathsan et al. 2018, Calus et al. 2018; Pomerantz et al. 2018; Srivathsan et al.
348 2019). Such an approach can be useful for species-level taxonomic assignment (Benítez-Páez et al.
349 2016) and even species discovery (Srivathsan et al. 2019). Our pipeline uses density-based clustering
350 which is a promising approach when studying species diversity in mixed samples, particularly with
351 Nanopore sequencing. The density-based clustering of Nanopore reads allows for a reference-free
352 approach by grouping reads with their replicates without having to map to a reference database
353 (Faucon et al. 2017). Conventional OTU threshold clustering approaches have shown to be a challenge
354 for nanopore data. Either each sequence was assigned to a unique OTU, or OTU assignment failed due
355 to the variable error profile (Ma et al. 2017), or the optimal threshold depended on the relative
356 abundance of species in a given sample (Mafune et al. 2017). Density-based clustering is advantageous
357 because it can adaptively call cluster boundaries based on other objects in the neighborhood (Ankerst
358 et al. 1999). Clusters correspond to the regions in which the objects are dense, and the noise is
359 regarded as the regions of low object density (Ankerst et al. 1999). For DNA sequences, such a
360 clustering approach requires sufficient read coverage around a true amplicon so that the novel clusters
361 can be detected and are not treated as noise. With sufficient sample size, density-based approaches can
362 allow us to obtain any possible known or novel species clusters with high accuracy and without the
363 need for a reference database. ASHURE is not limited to RCA data, as it performs a search for primers
364 in the sequence data, splits the reads at primer binding sites, and stores the information on start and
365 stop location of the fragment as well as its orientation. The pipeline can be used to process outputs of
366 other isothermal amplification methods generating concatenated molecules by simply providing
367 primer/UMI sequences that link each repeating segment.

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369 **Conclusion**

370 This study demonstrates the feasibility of bulk sample metabarcoding with Oxford Nanopore
371 sequencing using a modified molecular and novel bioinformatics workflow. We highly recommend the
372 use of isothermal amplification techniques to obtain longer repetitive reads from a bulk sample. With
373 our pipeline ASHURE, it is possible to obtain high-quality consensus sequences with up to 99.3%
374 median accuracy and to apply a reference-database free approach using density-based clustering. This
375 study was based on aquatic invertebrates, but the pipeline can be extended to many other taxa and
376 ecological applications. By offering portable, highly accurate, and species-level metabarcoding,

377 Nanopore sequencing presents a promising and flexible alternative for future bioassessment programs
378 and it appears that we have reached a point where highly accurate and potentially field-based DNA
379 metabarcoding with this instrument is possible.

380 **Table 1:** Varying RCA conditions for experimental protocols A and B

Dataset	Protocol A	Protocol B
RCA duration (hrs)	5	6
Number of target sequences per RCA fragment	12	15
Enzymatic branching (min)	5	2
Mechanical fragmentation	4200 rpm, 2 min	None
Primer pairs used	HCOA-LCO, HCOC2-LCOC2	HCOA2-LCOA2, HCOC2-LCOC2

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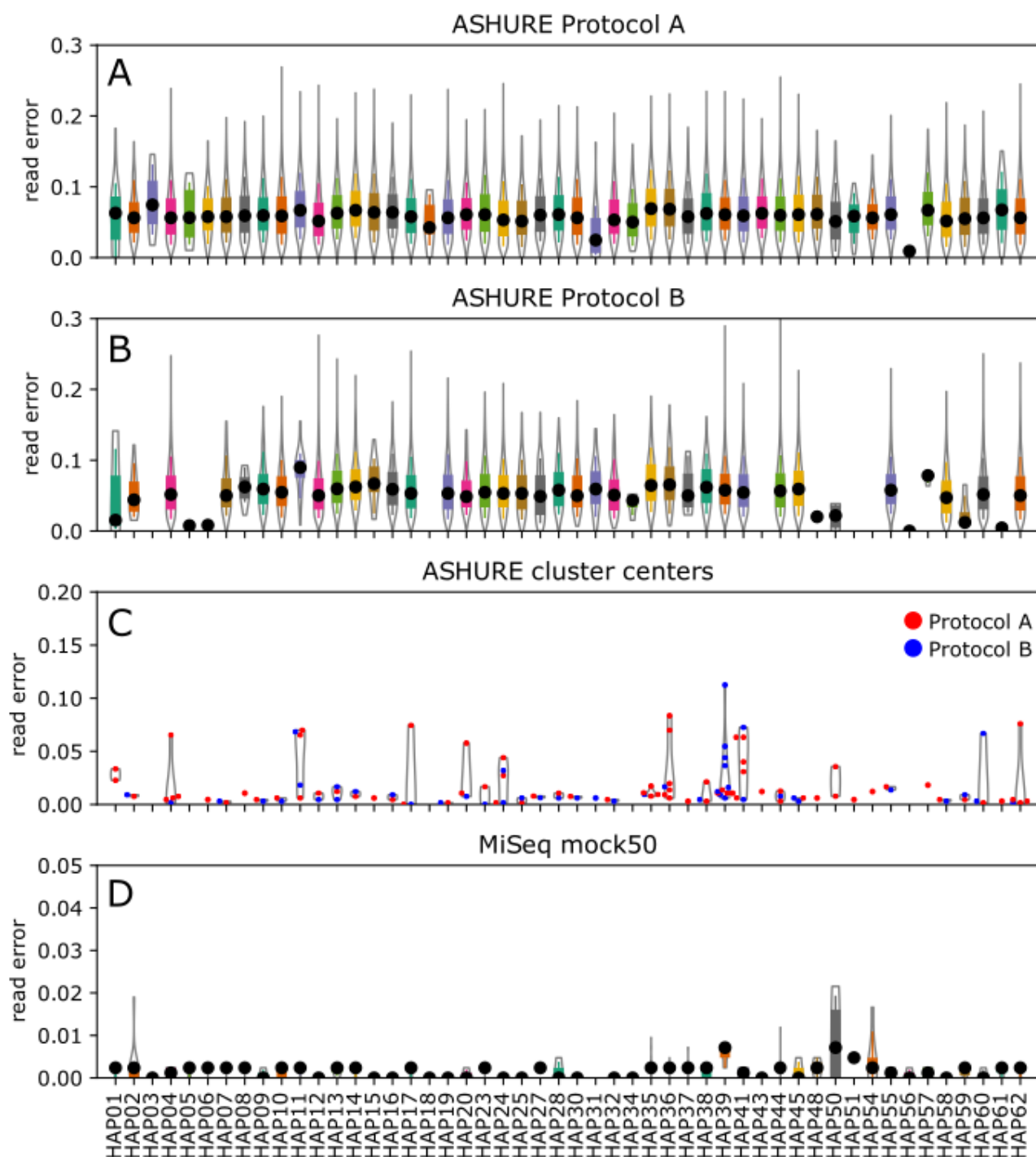
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383 **Table 2:** Consensus reads, median accuracy, and the number of OTUs/species detected at different
 384 thresholds for Protocol A and B analyzed with ASHURE and C3POa.

ASHURE pipeline						
	Protocol A			Protocol B		
Consensus read criterium	# of reads	Median accuracy (%)	# of OTUs detected	# of reads	Median accuracy (%)	# of OTUs detected
unfiltered	269,620	93.6	198	245,827	93.4	188
post filtering non-target data based on MiSeq	199,709	92.16	50	214,782	92.87	50
RCA > 15	1,434	97.39	36	2,884	97.62	38
RCA > 20	292	97.86	28	1,009	98.10	34
RCA > 25	78	98.22	19	455	98.35	30
RCA > 30	20	98.46	11	217	98.57	26
RCA > 35	7	99.05	5	106	98.82	22
RCA > 40	3	99.52	2	57	99.05	18
RCA > 45	2	99.60	2	30	99.29	13
RCA > 50	1	99.68	1	21	98.82	8
C3POa						
unfiltered	322,884	94.5	180	128,353	94.7	118
post filtering non-target data based on MiSeq	314,041	94.5	50	124,131	94.7	40

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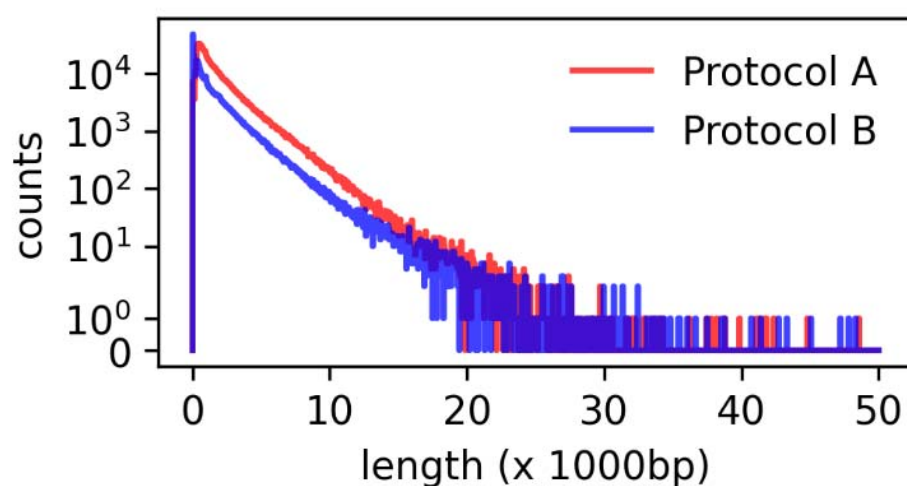
388 **Figure 1:** Nanopore sequencing read error per species for (A) Protocol A and (B) Protocol B obtained
389 with ASHURE using all reads. (C) Nanopore sequencing read error obtained with OPTICS in
390 ASHURE using cluster centers for each RCA condition. (D) MiSeq sequencing read error per species.

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396 **Figure 2:** Read length distribution for both sequencing protocols. The number of reads is provided in a
397 logarithmic scale on the y-axis.

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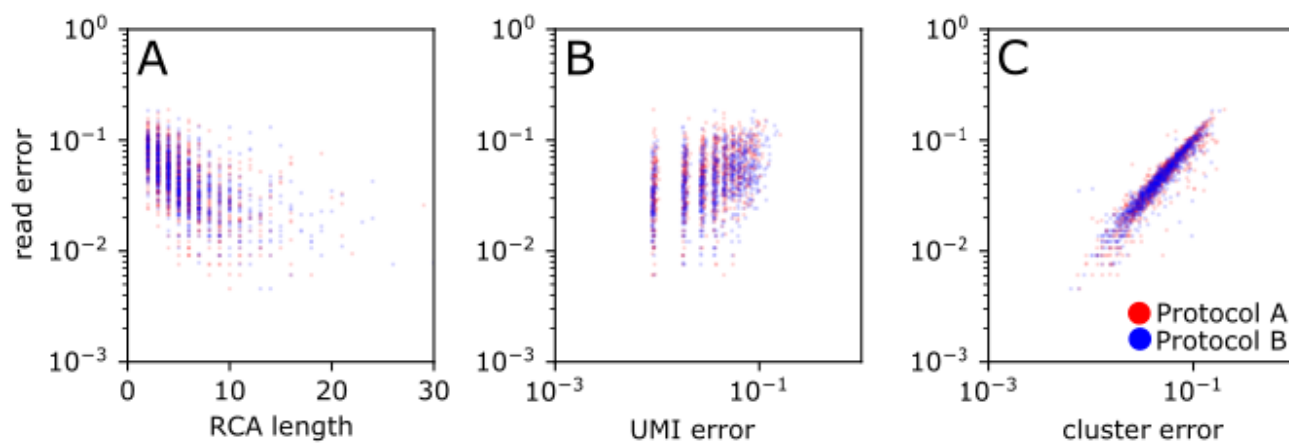
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414 **Figure 3:** Comparison of consensus error versus (A) RCA length, (B) UMI error, and (C) cluster
415 center error using the ASHURE pipeline for two RCA conditions.

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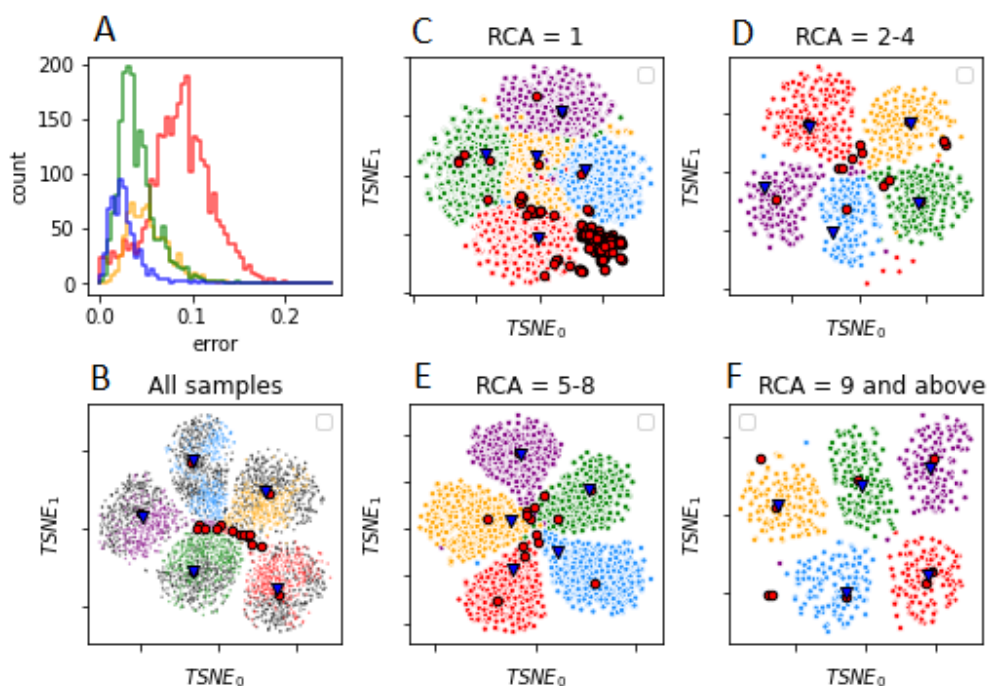
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433 **Figure 4:** tSNE visualization of reference-free clustering using OPTICS for five randomly selected
434 haplotypes. (A) The number of reads and percentage of error for each filtering criteria, red: reads with
435 1 RCA fragment, yellow: reads with 2-4 RCA fragments, green: reads with 5-8 RCA fragments, and
436 blue: reads with 9 or more RCA fragments. tSNE visualization of OPTICS clusters for reads with (B)
437 no filtering, (C) one RCA fragment, (D) 2-4 RCA fragments, (E) 5-8 RCA fragments, (F) 9 and more
438 RCA fragments. True haplotypes (blue triangles) and cluster centers obtained with reference-free
439 clustering (red circles) overlap more as the number of RCA fragments increases. Colors in B-F
440 correspond to: HAP04 (red), HAP11 (blue), HAP17 (purple), HAP39 (orange), HAP41 (green). Grey
441 dots in (B) indicate outliers.

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459 Author contributions

460 BB, VE, TB, and DS designed the experiments; BB and SM assembled the mock community, BB did
461 lab work; VE did the MiSeq experiment, BB and ZC analyzed the data; BB and DS wrote the
462 manuscript, all authors contributed to the manuscript.

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