

1 **Evaluation of Oxford Nanopore MinION™ Sequencing for 16S**  
2 **rRNA Microbiome Characterization**

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## 23 **Abstract**

24 In this manuscript we evaluate the potential for microbiome characterization by sequencing of  
25 near-full length 16S rRNA gene region fragments using the Oxford Nanopore MinION (hereafter  
26 ‘Nanopore’) sequencing platform. We analyzed pure-culture *E. coli* and *P. fluorescens*, as well  
27 as a low-diversity mixed community sample from hydraulic fracturing produced water. Both  
28 closed and open reference operational taxonomic unit (OTU) picking failed, necessitating the  
29 direct use of sequences without OTU picking. The Ribosomal Database Project classifier against  
30 the Green Genes database was found to be the optimal annotation approach, with average pure-  
31 culture annotation accuracies of 93.8% and 82.0% at the phyla and genus levels, respectively.  
32 Comparative analysis of an environmental sample using Nanopore and Illumina MiSeq  
33 sequencing identified high taxonomic similarity when using a weighted metric (Bray-Curtis), and  
34 significantly reduced similarity when using an unweighted metric (Jaccard). These results  
35 highlight the great potential of Nanopore sequencing to analyze broad microbial community  
36 trends, and the challenge of applying Nanopore sequencing to discern rare taxa in mixed  
37 microbial communities. Finally, we observed that between-run carryover following washes on  
38 the same flowcell accounted for >10% of sequence reads, necessitating future development to  
39 either prevent carryover or filter sequences of interest (e.g. barcoding).

40

## 41 **Introduction**

42 Interest in studying the microbiome, microbiota associated with various environments, has  
43 exploded in recent years largely due to the rapid expansion in ‘next-generation sequencing’  
44 capabilities and subsequent reduction in cost. It has been recognized that the human microbiome  
45 plays an important role in many different clinical outcomes, including obesity (1), immune state

46 (2), and infection (3). The human microbiome is comprised of a diverse array of commensal  
47 microorganisms in and on the human body, and emerging research has suggested a clinical role  
48 for the microbiome in either therapeutic development (e.g. probiotics) (4) or diagnostics (5).  
49 Concurrently, significant interest has emerged in the microbiomes of various other environments,  
50 such as buildings (6-8) and water systems (9).

51 Currently, the most common microbiome analysis approach is high-throughput amplicon  
52 sequencing of the 16S rRNA gene region. The most widely used sequencing technologies are  
53 Illumina sequencing platforms (e.g. the MiSeq and HiSeq). These platforms are accurate and  
54 generate a large amount of data, but are limited by capital costs, the necessity to pool samples to  
55 reduce per-sample costs, sequence read length, and a turnaround time of days to weeks that may  
56 be inadequate for many applications. Recently, Oxford Nanopore has released a small and  
57 inexpensive sequencing platform called the MinION<sup>TM</sup>, which has been previously reviewed (10,  
58 11). Capital costs are reduced to per-run costs that are comparable with the current Illumina  
59 platforms, and data analysis is possible in near-real time, enabling investigators to generate  
60 sequence data as-needed. For example, samples were correctly assigned to the *Salmonella*  
61 species in 20 minutes and serotype in 40 minutes (12). Amplicon (non-metagenomic) sequencing  
62 has also been successfully employed to identify both bacterial and viral origin (13, 14).  
63 Additionally, this technology routinely produces sequences >10kb in length, enabling  
64 sequencing of the full 16S rRNA gene region and more reliable taxonomic placement. Despite  
65 the benefits, the primary drawback to this technology is relatively high error rates (currently  
66 reported to be ~8%), hindering metagenomic analysis of highly diverse microbiome samples and  
67 requiring additional development and validation. Previous investigations have demonstrated  
68 successful 16S rRNA microbial community classification via sequencing of a single mouse gut

69 microbiome (15) and a single mock microbial community (16), including species-level  
70 assignment (16); however, a formal assessment of 16S rRNA sequencing on the Nanopore  
71 platform, including analysis of pure-culture samples for annotation validation, is currently  
72 lacking. Development of Nanopore technology for microbiome analysis would enable rapid (<12  
73 hours from sample to results) and low cost microbiome characterization that would be applicable  
74 to both the clinic and other applications.

75  
76 In the current study we evaluate the potential for microbiome characterization via Nanopore  
77 sequencing of near full-length 16S rRNA PCR amplicons. First, we evaluate the accuracy and  
78 annotation strategies of sequences from pure culture *E. coli* and *P. fluorescens* to determine the  
79 most appropriate sequence analysis approach. We then compare performance of Nanopore  
80 sequencing against the current state of the art Illumina sequencing using a sample from hydraulic  
81 fracturing wastewater. Finally, we investigate an apparent between-run carryover phenomena,  
82 and propose necessary future investigations to enable 16S rRNA microbiome characterization on  
83 the Nanopore platform.

84

85

## 86 **Results and Discussion**

### 87 *Sequencing*

88 Sequencing was performed on eight 16S rRNA libraries; however, due to apparent between-run  
89 carryover (discussed below), only the initial run on each of three flow cells was used for  
90 subsequent analyses of annotation approach and accuracy. These sequencing libraries were

91 derived from two pure-culture samples, *E. coli* and *P. fluorescens*, and a low-diversity sample  
92 from hydraulic fracturing produced water. Detailed sequencing results are shown in Table 1.  
93 Raw sequence data can be found on Figshare using DOI:  
94 <https://dx.doi.org/10.6084/m9.figshare.4515752.v3>.

95

### 96 ***Pure Culture Analyses***

97 We first evaluated the suitability of Operational Taxonomic Unit (OTU) picking methods using  
98 Nanopore sequence data. OTU picking via comparison with a reference library, i.e. closed-  
99 reference OTU picking, failed, with no sequences being assigned to an OTU (i.e. all sequences  
100 were excluded). *De novo* OTU clustering, i.e. OTU clustering by determining between-sequence  
101 similarity, was then evaluated using similarity thresholds between 90-100%. At the typically  
102 used similarity threshold of 97%, all sequences from both pure-culture samples were assigned to  
103 unique OTUs (i.e. the ratio of OTUs to sequences was one). Results from *de novo* OTU  
104 clustering evaluation are shown in Figure 1. These results highlighted the challenge of clustering  
105 reads from long, error-prone sequences, and necessitated analyzing the taxonomy of sequences  
106 individually without OTU picking.

107

108 We next evaluated the ability to accurately annotate the taxonomy of pure-culture Nanopore 16S  
109 rRNA sequences using three different annotation approaches: the naïve Bayesian Ribosomal  
110 Database Project (RDP) classifier with the Green Genes database; the RDP classifier against the  
111 RDP database; and BLAST against the Green Genes database. Results from this evaluation are  
112 shown in Figure 2. The RDP classifier against the Green Genes database was found to be the best  
113 performing annotation strategy. Using this approach, the annotation accuracy for *E. coli* was

114 96.7% and 81.9% at the phyla and genus level, respectively, and was 90.9% and 82.0% for *P.*  
115 *fluorescens* at the phyla and genus level, respectively.

116

### 117 ***Comparison with Illumina Sequencing***

118 We next evaluated Nanopore sequencing to characterize an environmental sample from  
119 hydraulic fracturing produced water (17). This sample was selected as it exhibited low alpha-  
120 diversity in previous analyses. The same DNA extract was used for both analyses.

121

122 By assigning taxonomy to each individual sequence, five phyla were identified by the Nanopore  
123 platform, and eleven phyla were identified by the Illumina platform. Among them, four phyla  
124 that together accounted for greater than 99% of sequence relative abundance were identified by  
125 both platforms (Table 2). Nine shared genera were detected by both platforms, accounting for  
126 relative abundances of 98.3% on the Nanopore platform and 81.6% on the Illumina platform  
127 (Table 2). Both the Nanopore and Illumina platforms revealed similar microbial community  
128 structure of the produced water sample. The Firmicutes Phylum dominated the microbial  
129 community with relative abundance higher than 90% with both platforms (Table S1). Phyla  
130 unique to the Nanopore and Illumina platforms accounted for less than 0.5% of relative  
131 abundance (Table S1). At the genus level, the produced water microbial community was  
132 dominated by the genus *Halanaerobium* (Table S2), with 95.6% of Nanopore sequences and  
133 76.8% of Illumina sequences being assigned to *Halanaerobium*. In addition, 14.6% of Illumina  
134 16S rRNA sequence reads were assigned to Clostridiales, which is within the same class with  
135 *Halanaerobium* (Clostridia) (Table S2).

136

137 We also calculated the Jaccard distance (solely based on presence/absence of each taxa) and  
138 Bray-Curtis distance (based on both presence/absence and relative abundance of each taxa)  
139 between the produced water microbial community revealed by Nanopore and Illumina platforms  
140 at different taxonomic levels. Jaccard and Bray-Curtis distances measure the degree of microbial  
141 community structure difference between two samples, with a value of one indicating no  
142 community structure overlap and a value of zero indicating identical microbial communities. We  
143 adopted these measures to evaluate the level of difference between the technical replicates of the  
144 same produced water sample between the Nanopore and Illumina sequencing platforms. The  
145 Jaccard distance increased from 0.62 at the phylum level to 0.89 at genus level; the Bray-Curtis  
146 distance increased from 0.04 at phylum level to 0.22 at genus level (Figure 3). Jaccard distances  
147 were higher than Bray-Curtis distances at all phylogenetic levels, because more taxa were  
148 assigned using short sequence reads by Illumina sequencing (Table 2) and Jaccard distance only  
149 accounts for the presence and absence of each assigned taxa whereas Bray-Curtis distances take  
150 relative abundance into account.

151  
152 Pearson correlation of the relative abundance of each taxon between Nanopore and Illumina data  
153 was conducted to further evaluate the reproducibility of sequencing results between the two  
154 platforms. Significant correlation was found at all phylogenetic levels from phylum to genus (R  
155 values > 0.98, p values < 0.001), indicating reproducible taxonomic assignment results can be  
156 obtained between Nanopore and Illumina platforms.

157  
158 It should be noted in these comparisons that different primer sets were used for the two analysis  
159 approaches, which has previously been shown to bias microbiome community structure as

160 analyzed by 16S rRNA sequencing (18). Despite this additional source of bias, the above  
161 analyses imply that the weighted community structure is comparable between the two platforms,  
162 encouraging future development.

163

#### 164 ***Between-Run Sample Carryover***

165 We noted an apparent carryover of sequences between pure culture runs of *P. fluorescens* and *M.*  
166 *smegmatis*. The potential for sequence carryover has been anecdotally reported in the literature  
167 (19). We subsequently excluded all runs except the first run on each flow cell from earlier  
168 analyses, and undertook a formal evaluation of sequence carryover.

169

170 Results from analysis of the *Mycobacterium smegmatis* run are shown in Figure 4. In this run,  
171 76.5% of sequences were assigned to the correct Actinobacteria phyla, 11.0% of sequences were  
172 incorrectly assigned to another domain or unable to be assigned, and 12.4% of sequences were  
173 incorrectly assigned to the Proteobacteria phyla, presumptively resulting from sequence  
174 carryover. 54.5% of sequences were assigned to the correct *Mycobacterium* genus whereas  
175 10.8% of sequences were assigned to the *Pseudomonas* genus.

176

177 Following the *P. fluorescens* and *M. smegmatis* runs, we subsequently ran an additional  
178 environmental sample derived from river water and re-ran a *M. smegmatis* sample using 1D  
179 technology. In the second *M. smegmatis* run, we observed only 0.5% of sequences to be assigned  
180 to the correct Mycobacteriaceae family, compared with 54.5% in the first *M. smegmatis* run.  
181 These results imply that continued carry-over serves to significantly decrease output quality;



182 however, additional validation with controlled microbial community composition is necessary to  
183 confirm this observation.

184

### 185 ***Areas of Future Development***

186 This investigation has identified multiple necessary areas of future development to enable 16S  
187 rRNA microbiome characterization on the Nanopore platform. First, strategies to exclude  
188 between-run carryover, either via improved washing between runs or a barcode approach, would  
189 enable multiple runs on the same flow cell, significantly reducing per-run costs. In this  
190 investigation we performed six runs on the same flow cell while observing minimal output loss.  
191 Second, improved bioinformatics strategies are necessary to exclude poor quality sequences. In  
192 the *E. coli* and *P. fluorescens* runs, 3.3 and 9.0% of sequences, respectively, were not assigned to  
193 any phyla, suggesting poor sequence quality and confounding both alpha- and beta-diversity  
194 analyses. Finally, it would be beneficial to develop a 16S rRNA annotation pipeline based upon  
195 optimized analysis strategies that provides output in near real-time, facilitating field and clinic  
196 applications and alleviating current bioinformatics challenges from interested investigators.  
197 Ultimately, the development of a rapid and low-cost microbiome approach will facilitate the  
198 application of clinical and environmental microbiome technologies.

### 199 **Materials and Methods**

#### 200 **Overview of DNA samples and sequencing libraries**

201 Three pure culture bacterial DNA samples and an environmental DNA sample were analyzed in  
202 the current study. The three pure culture bacterial DNA samples were extracted from pure  
203 cultures of: (1) *Escherichia coli* (ATCC 15597), (2) *Pseudomonas fluorescens* (ATCC 13525),  
204 and (3) *Mycobacterium smegmatis* str. mc2 155, respectively. The environmental DNA samples,

205 which had previously undergone 16S rRNA gene sequencing using the Illumina MiSeq platform,  
206 were a hydraulic fracturing produced water sample (17) and a river water sample (unpublished).

207  
208 All Nanopore sequencing runs were conducted using the MinION Mk IB platform following  
209 recommended sequencing protocols (Oxford Nanopore Technologies). The *E. coli* 16S rRNA  
210 amplicon 2D library and the hydraulic fracturing produced water 16S rRNA amplicon 2D library  
211 were sequenced individually using a Nanopore MIN-105 flow cell and a Nanopore MIN-106  
212 flow cell, respectively. The remaining libraries were sequenced on a MIN-106 flow cell  
213 following a sequential order: (1) *P. fluorescens* 2D library; (2) *M. smegmatis* 2D library; (3) river  
214 water sample 2D library; (4) *M. smegmatis* 1D library. For the sequential sequencing runs, flow  
215 cell washing was conducted immediately following the completion of the previous sequencing  
216 run using a Nanopore washing kit WSH002 (Oxford Nanopore). The Oxford Nanopore  
217 recommended washing protocol was used between runs, namely 150  $\mu$ L of WSH002 solution A  
218 was loaded to the flow cell through priming port and incubated at room temperature for 10  
219 minutes, then 150  $\mu$ L of WSH002 solution B was loaded through the priming port before the  
220 next sequencing run and incubated for another 10 minutes at room temperature.

221

## 222 **Nanopore sequencing library preparation**

223 Previously described universal primers targeting the 16S rRNA gene region (S-D-bact-0008-c-  
224 S20 and S-D-bact-1391-a-A-17) (20) were used for PCR. Each PCR was conducted in a total  
225 volume of 50  $\mu$ L, containing 5  $\mu$ L 10x buffer, 5  $\mu$ L dNTP mix, 2.5  $\mu$ L of each forward and  
226 reverse primer, 0.25  $\mu$ L DreamTaq, 1  $\mu$ L template DNA, and 33.75  $\mu$ L nuclease free molecular  
227 grade water. The temperature condition for the PCR was 3 minutes at 95°C; 30 cycles composed

228 of 20 seconds at 95 °C, 30 seconds at 47 °C for annealing, 1 minute at 72 °C; and a final  
229 elongation at 72 °C for 15 minutes. All PCR products were purified using Ampure XP beads and  
230 normalized to 45 µL containing 1 µg of purified PCR products. Negative controls were used for  
231 all PCR reactions and DNA extractions, and all controls were negative.

232  
233 2D libraries were prepared using a Nanopore NSK007 sequencing kit and recommended protocol  
234 (Oxford Nanopore Technologies). The end repair step of the purified PCR products was  
235 conducted by adding 7 µL Ultra II End-Prep buffer, 3 µL Ultra II End-Prep enzyme mix (New  
236 England Biolabs), and 5 µL control DNA provided with Nanopore NSK007 sequencing kit. The  
237 end repair reaction mix was incubated at 20°C for 5 minutes and 65°C for 5 minutes. The end-  
238 repaired PCR products were further purified using AMPure XP beads and ligated to the  
239 sequencing adapters by adding 8 µL molecular grade water, 10 µL Nanopore NSK007 adapter  
240 mix, 2 µL Nanopore NSK007 HPA solution, and 50 µL Blunt/TA Master Mix (New England  
241 Biolabs), and then incubated at room temperature for 10 minutes. 1 µL HPT solution from the  
242 NSK007 kit was added and incubated for an additional 10 minutes at room temperature. The  
243 ligated and tethered 2D libraries were purified by using MyOne C1 beads (Thermo Scientific)  
244 and eluted in 25 µL elution buffer (Oxford Nanopore Technologies). A description of the *M.*  
245 *smegmatis* 1D library preparation is included in the Supplementary Information.

246  
247 All sequencing flow cells were primed using 500 µL Running Buffer Fuel Mix diluted in 500 µL  
248 molecular grade water following the recommended priming protocol (Oxford Nanopore  
249 Technologies). After priming, 6 µL of each 2D sequencing library was mixed with 37.5 µL

250 Running Buffer Fuel Mix (Oxford Nanopore Technologies) and 31.5  $\mu$ L molecular grade water,  
251 then loaded to Nanopore flow cell for sequencing.

252

## 253 **Sequence data processing**

### 254 **Base-calling and initial format conversion**

255 The raw FAST5 files were base-called using Metrichor v2.42.2 with 2D Basecalling for FLO-  
256 MIN106 250bps workflow and 1D Basecalling for FLO-MIN106 450bps workflow. Passed 2D  
257 reads of 2D sequencing libraries and passed template reads of 1D sequencing libraries were  
258 converted to FASTA files for downstream analysis using Poretools (21).

259

### 260 **Operational Taxonomic Unit Evaluation**

261 Operational taxonomic unit (OTU) picking was conducted by using both closed-reference and *de*  
262 *novo* picking strategies implemented in QIIME 1.9.2 (22). Closed-reference OTU picking was  
263 conducted by using Greengenes 13.8 as the reference database (23).

264

### 265 **Taxonomy Assignment**

266 For pure culture *E. coli* and *P. fluorescens* sequencing data, taxonomy was assigned to each  
267 sequence read within QIIME 1.9.0 (24) using the RDP classifier (25) against Greengenes 13.8  
268 (23) and RDP 16S rRNA training set (25) as the reference database, respectively; as well as  
269 using BLASTn (26) against Greengenes 13.8 (23) as a reference database. For subsequent  
270 analyses, taxonomy was assigned to each individual sequence read using the RDP classifier (25)  
271 against Greengenes 13.8 (23) as this approach was found to achieve the highest taxonomy  
272 assignment accuracy for the pure culture *E. coli* and *P. fluorescens* sequence data.

273  
274 Illumina 16S rRNA amplicon sequencing data of the produced water and river water samples  
275 were re-processed using the same approach as Nanopore sequence data. The Illumina data were  
276 clustered into OTUs using 100% similarity threshold, i.e. each identical Illumina sequencing  
277 read was assigned taxonomy using RDP classifier (25) against Greengenes 13.8 (23).

278  
279 Jaccard and Bray-Curtis distances, which are dissimilarity distances measuring level of  
280 dissimilarity between two microbial communities, were calculated using QIIME 1.9.2 (22) to  
281 measure the degree of similarity of the produced water microbial community similarity between  
282 Nanopore and Illumina platforms. Significance of correlation between the two technical  
283 replicates of the produced water sample across sequencing platforms was conducted using  
284 Pearson's correlation implemented in Minitab 16.

285  
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289 initially provided access to the sequencing platform at reduced cost.

290  
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378

379 Figure 1. Number of observed *de novo* operational taxonomic units (OTUs) per number of  
380 sequences at different similarity thresholds for *E. coli* and *P. fluorescens* pure culture DNA  
381 sample sequenced by Nanopore.

382 Figure 2. Accuracy of taxonomy assignment at different phylogenetic levels for (A) *E. coli*; and  
383 (B) *P. fluorescens*

384 Figure 3. Jaccard (unweighted) and Bray-Curtis (weighted) dissimilarity values for a hydraulic  
 385 fracturing produced water sample analyzed by Nanopore and Illumina sequencing; distance  
 386 value of one indicating no community structure overlap, and a value of zero indicating identical  
 387 community structure

388 Figure 4. Taxonomy assignment of *Mycobacterium smegmatis* 16S rRNA pure culture  
 389 sequencing following *Pseudomonas fluorescens* sequencing.

390

Table 1. Sequencing results and output from Oxford Nanopore MinION runs.

Sample	Sequencing Chemistry	Total Reads	Passed Reads (%)	Average Q-Score of Pass Reads ( $\pm$ SD)	Average Read Length ( $\pm$ SD)	Run Time (Min)
<i>E. coli</i>	2D	4093	1447 (35.4%)	$11.95 \pm 1.44$	$1304.23 \pm 302.70$	15
<i>P. fluorescens</i>	2D	7374	2671(36.2%)	$13.73 \pm 1.67$	$1277.99 \pm 321.87$	20
<i>M. smegmatis</i> *	2D	11627	4502(38.7%)	$13.56 \pm 1.7$	$1263.30 \pm 300.29$	20
Produced Water	2D	16525	5461 (33.0%)	$13.79 \pm 1.66$	$1320.96 \pm 215.06$	20
River Water	2D	9745	3143 (32.3%)	$13.0 \pm 1.76$	$1233.62 \pm 385.52$	20
<i>M. smegmatis</i>	1D	10383	7463 (71.9%)	$7.82 \pm 0.78$	$1414.44 \pm 697.89$	20

\**M. smegmatis* 2D, river water sample, and *M. smegmatis* 1D library were excluded from primary analyses due to apparent carryover from previous runs.

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Table 2. Comparison of number of taxa at different phylogenetic levels for the produced water sample

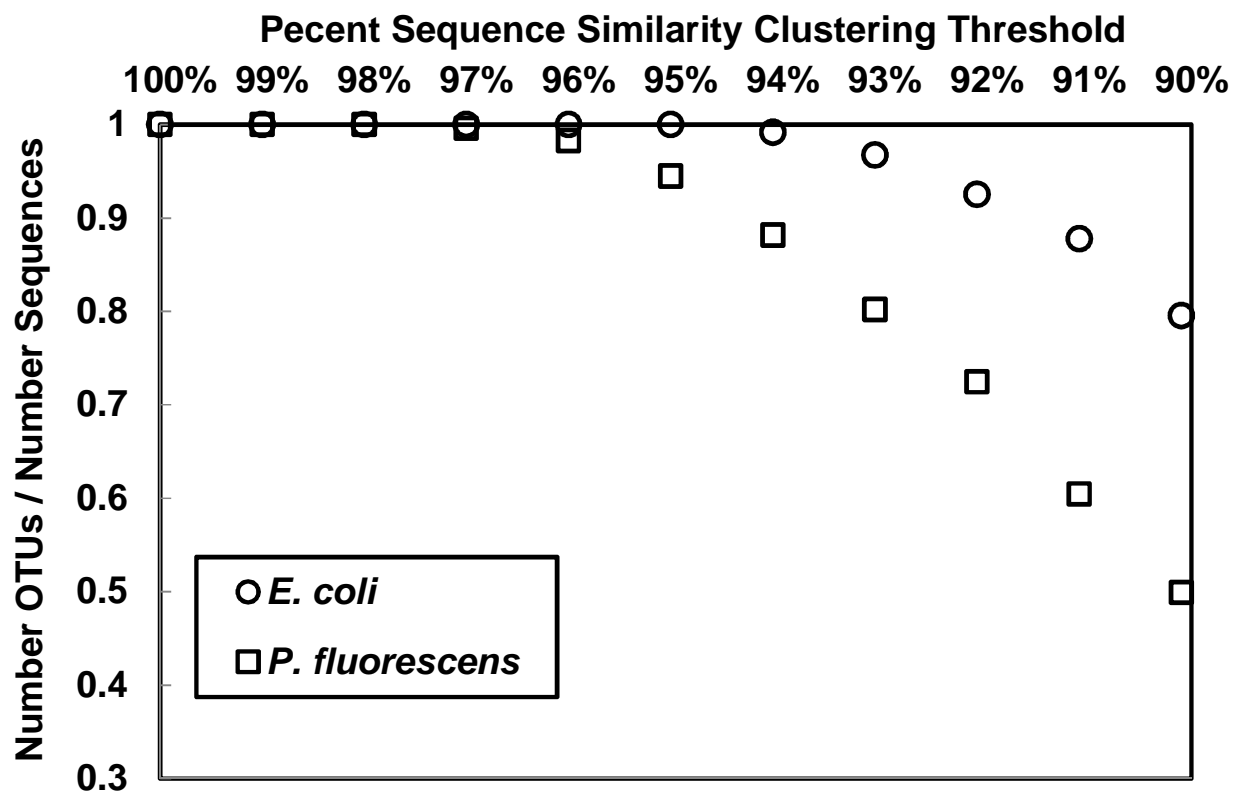
	Phylum	Class	Order	Family	Genus
Number of taxa assigned using Nanopore data	5	9	13	17	20
Number of taxa assigned using Illumina data	11	21	36	63	84
Number of shared taxa	4	8	11	13	9
Total relative abundance of shared taxa (Nanopore)	99.9%	100.0%	99.9%	98.5%	98.3%
Total relative abundance of shared taxa (Illumina)	99.4%	98.0%	96.3%	81.8%	81.6%



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395 Table S1. Relative abundance distribution at phylum level for the produced water sample.

396 Table S2. Relative abundance distribution of Firmicutes phylum for the produced water sample.

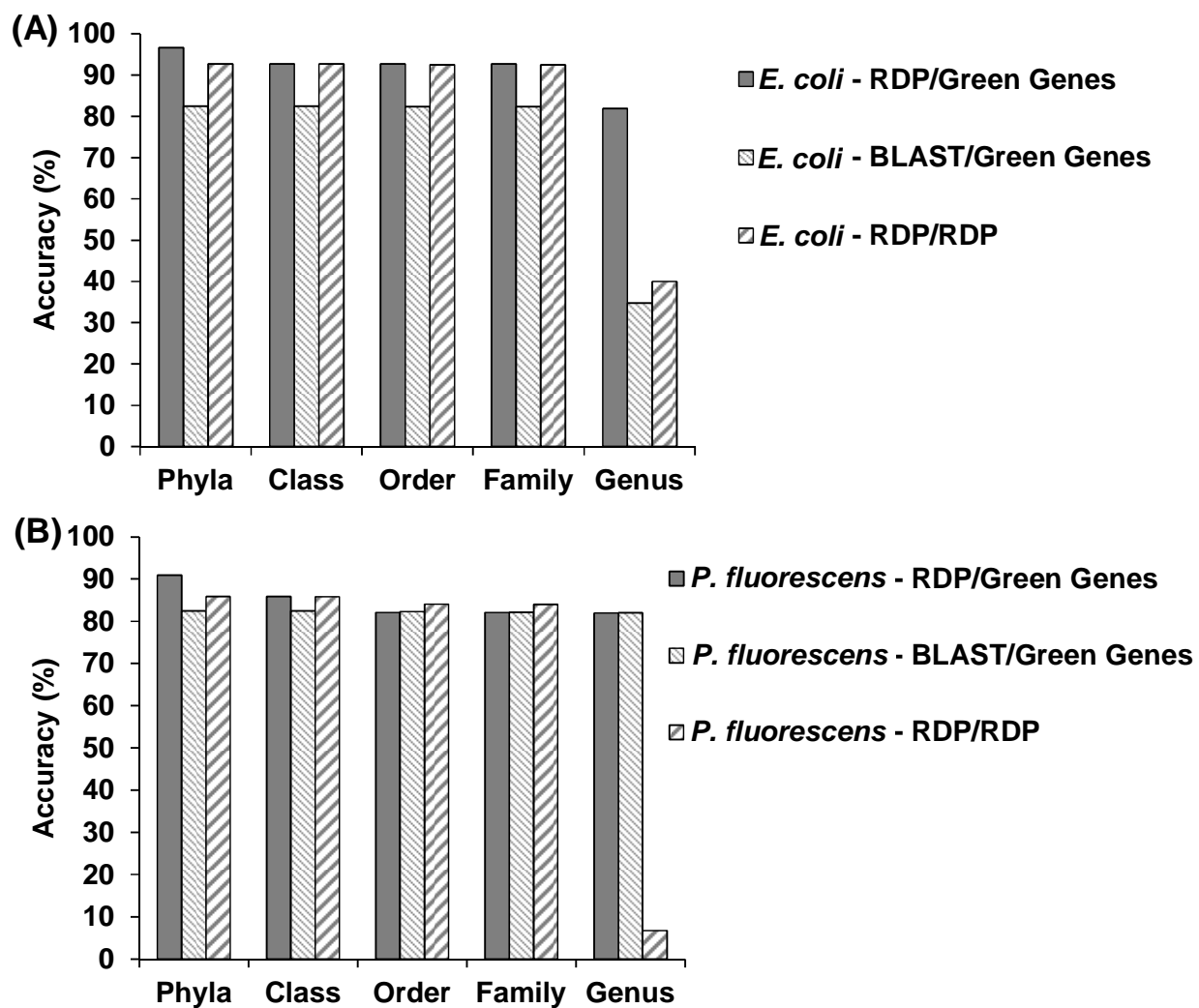


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398 Figure 1.

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402 Figure 2.

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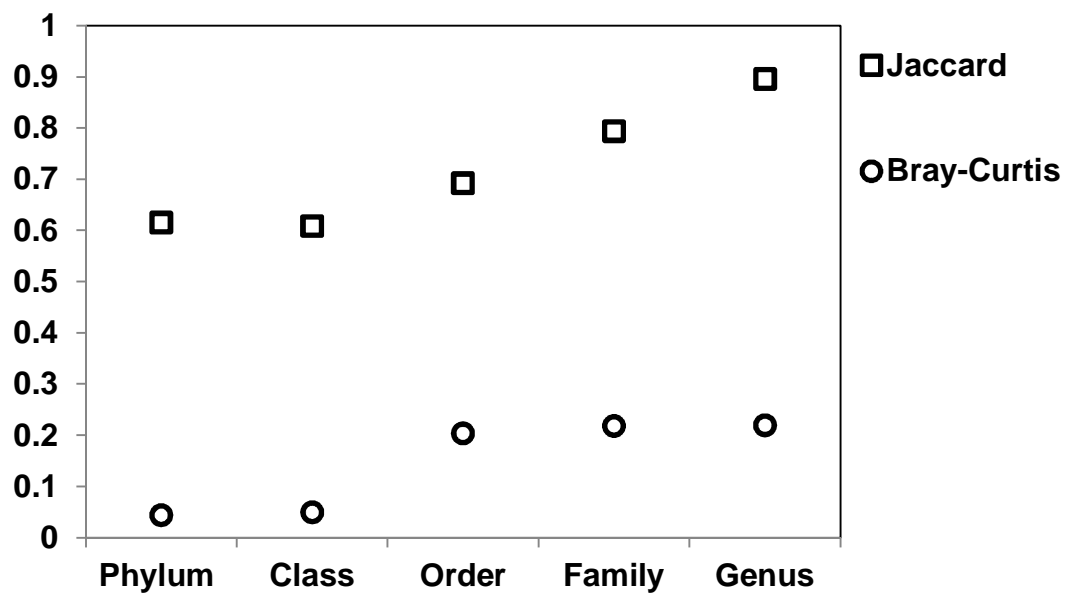
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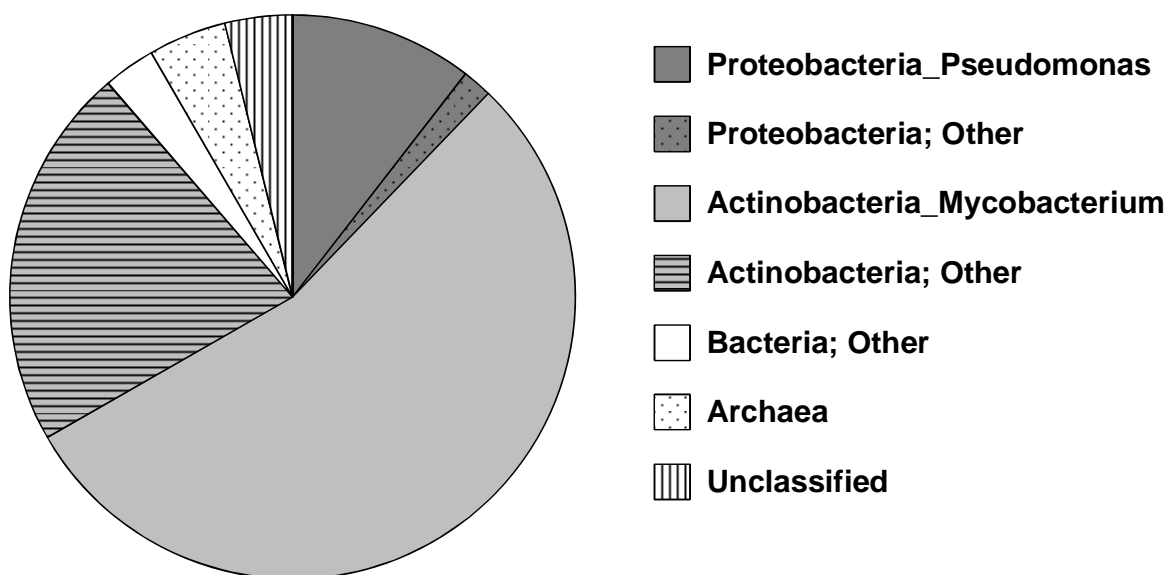
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410 Figure 3.

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413 Figure 4.

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