

doi: 10.1093/gigascience/gix043 Advance Access Publication Date: 12 June 2017 Research

RESEARCH

Multi-locus and long amplicon sequencing approach to study microbial diversity at species level using the MinIONTM portable nanopore sequencer

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Abstract

The miniaturized and portable DNA sequencer MinIONTM has demonstrated great potential in different analyses such as genome-wide sequencing, pathogen outbreak detection and surveillance, human genome variability, and microbial diversity. In this study, we tested the ability of the MinIONTM platform to perform long amplicon sequencing in order to design new approaches to study microbial diversity using a multi-locus approach. After compiling a robust database by parsing and extracting the rrn bacterial region from more than 67 000 complete or draft bacterial genomes, we demonstrated that the data obtained during sequencing of the long amplicon in the MinIONTM device using R9 and R9.4 chemistries were sufficient to study 2 mock microbial communities in a multiplex manner and to almost completely reconstruct the microbial diversity contained in the HM782D and D6305 mock communities. Although nanopore-based sequencing produces reads with lower per-base accuracy compared with other platforms, we presented a novel approach consisting of multi-locus and long amplicon sequencing using the MinION[™] MkIb DNA sequencer and R9 and R9.4 chemistries that help to overcome the main disadvantage of this portable sequencing platform. Furthermore, the nanopore sequencing library, constructed with the last releases of pore chemistry (R9.4) and sequencing kit (SQK-LSK108), permitted the retrieval of the higher level of 1D read accuracy sufficient to characterize the microbial species present in each mock community analysed. Improvements in nanopore chemistry, such as minimizing base-calling errors and new library protocols able to produce rapid 1D libraries, will provide more reliable information in the near future. Such data will be useful for more comprehensive and faster specific detection of microbial species and strains in complex ecosystems.

Keywords: MinION; nanopore sequencer; ribosomal operon; long amplicon sequencing; microbial diversity; long-read sequencing

Background

During the last 2 years, DNA sequencing based on singlemolecule technology has completely changed the perception of genomics for scientists working in a wide range of scientific fields. This new perspective is not only supported by the technology itself but also by the affordability of these sequencing instruments. In fact, unprecedentedly, Oxford Nanopore Technologies (ONT) released the first miniaturized and portable DNA

Received: 29 September 2016; Revised: 8 March 2017; Accepted: 9 June 2017

© The Author 2017. Published by Oxford University Press. This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0/), which permits unrestricted reuse, distribution, and reproduction in any medium, provided the original work is properly cited. sequencer in early 2014, within the framework of the MinIONTM Access Programme. Recently, the MinION Analysis and Reference Consortium (MARC) has published results related to the study of the reproducibility and global performance of the MinIONTM platform. These results indicate that this platform is susceptible to a large stochastic variation, essentially derived from the wet-lab and MinIONTM operative methods, but also that variability has minimal impact on data quality [1].

The coordinated and collaborative work and mutual feedback between industry and the scientific community have enabled ONT to develop rapidly toward improving its portable platform for DNA sequencing, minimizing the stochastic variation during DNA library preparation. Consequently, in late Autumn 2015, ONT released MkIb, the latest version of MinION[™], and in April 2016 the fast mode chemistry (R9) was released, increasing the rate of sensing DNA strands from 30–70 to 280–500 bp/sec and reaching up to 95% of per-base accuracy in 2D reads (Clive G. Brown, CTO ONT, personal communication).

One of the most attractive capabilities of the MinIONTM platform is the sequencing and assembly of complete bacterial genomes using exclusively nanopore reads [2] or through hybrid approaches [3, 4]. Notwithstanding, the MinION[™] platform has also been demonstrated as useful in other relevant areas, including human genetic variant discovery [5, 6], detection of human pathogens [7, 8], detection of antibiotic resistance [9, 10], and microbial diversity [11, 12]. Regarding the latter, microbial diversity and taxonomic approaches are common and in high demand to analyse the microbiota associated with a wide variety of environment- and human-derived samples. However, these analyses are greatly limited by the short-read strategies commonly employed. Thanks to improvements in the chemistry of the most common popular sequencing platforms in recent years, it is now possible to characterize microbial communities in detail, down to the family or even genus level, using genetic information derived from roughly 30% (~500nt) of the full 16S ribosomal RNA (rRNA) gene. Despite the massive coverage achieved with short-read methods, the limitation in terms of read length means taxonomic assignment at the species level is still unfeasible. For instance, taxonomy strategies based on short reads from the Illumina MiSeq platform offer limited information that underestimates the microbial diversity of complex samples when compared with alternative approaches based on long DNA reads [13]. Consequently, implementation of longread sequencing approaches to study larger fragments of marker genes will permit the design of new studies to provide evidence for the central role of precise bacterial species/strains in a great variety of microbial consortia. Recent studies at this regard have showed important advances in taxonomy analysis using long reads generated by single molecule technologies [11, 14, 15], indicating that the expansion or inclusion of more hypervariable regions in the analysis overcomes the disadvantage of working with error-prone DNA reads. With respect to the above, we have recently explored the performance of the MinIONTM device. Our study demonstrates that data obtained from sequencing nearly full-length 16S rRNA gene amplicons are effective in the study of microbial communities through nanopore technology [11]. We wanted to move a step forward in this type of strategy, thus gaining more specificity when including several hypervariable markers in the analysis, at sequence and structural levels, by designing a multi-locus and long amplicon sequencing method to study microbial diversity. At the same time, we also wanted to explore the affordability of $MinION^{TM}$ technology to perform microbial diversity analyses by multiplexing several samples in 1 single MinION[™] flowcell. Accordingly, here we present a study of the 16S, 23S, and the internal transcribed spacer (ITS; which frequently encodes tRNA genes) simultaneous sequencing, using the MinIONTM MkIb device and R9 chemistry, with prior generation of ~4.5 kb DNA fragments by amplifying the nearly full-length operon encoding the 2 larger ribosomal RNA genes in bacteria, the *rrn* region (*rrn* hereinafter). We have studied the *rrn* of 2 mock microbial communities, composed of genomic DNA from 20 and 8 different bacterial species, obtained, respectively, from BEI Resources and ZYMO Research Corp., using the MinIONTM sequencing platform in multiplex configuration.

Data Description

The R9 raw data collected in this experiment was obtained as fast5 files using MinKNOWTM v. 0.51.3.40 (Oxford Nanopore Technologies) after conversion of electric signals into base calls via the MetrichorTM agent v. 2.40.17 and the Barcoding plus 2D Basecalling RNN for SQK-NSK007 workflow v. 1.107, whereas the R9.4 raw data was generated by MinKNOWTM v. 1.5.5 (Oxford Nanopore Technologies) with the respective local basecalling algorithm implemented in that version of the MinIONTM controller software. Base-called data passing quality control and filtering were downloaded, and data were converted to fasta format using the *poRe* package [16]. Fast5 raw data can be accessed at the European Nucleotide Archive under the project ID PRJEB15264. Only 2 data sets were generated after a sequencing run of MinIONTM MkIb.

Analysis

Defining the arrangement of the rrn region

The complete or partial gene sequence of the RNA attached to the small subunit of the ribosome is classically used to perform taxonomy and diversity analysis in complex samples containing hundreds of microbial species. In the case of bacterial species, the 16S rRNA gene is the most widely used DNA marker for taxonomic identification of a particular species, given the relatively high number of hypervariable regions (V1 to V9) present across its sequence. Nowadays, it is possible to study the complete or almost full-length sequence of the 16S rRNA molecule thanks to single-molecule sequencing approaches [11, 14, 15, 17]. The identification of complex microbial communities at the species level with raw data obtained from MinIONTM or PacBIO platforms is improving; however, uncertainty in taxonomic assignation is still noteworthy given the high proportion of errors in their reads. While future technical advances may improve the quality of DNA reads generated by third-generation sequencing devices, new strategies can also be adopted to enhance the performance of these approaches. Consequently, we postulate that a good example of this is to study a common multi-locus region of the bacterial genome, which enables the simultaneous study of more variable regions and locus arrangements (sequence and structural variation), such as the operon encoding the ribosomal RNA. Using a complex sample where hundreds of microbial species are potentially present (DNA from human faeces), we carried out preliminary experiments to amplify the rrn. We observed that from the hypothetical configurations envisaged for the rrn (Fig. 1A), we only obtained a clear amplification using the primer pairs S-D-Bact-0008-c-S-20 and 23S-2241R, indicating that the rrn preferentially seems to be transcriptionally arranged as follows: 16S-ITS-23S. A detailed evaluation of the fragment size determined that the main polymerase chain reaction (PCR) products ranged from 4.3 to 5.4 kbp (Fig. 1B-D),



Figure 1: Organization of the *rm* region in bacteria. (A) Hypothetical transcriptional arrangements expected for *rm* and tested experimentally using 2 sets of primer pairs (see small arrows drawn in each configuration). (B) Agarose gel electrophoresis of PCR reactions performed under the 2 hypothetical arrangements of *rm*; lanes: (i) 1 kb ruler (Fermentas), (ii) PCR reaction from the top configuration in (A), (iii) PCR reaction from the bottom configuration in (A). The GelAnalyser Java application was used to perform the band size analysis of the 1 kb ruler standard (C) and the amplicons obtained from human faecal DNA (D).

consistent with the expected size of PCR products amplifying the 16S, ITS, and 23S regions from several microbial species. The next step involved designing a multiplex sequencing approach to try to analyse more than 1 sample per sequencing run in 1 flowcell of MinIONTM; therefore, the primers were re-designed to include a distinctive barcode region at 5′ (Table 1). During PCR of the *rrn*, we tagged the amplicon derived from the mock community HM782D with the barcode *bc01* in a dual manner, whereas the amplicons derived from sample D6305 were tagged with barcode *bc08* in similar way. Parallel experiments were conducted on HM782D and D6305 DNA, with comparative aims, using a conventional protocol of microbial diversity analysis and consisting of V4-V5 16S amplicon sequencing using the Illumina MiSeq paired-end approach (see the Methods section).

The reference database

One of the major handicaps when proposing this new rm region to be used for taxonomy analysis is the need to compile a reference database to compare the reads produced by the MinIONTM device. Therefore, we proceeded to parse the genetic information of over 67 000 bacterial genomes, whose sequences are publicly available in GenBank at the National Center for Biotechnology Information database. In this way, we retrieve and compile more than 47 000 rm sequences that were subjects of a clustering analysis to reduce the level of redundancy and to disclose the variability intrinsically associated with the rm itself and with its individual components as well (Fig. 2).

After normalization of cluster numbers against the median size of the respective regions analysed and referenced against the numbers obtained for the 16S region at 97% sequence identity, we found that the rrn region, comprising the 16S, ITS, and 23S coding regions, exhibits more than 4-fold more variation than that observed for the 16S molecule alone (at 100% sequence identity). As expected, the 23S region exhibited more diversity by containing more hypervariable regions than the 16S region and getting almost 2-fold more diversity. Strikingly, the ITS regions showed similar levels of genetic diversity despite having almost one-fourth the size of the 16S region on average. When parsing the genetic information of over 67 000 bacterial genomes, we observed that the ITS region frequently encodes 1 or several tRNA genes and it possess high variability in terms of length as well. Consequently, the variability observed in the rrn was the largest observed and thought to be meaningful for the aims of this study. We obtained data supporting the above notion by searching the number of rrn clusters (at 100% identity) matching the most predominant species in the database, thus retrieving 1713, 1276, and 1273 rrn clusters annotated for Escherichia coli, Streptococcus pneumoniae, and Staphylococcus aureus, respectively. In consequence, the rrn is able to accumulate enough sequence variability to discern taxonomy even at the strain level.

Performance of the R9 chemistry

Once we could compile a reference database for comparison aims, we proceeded with the amplicon library construction and

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Sample	Barcode	Primer	Barcode extended ^a
HM-782D	(bc01) GGTGCTGAAGAAAGTTGTCGGGTGTCTTTGTGTGTTAACCT	(S-D-Bact-0008-c-S-20) AGAGTTTGATCMTGGCTCAG	(bc01F) GGTGCTGAAGAAGTTGTCGGTGTCTTTGTGTTAACCTAGAGTTTGATCMTGGCTCAG
	(bc01)	(23S-2241R)	(bc01R)
	GGTGCTGAAGAAGTTGTCGGTGTCTTTGTGTTAACCT	ACCGCCCCAGTHAAACT	<u>GGTGCTGAAGAAGTTGTCGGTGTCTTTGTGTTAACCT</u> ACCGCCCCAGTHAAACT
D6503	(bc08)	(S-D-Bact-0008-c-S-20)	(bc08F)
	GGTGCTGTTCAGGGAACAAACCAAGTTACGTTTAACCT	AGAGTTTGATCMTGGCTCAG	GGTGCTGTTCAGGGAACAACCAAGTTACGTTTAACCTAGAGTTTTGATCMTGGCTCAG
	(bc08)	(23S-2241R)	(bc08R)
	GGTGCTGTTCAGGGAACAAACCAAGTTACGTTTAACCT	ACCGCCCCAGTHAAACT	<u>GGTGCTGTTCAGGGAACAAACCAAGTTACGTTTAACCT</u> ACCGCCCAGTHAAACT
Other primer	s used		
Human		(S-D-Bact-1391-a-A-17)	
fecal DNA		GACGGGCGGTGWGTRCA	
		(23S-129F)	
		CYGAATGGGRVAACC	

Table 1: Barcodes and primers used to generate amplicon libraries





Figure 2: Variability of the rrn region and its functional domains. The rrn database, compiled after parsing more than 67 000 draft and complete bacterial genomes, was assessed by clustering analysis at different levels of sequence identity: 97% (white bars), 98% (light grey bars), 99% (dark grey bars), and 100% (black bars). For comparative aims, the functional DNA sequences encoded into the rrn region were also individually studied. The normalized diversity (y-axis) resulted from calculating the number of clusters obtained for each analysis, normalized with the median sizes of the respective regions in terms of kb and referenced against the value obtained for 16S sequences clustered at 97%, the canonical threshold for species assignment.

sequencing run, obtaining raw data consisting of 17 038 reads, almost all of which were classified as 1D reads. For general knowledge, the DNA reads derived from the MinIONTM device can be classified into 3 types: "1D template," "1D complement," and "2D" reads. 2D reads are products of aligning and merging sequences from the template (read from the leader adapter) and complement reads (a second adapter called a hairpin, or HP, adapter must be generated) produced from the same DNA fragment. These contain a lower error rate, owing to strand comparison and mismatch correction. In addition to the technical issues indicative of a bad ligation of the HP adapter, we obtained 93% of reads (~15 900 reads) during the first 16 hours of the run; thus, we obtained lower sequencing performance after re-loading with the second aliquot of the sequencing library and extended the run for another 24 hours (40 hours in sum). The fasta sequences were filtered by retaining those between 1500 and 7000 nt in length, obtaining at least enough sequence information to compare a DNA sequence equivalent to the 16S rRNA gene length. After this filtering step, we retained 72% of sequences (12 278) and then we performed the respective barcode splitting. For this purpose, we modified the default parameters of the "split_barcodes.pl" perl script (Oxford Nanopore Technologies) by incorporating the information of the extended barcodes (Table 1), rather than the barcode information alone, and simultaneously increased the stringency parameter to 25 (14 by default). Afterwards, the concatenation of the reads was obtained from respective forward and reverse extended barcodes; then we retrieved a total of 2019 (52% from forward and 48% from reverse barcodes) and 1519 (53% from forward and 47% from reverse barcodes) 1D reads for HM782D and D6305 mock communities, respectively. Read-mapping was performed against the rrn database, compiling more than 22 000 rrn regions, retrieved from more than 67 000 genomes available in GenBank (see Availability of supporting data). The taxonomy associated with the best hit based on the competitive alignment score followed by filtering steps (see the Methods) was used to determine the structure of each mock community. The MinIONTM sequencing data produced the microbial structure presented in Fig. 3 for the mock communities HM782D and D6305, respectively.

Fig. 3 shows the bacterial species and their respective relative proportions retrieved from the analysis of the mock communities HM782D and D6305, respectively. With respect to the HM782D mock community, we were able to recover 20 representative species, accounting for 16 out of 20 species present in that artificial community (Fig. 3A). However, the remaining 4 species that apparently are absent in this community have a close relationship to others detected correctly, e.g., Bacillus subtilis, Bacillus thuringensis, Bacillus anthracis, and Propionibacterium sp. Furthermore, we were unable to report the presence of just 4 species present in HM782D because proportions of Rhodobacter sphaeroides and Actinomyces odontolyticus were below the predominance threshold (1%), being present in 0.25% and 0.12%, respectively. Similarly, another 40 different species, but close to that present in the HM782D mock community (Bacillus spp., Streptococcus spp., Clostriudium spp., Neisseria spp., Staphylococcus spp., and Listeria spp.), had minor representation in data derived from rrn sequencing. With respect to the lower proportions of Rhodobacter sphaeroides and Actinomyces odontolyticus, we have previously demonstrated that the low levels of 16S reads are a consequence of amplification bias derived from the PCR reaction and not from sequencing itself [11]. In this case, the new primer pair used to generate the long amplicons would seem to work more efficiently than those previously used, but apparently they still present issues at the bacterial coverage level. When we revised the whole taxonomy contained in our rrn database, the compiling of non-rrn regions for Deinococcus radiodurans and Helicobacter pylori partially explained the lack of these species in HM782D analysed by the present approach. However, a new alignment process using individual 16S and 23S rRNA sequences obtained from GenBank and including those for D. radiodurans and H. pylori, respectively, demonstrated that at least D. radiodurans could be identified in a higher proportion than A. odontolyticus and R. sphaeroides, albeit in a lower proportion than our predominance threshold (data not shown). Regarding the results obtained from the D6305 mock community, we found a total of 10 bacterial species present in this mixed DNA sample; 8 of them matched the expected structure of the community, and additionally 18 close species had minor representation (Bacillus spp., Enterococcus spp., Klebsiella spp., Lactobacillus spp., Streptococcus spp., and Staphylococcus spp.). Using the MinION^{\rm TM} data, we were able to recover 100% of the species present in this sample and the 2 additional members identified also have a close relationship within the Bacillus genus, as observed in the HM782D sample (Fig. 3B). We have determined that the coverage needed to retrieve all expected species in a non-even mock community with an abundance above 1% is $\sim \times 13$ in terms of the number of species of that community.

When compared to reference values and proportions theoretically expected for the species present in the 2 mock communities, we observed some deviations that were greater in certain species. Particularly, in the HM782D sample, the lowest coverage biases were observed for Actinomyces odontolyticus (-5.36), Rhodobacter sphaeroides (-4.36), and Enterococcus faecalis (-2.04). This indicates that such species, in addition to D. radiodurans and H. pylori, are more difficult to detect with the primers and PCR used here. By contrast, Escherichia coli (1.79) seems to be preferentially amplified, given that this species exhibited the highest positive coverage bias value (Fig. 3C). We again found that coverage bias is linearly correlated with PCR products generated by quantifying E. coli, L. gasseri, and B. vulgatus amplicons (Pearson's R = 0.82, P = 0.047), data indicating that there are not major issues during taxonomy assignation by overrepresentation of certain species in the reference database. The values obtained for D6305 were more homogeneous, and the lowest coverage bias was observed for *Lactobacillus fermentum* (–2.18) (Fig. 3D). Additional analysis indicated that there was not significant correlation between coverage bias and GC content in *rrn*. Although the low coverage bias for some species can be solved by selecting another pair of primers, the ability to recover almost all of them, at least in a low proportion, in itself represents an important attribute of this approach for inter-sample comparisons. Interestingly, we observed a similar pattern of overrepresentation of *Bacillus* spp. sequences (>50%) in the D6305 sample but not for *Escherichia* spp. sequences (~4%) in the HM782D mock community when Illumina MiSeq data were assessed (Fig. 3C and D).

The high error rate of the 1D reads (ranging between 70% and 87% sequence identity, according to high-quality alignments) makes barcode de-multiplexing a difficult task in nanopore data. However, our results indicate that with the configuration and parameters presented here, we could efficiently distinguish the reads generated from HM782D and D6305 amplicons. As a consequence, the performance of this long amplicon approach to properly assigning microbial communities to samples was efficiently assisted by the parameters during the de-multiplexing process, which were central to discerning reads obtained from respective samples multiplexed in the MinION flowcell. For instance, the distribution of read matching with closely related species, such as Lactobacillus gasseri and Lactobacillus fermentum, contained distinctively in HM782D and D6305 samples, was indicative of the adequate execution of the de-multiplexing pipeline. The above was also exemplified for Salmonella enterica sequences that were determined only in D6305 despite its close relationship with E. coli at the 16S and 23S sequence levels (close to 100%). Regarding the latter, the multiple sequence alignment built with rrn regions from both species was inspected, directly distinguishing the ITS as the major source of variation between the 2 species (data not shown). Indeed, this was corroborated by the comparative analysis performed during the clustering step of the reference samples to create our rrn database.

Performance of R9.4 chemistry

During the course of the present work, the MinION R9.4 chemistry was delivered in Autumn 2016. Therefore, we wanted to perform a replicate experiment using this type of chemistry in order to disclose how much improvement our approach would gain in terms of sensibility and specificity. With only a 3-hour run, we observed a notable improvement of throughput and perbase accuracy and the MinIONTM produced almost 40 000 reads with a predominant QScore distribution between 8 and 12, suggesting a theoretical error rate of reads between 0.15 and 0.06, respectively, lower than obtained from R9 reads (0.25 to 0.15). After compiling all sequences in a fasta file, we proceeded to perform filtering in a more equal manner than previously done for R9 data. Consequently, we retained more than 33 000 reads (86%) for further processing and taxonomy assignment. The major results from comparison among R9 and R9.4 runs are summarized in Table 2. As expected, the R9.4 dataset was more accurate and its reads showed a lower per-base error rate; therefore, the taxonomy analysis based on these reads would be more precise than that observed with R9 reads. Globally, the results obtained from R9.4 chemistry are very similar to those observed with R9 chemistry, but the level of uncertainty was diminished by reducing the number of close species to that contained in respective



Figure 3: Microbial structure of the mock communities. (A and B) Microbial species and respective relative proportions determined to be present in the HM782D and D6305 mock communities, respectively, following the analysis of raw data obtained from *rm* amplicon sequencing in MinIONTM and chemistry R9. (C and D) Comparative analysis of the expected microbial species and proportions against the data obtained after mapping reads generated by a 4.5-kbp amplicon PCR and sequenced in the MinIONTM device with R9 and R9.4 chemistries, for HM782D and D6305, respectively. (E) Linear correlation analysis of relative read proportions obtained for all bacterial species present in HM872D and D6305 mock communities with R9 and R9.4 chemistries.

0.050 0.100 0.150 0.200 0.250

Relative proportion of R9 reads

0.000

Table 2: Basic stats comparison of R9 and R9.4 reads after processing.

	R9	R9.4
Total raw data	17 038 (100%)	39 216 (100%)
Reads >1.5 kb	12 278 (72%)	33 764 (86%)
Read length distribution	25th percentile = 2847 nt, median = 3303 nt;	25th percentile = 3730 nt, median = 3976 nt;
	75th percentile = 3754	75th percentile = 4135 nt
Reads aligned (bc1 $+$ bc8)	3227 (19%)	14 392 (43%)
Alignment identity distribution	25th percentile = 66.5%, median = 69%; 75th	25th percentile = 81%, median = 85%; 75th
	percentile = 73%	percentile = 87%
Maximum identity	86.7%	92%

mock communities exhibiting very low abundance (<1%), thus decreasing from 40 species to 15 for the HM782D and from 18 to 16 for the D6305. We were unable again to recover D. *radiodurans* and H. *pylori* reads, but we improved the sensitivity for A. *odontolyticus* and R. *sphaeroides* (Fig. 3C and D), whose relative proportions were almost duplicated in R9.4 data (R. *sphaeroides* = 0.44%, A. *odontolyticus* = 0.31%). We compared the respective proportions obtained from R9 and R9.4 chemistries obtaining consistent results (Fig. 3E), indicating that our approach is reproducible with no major changes despite the different chemistry and kits for library preparation using during both sequencing runs.

Comparison with Illumina MiSeq data

The Illumina MiSeq data obtained after sequencing the V4-V5 16S region permitted characterization of the genus distribution in the HM782D sample with the RDP classifier. As a result, we compiled distribution of all 17 genera represented in the HM782D mock community (Supplementary Material 1) and 4 additional genera with very low abundance (2 reads/8409 assigned). Moreover, when an operational taxonomic unit (OTU)picking approach was conducted, we recovered 41 OTUs whose identity was evaluated in the SILVA Incremental Aligner (SINA) server (Supplementary Material 2). Globally, we recovered taxonomy identification of all genera expected, but only 3 species were well identified based on the Greengenes taxonomy (H. pylori, P. acnes, and R. sphaeroides) whereas 1 was wrongly identified (Neisseria cinerea). For the D6305 mock community, we could recover the 8 different components at the genus level of this mock community (Supplementary Material 1), plus 7 additional unrelated genera with very low abundance (<7 reads/8046 assigned). At the OTU level, we retrieved a total of 14 sequences, whose taxonomy identification is presented in the Supplementary Material 3 document. In this case, only S. enterica could be identified at the species level. Given that data derived from this short-read approach normally cannot reach a reliable taxonomy assignment down to the species level, we proceeded to make comparisons with R9 and R9.4 data by compiling this last information to the genus level in order to evaluate the performance of our approach with a commonly used procedure. In the Supplementary Material 1 file and Table 3, a comparison in terms of the relative read proportion and coverage bias is depicted. We observed no larger deviations in data retrieved with MinION regarding those numbers obtained with conventional approaches such as the study of V4-V5 regions with the MiSeq platform. Interestingly, we observed a similar pattern of important negative coverage bias in all 3 approaches for Actinomyces spp., Enterococcus spp., and Rhodobacter spp. species in the HM782D community and for Lactobacillus spp., and Listeria spp. in the D6305 community, suggesting that species of such genera are equally

underrepresented no matter the type of amplicon, sequencing platform, or sequencing chemistry of study. Conversely, only the Bacillus spp. from the D6305 exhibited large positive coverage bias values in all 3 approaches. Globally, all methods compared to study microbial communities at this level have a pattern of underrepresentation for all species present in the mock communities given the average and median values obtained. Moreover, the MiSeq V4-V5 approach also showed important coverage bias, indicating that this issue is not strictly associated with the MinIONTM-based approach presented here and that it is probably inherent to the amplification process of target DNA. Finally, correlation tests indicate that despite the coverage bias observed, all configurations used to study the mock communities replicate fairly well the composition of the mock communities and that data obtained from R9 and R9.4 experiments show a slight improvement in this regard, with no major differences when compared with data from the MiSeq platform (Table 3).

Discussion

The inventory of microbial species based on 16S DNA encoding for ribosomal RNA sequencing is frequently used in biomedical research to determine microbial organisms inhabiting the human body and their relationship with disease. Recently, thirdgeneration DNA sequencing platforms have developed rapidly, facilitating the identification of microbial species and overcoming the read length issues inherent to second-generation sequencing methods. These advances allow researchers to infer taxonomy and analyse diversity from the almost full-length bacterial 16S rRNA sequence [11, 14, 15, 17]. Particularly, the ONT platform deserves special attention, given its portability and its fast development since $MinION^{TM}$ became available in 2014. Notwithstanding, this technology is susceptible to a large stochastic variation, essentially derived from wet-lab methods [1]. We corroborated this issue by obtaining a sequencing run where the raw data predominantly consisted of 1D reads as a consequence of the HP adapter ligation failure, despite following the manufacturer's instructions. However, we were able to develop an efficient analysis protocol where the higher read quality offered by R9 chemistry and the updated Metrichor basecaller protocol proved pivotal to obtaining 1D reads with a range of identity between 70 and 86%, with sufficient per-base accuracy to successfully perform the taxonomic analyses described herein. Moreover, during the course of this study, the R9.4 flowcells were released and we were able to replicate our approach using this improved pore chemistry and the SQK-LSK108 for 1D libraries obtaining reads with a sequence identity of up to 92%.

Our preliminary results indicated that the *rm* region in bacteria preferentially has a unique conformation (with the transcriptional arrangement of 16S-ITS-23S), and we could amplify this ~4.5-kbp region with the selected S-D-Bact-0008-c-S-20 and

	Relative read proportion				Coverage bias		
Genera HM782D	Reference	PCR+MiSeq	PCR+MinION-R9	PCR+MinION-R9.4	PCR+MiSeq	PCR+MinION-R9	PCR+MinION-R9.4
Acinetobacter spp.	0.050	0.019	0.046	0.032	-1.42	-0.12	-0.62
Actinomyces spp.	0.050	0.010	0.001	0.003	-2.36	-5.36	-4.02
Bacillus spp.	0.050	0.017	0.102	0.045	-1.57	1.03	-0.16
Bacteroides spp.	0.050	0.106	0.059	0.037	1.08	0.25	-0.42
Clostridium spp.	0.050	0.125	0.027	0.032	1.32	-0.91	-0.66
Deinococcus spp.	0.050	0.109	0.000	0.000	1.12	ND	ND
Enterococcus spp.	0.050	0.022	0.012	0.013	-1.17	-2.04	-1.93
Escherichia/Shigella spp.	0.050	0.038	0.172	0.209	-0.38	1.79	2.07
Helicobacter spp.	0.050	0.040	0.000	0.000	-0.32	ND	ND
Lactobacillus spp.	0.050	0.065	0.051	0.068	0.38	0.03	0.45
Listeria spp.	0.050	0.024	0.074	0.133	-1.04	0.57	1.41
Neisseria spp.	0.050	0.099	0.064	0.056	0.98	0.36	0.17
Propionibacterium spp.	0.050	0.021	0.079	0.097	-1.25	0.66	0.96
Pseudomonas spp.	0.050	0.038	0.018	0.079	-0.39	-1.46	0.67
Rhodobacter spp.	0.050	0.013	0.002	0.004	-1.91	-4.36	-3.51
Staphylococcus spp.	0.100	0.037	0.125	0.086	-1.44	0.32	-0.22
Streptococcus spp.	0.150	0.217	0.115	0.093	0.53	-0.38	-0.69
Genera D6305							
Bacillus spp.	0.139	0.574	0.383	0.340	2.04	1.46	1.29
Enterococcus spp.	0.088	0.078	0.057	0.082	-0.17	-0.64	-0.11
Escherichia/Shigella spp.	0.113	0.035	0.167	0.137	-1.67	0.56	0.28
Lactobacillus spp.	0.198	0.104	0.049	0.049	-0.93	-2.01	-2.01
Listeria spp.	0.163	0.046	0.080	0.118	-1.82	-1.03	-0.46
Pseudomonas spp.	0.058	0.060	0.038	0.039	0.05	-0.62	-0.56
Salmonella spp.	0.115	0.049	0.099	0.138	-1.22	-0.22	0.26
Staphylococcus spp.	0.126	0.051	0.094	0.095	-1.30	-0.42	-0.41
Average	0.080	0.080	0.077	0.079	-0.51	-0.55	-0.36
Median	0.050	0.048	0.062	0.074	-0.72	-0.30	-0.29
Min	0.050	0.010	0.000	0.000	-2.36	-5.36	-4.02
Max	0.198	0.574	0.383	0.340	2.04	1.79	2.07
Pearson's R ^a (P-value) Pearson's R ^b (P-value)	-	0.39 (0.0504) –	0.43 (0.0306) 0.73 (0.0001)	0.41 (0.0417) 0.64 (0.0005)			

Table 3: Comparative analysis among data generated from MiniON and MiSeq platfor
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^aPearson's R calculated from comparisons of R9, R9.4, and MiSeq data with reference proportions, respectively.

^bPearson's R calculated from comparisons of R9 and R9.4 data with MiSeq output, respectively.

23S-2241R primer pair. Once we were able to distinguish the feasibility of amplifying the *rrn*, our approach comprised the study of 2 different mock communities in a multiplex manner, to be combined into 1 single MinIONTM flowcell. By designing the respective forward and reverse primers tagged with specific barcodes recommended by ONT, we were able to retrieve extended barcode-associated reads, in spite of the large proportion of perbase errors contained in these types of reads. Using MinIONTM data based on multi-locus markers and long amplicon sequencing, we could reconstruct the structure of 2 commercially available mock communities. Although the expected proportions of some species in each community exhibited an important coverage bias, we were able to recover 80% (HM782D) and 100% (D6305) of bacterial species from the respective mock communities. Consequently, future analyses should be conducted to find an appropriate PCR approach using primers with a higher coverage for bacterial species.

We have analysed a great amount of genetic information, with the aim of compiling a valuable database containing the genetic information for the *rm* present in over 67 000 draft and complete bacterial genomes. The global length distributions in the region indicated that the *rm* was 4993 \pm 187 bp in length whereas the 16S, ITS, and 23S sub-regions were 1612 \pm 75, 488 \pm 186, and 3036 \pm 160 bp in length, respectively. Using

this genetic information of the rrn, clustered at 100% sequence identity, enabled us to establish a multi-locus marker able to discriminate the taxonomy of 2 mock communities containing very close species. The latter was possible given that simultaneous analysis of the 16S, ITS, and 23S molecules offered almost 40-fold more diversity than studying the 16S, ITS, or 23S sequences separately and at 97% sequence identity. Moreover, the ITS was distinguished individually as an important variable genetic region in terms of sequence and length. Furthermore, it contributes notably to the higher variability observed in the rrn region, a fact evidenced in previous studies [18-21]. The accumulation of a larger number of variable sites in the *rrn* region, together with the particular structural variation of the ITS to potentially accommodate and encode tRNA genes, is thought to be central to discriminating bacterial species, despite the large proportion of per-base errors contained in MinIONTM reads. Our data indicate that our MinION reads produce alignments with averaged lengths of 2463 and 3191 bases for HM782D and D6305, respectively, using R9 chemistry and 4173 and 4115 bases for HM782D and D6305, respectively, using R9.4 chemistry. Consequently, the taxonomy assignment was predominantly based on the variability of more than 2 out of the 3 markers included in the rrn, no matter if reads were produced from the 16S or 23S edges of rrn amplicons. We expect that this type of analysis will likely

become more accurate over time as nanopore chemistry improves in the near future, with the concomitant increase in throughput, which is pivotal to disclosing the hundreds of species present in complex microbial communities for analysis in human or environmental studies. Therefore, the multilocus, long, and multiplex methods described here represent a promising analysis routine for microbial and pathogen identification, relying on the sequence variation accumulated in approximately 5 kbp of DNA, roughly accounting for the assessment of 1.25% of an average bacterial genome (~4 mbp). Notwithstanding, we cannot obviate that the current state of this approach presents some limitations in terms of the completeness of the rrn database created as well as the efficiency of the primers used to generate the long amplicons that have to be revisited in order to improve and increase the coverage of bacterial species. To date, our database includes rrn sequences from 2479 different species grouped into 918 different genera. In consequence, urgent studies must be undertaken to generate a more complete database that includes the rrn genomic information from species inhibiting complex and real samples such as those derived from that human body.

Methods

Bacterial DNA and rrn amplicons

The complex DNA sample for preliminary studies of rrn region arrangement consisted of DNA isolated from faeces, kindly donated by a healthy volunteer upon informed consent. An aliquot of 200 mg of human faeces was used to isolate microbial DNA using the QIAamp DNA Stool Mini Kit (Qiagen, Venlo, The Netherlands), following the manufacturer's instructions. Finally, DNA was eluted in 100 μ L nuclease-free water and a DNA aliquot at 20 ng/ μ L was prepared for PCR reaction using the primer pairs S-D-Bact-0008-c-S-20 and 23S-2241R or 23S-129F and S-D-Bact-1391-a-A-17 for testing configurations, shown in Fig. 1A (Table 1). The band size was analysed using the Java-based GelAnalyzer tool [22]. Genomic DNA for the reference mock microbial communities was kindly donated by BEI Resources [23] and ZYMO Research Corp. [24]. The composition of the mock communities was as follows: (i) HM782D is a genomic DNA mixture of 20 bacterial species containing equimolar ribosomal RNA operon counts (100 000 copies per organism per μ L), as indicated by the manufacturer and (ii) ZymoBIOMICS Cat No. D6305 (D6305 hereinafter) is a genomic DNA mixture of 8 bacterial species (and 2 fungal species) presented in equimolar amounts of DNA. According to the manufacturer's instructions, 1 μ L of DNA from each mock community was used to amplify all the genes contained in the rrn. DNA was amplified in triplicate by 27 PCR cycles at 95°C for 30 seconds, 49°C for 15 seconds, and 72°C for 210 seconds. Phusion High-Fidelity Taq Polymerase (Thermo Scientific, Waltham, MA, USA) and the primers S-D-Bact-0008-c-S-20 (mapping on 5' of 16S gene) and 23S-2241R (mapping on 3' of 23S gene) target a wide range of bacterial 16S rRNA genes [25, 26]. For the Illumina MiSeq sequencing, the V4-V5 hypervariable regions from the bacterial 16S rRNA gene were amplified using 1 $\mu \rm L$ of DNA from each mock community and 25 PCR cycles at 95°C for 20 seconds, $40^{\circ}C$ for 30 seconds, and $72^{\circ}C$ for 20 seconds. Phusion High-Fidelity Taq Polymerase (Thermo Scientific, Waltham, MA, USA) and the 6-mer barcoded primers [S-D-Bact-0563-a-S-15 (AYTGGGYDTAAAGNG) and S-D-Bact-0907-a-A-20 (CCGT-CAATTYMTTTRAGTTT)] were used to generate \sim 380bp dual barcoded PCR amplicons. As we wished to multiplex the sequencing of both mock communities into 1 single $\mathsf{MinION}^{\mathsf{TM}}$ flowcell, we designed a dual-barcode approach, where respective primers

were synthesized and fused with 2 different barcodes recommended by ONT (Table 1). Amplicons consisted of ~4.5 kbp of blunt-end fragments for the MinIONTM approach and ~380 bp for the Illumina MiSeq approach, and those were purified using the Illustra GFX PCR DNA and Gel Band Purification Kit (GE Healthcare, Little Chalfont, UK). Amplicon DNA was quantified using a Qubit 3.0 fluorometer (Life Technologies, Carlsbad, CA, USA). Quantification of certain PCR products to correlate with sequencing coverage bias by qPCR was assessed as previously described [11].

Amplicon DNA library preparation

The Genomic DNA Sequencing Kit SQK-MAP006 was ordered from ONT and used to prepare the amplicon library for loading into MinIONTM. Approximately 0.9 μ g of amplicon DNA (0.3 per mock community plus 0.3 μ g of an extra query sample consisting of amplicons obtained from a genomic DNA mix of several uncharacterized microbial isolates) were processed for end repair using the NEBNextUltra II End Repair/dA-tailing Module (New England Biolabs, Ipswich, MA, USA), followed by purification using Agencourt AMPure XP beads (Beckman Coulter, Brea, CA, USA) and washing twice with 1 volume of fresh 70% ethanol. Subsequently, and according to the manufacturer's suggestions, we used 0.2 pmol of the purified amplicon DNA (~594 ng, assuming fragments 4.5 kbp in length) to perform the adapter ligation step. Ten μ L of adapter mix, 2 μ L of HP adapter, and 50 μ L of Blunt/TA ligase master mix (New England Biolabs, Ipswich, MA, USA) were added in that order to the 38 μ L end-repaired amplicon DNA. The reaction was incubated at room temperature for 15 minutes, and 1 $\mu\mathrm{L}\,\mathrm{HP}$ Tether was added and incubated for an additional 10 minutes at room temperature. The adapter-ligated amplicon was recovered using MyOne C1-beads (Life Technologies, Carlsbad, CA, USA) and rinsed with washing buffer provided by the Genomic DNA Sequencing Kit SQK-MAP006 (Oxford Nanopore Technologies, Oxford, UK). Finally, the sample was eluted in the MyOne C1-beads by adding 25 μ L of elution buffer and incubating for 10 minutes at 37°C before pelleting in a magnetic rack. The R9.4 sequencing library was obtained by processing 600 ng of purified amplicon DNA (0.15 per mock community plus 0.15 μg of 2 extra query samples consisting of amplicons obtained from a genomic DNA mix of several uncharacterized microbial isolates) with SQK-LSK108 sequencing (Oxford Nanopore Technologies, Oxford, UK) for 1D reads, following the manufacturer's instructions. Briefly, the 600 ng of amplicon DNA diluted in 50 μ L nuclease-free water were processed for end repair using the NEBNextUltra II End Repair/dA-tailing Module (New England Biolabs, Ipswich, MA, USA), followed by purification using Agencourt AMPure XP beads (Beckman Coulter, Brea, CA, USA) and washing twice with 200 $\mu \rm L$ of fresh 70% ethanol. The ligation step was performed with 30 $\mu \rm L$ of DNA end-prepped, 20 $\mu \rm L$ adapter mix, and 50 μ L of Blunt/TA ligase master mix (New England Biolabs, Ipswich, MA, USA). The reaction was incubated at room temperature for 15 minutes. The adapter-ligated amplicon was recovered again using Agencourt AMPure XP beads (Beckman Coulter, Brea, CA, USA), washing twice with the ABB buffer supplied in the SQK-LSK1008 sequencing kit (Oxford Nanopore Technologies, Oxford, UK), and was eluted in Agencourt AMPure XP beads by adding 25 μ L of elution buffer and incubating for 10 minutes at 37°C before pelleting in a magnetic rack. Samples for the Illumina MiSeq approach were sent to the National Center for Genomic Analaysis (CNAG, Barcelona, Spain) for multiplex sequencing in 1 lane of MiSeq instrument with 2 \times 300 pairedend configuration.

Flowcell set-up

A brand new, sealed R9 flowcell was acquired from ONT and stored at 4°C before use. The flowcell was fitted to the MinIONTM MkIb prior to loading the sequencing mix, ensuring good thermal contact. The R9 flowcell was primed twice using 71 μ L premixed nuclease-free water, 75 μ L × 2 running buffer, and 4 μ L fuel mix. At least 10 minutes were required to equilibrate the flowcell before each round of priming and before final DNA library loading. For the replicate experiment, a R9.4 flowcell was fitted to the MinIONTM MkIb prior to loading the sequencing mix, ensuring good thermal contact. The R9.4 flowcell was primed with 800 μ L of running buffer (0.5 mL nuclerase-free water plus 0.5 mL RBF buffer). At least 10 minutes were required to equilibrate the flowcell, and then the remaining 200 μ L of running buffer were injected into the R9.4 flowcell with the SpotON port open.

Amplicon DNA sequencing

The sequencing mix was prepared with 59 μ L nuclease-free water, 75 μ L ×2 running buffer, 12 μ L DNA library, and 4 μ L fuel mix. A standard 48-hour sequencing protocol was initiated using MinKNOWTM v. 0.51.3.40. Base-calling was performed through data transference using the MetrichorTM agent v. 2.40.17 and the Barcoding plus 2D Basecalling RNN for SQK-NSK007 workflow v. 1.107. During the sequencing run, 1 additional freshly diluted aliquot of DNA library was loaded after 16 hours of initial input. The raw sequencing data derived from the 2 mock communities studied here were expected to account for two-thirds of the data produced by the R9 flowcell used. The R9.4 run was performed with a 75 μL DNA library (37.5 μL RBF buffer, 25.5 μL LLB, 12 $\mu \rm L$ DNA library) loaded into the R9.4 flowcell through the SpotON port. A standard 48-hour sequencing protocol was initiated using MinKNOWTM v. 1.5.5 with the respective local basecalling algorithm implemented in the $MinKNOW^{TM}$ software. The R9.4 raw data were generated during a sequencing run of only three hours.

The rrn database

We built a database containing the genetic information for the 16S and 23S rRNA genes and the ITS sequence in all the complete and draft bacterial genomes available in the National Center for Biotechnology Information database [27]. A total of 67 199 genomes were analysed by downloading the "fna" files and parsing for rRNA genes in the respective "gff" annotation file. Chromosome coordinates for rrn regions were parsed and used to extract DNA sequences from complete chromosomes or the DNA contigs assembled. The resulting rrn sequences were analysed, and the length distribution was assessed. We retrieved a total of 47 698 rrn sequences, with an average of 4993 nt in length. By selecting a size distribution equal to the 99th percentile (2-sided), we discarded potential incomplete or aberrant annotated rrn sequences and observed that rrn sequences can be found between 4196 and 5790 nt; under these boundaries, our rrn database finally accounted for a total of 46 920 sequences. Equivalent databases were built by parsing the respective rrn sequences with the RNammer tool to discriminate the 16S, ITS, and 23S rRNA sequences [28]. To remove the level of redundancy of our *rrn* database and to maintain the potential discriminatory power at the strain level, we performed clustering analysis using the USEARCH v. 8 tool for sequence analysis and the option -otu_radius_pct equal 0 [29], thus obtaining a total of 22 350 reference sequences. For comparative aims, the rrn database and the 16S, ITS, and 23S databases were also analysed using the option -otu_radius_pct with values ranging from 1 to 3. For access to the *rrn* database and the respective species annotation, see the Availability of supporting data section.

MinION data analysis

Read-mapping was performed using the LAST aligner v. 189 (LAST, RRID:SCR_006119) [30] with parameters -q1 -b1 -Q0 -a1 r1. Each 1D read was compared in a competitive way against the entire rrn database, and the best hit was selected by obtaining the highest alignment score. Alignment length and alignment coordinates in target and query sequences were parsed from the LAST output, and the sequence identity between matched regions was calculated using the python Levenshtein distance package. Iterative processing was used to determine thresholds for detection by evaluating the taxonomy distribution with read subsampling and different levels of sequence identity in top scored alignments. High-quality alignments were selected by filtering out those with identity values up to the 50th percentile of the distribution of identity values of all reads per sample (\sim 69%) in the R9 run. Therefore, taxonomy assignment was based exclusively on alignments with \geq 70% identity. For data derived from R9.4 chemistry, high-quality alignments were selected by filtering out those with identity values up to 25th percentile of the distribution, thus retaining alignments with \geq 81% identity. Basic stats, distributions, filtering, and comparisons were performed in R v. 3.2.0 [31]. For relative quantification of species, the singletons were removed, and the microbial species considered to be predominantly present in the mock communities were those with a relative proportion $\geq 1\%$, a value that has been demonstrated to be discriminative to always obtaining the expected microbial diversity during the iterative processing of alignments. The coverage bias was calculated by obtaining the fold-change (Log₂) of species-specific read counting against the expected (theoretical) average for the entire community according to information provided by the manufacturers.

Illumina data analysis

Fastq paired-end raw data were assembled using Flash software [32]. The HM782D and D6305 reads were de-multiplexed, and barcode and primers were removed using Mothur v. 1.36.1 (mothur, RRID:SCR_011947) [33]. The sequences were then processed for chimera removal using the Uchime algorithm [34] and SILVA reference set of 16S sequences [35]. A normalized subset of 10 000 sequences per sample was created by random selection after shuffling (×10 000) the original dataset. Taxonomy assessment was performed using the RDP classifier v. 2.7 [36]. The OTU-picking approach was performed with the normalized subset of 10 000 sequences, the uclust algorithm implemented in USEARCH v. 8.0.1623, the options -otu_radius_pct equal to 3 for clustering at 97%, and -minsize 2 for removing singletons [29]. The SINA server was used for taxonomy identification of OTUs recovered from Illumina MiSeq data [37].

Additional files

Supplementary data are available at GIGSCI online.

Supplementary Material 1. Comparison of MinION[™] and MiSeq outputs. Data obtained from Illumina MiSeq sequencing of the V4-V5 16S region from respective mock communities were compared with outputs from MinION[™] R9 and R9.4. Given that taxonomy identification of MiSeq reads at the species level only retrieved very few assignments, we compiled the species distribution of $MinION^{TM}$ data at the genus level.

Supplementary Material 2. MS Excel file compiling the output information retrieved from the SINA server (https://www.arb-silva.de/aligner/) for taxonomy identification of 41 OTUs derived from analysis of HM782D with the Illumina MiSeq approach. Information regarding sequence quality, identity percentage, mapping coordinates against the E. coli reference, length and taxonomy based on SILVA, Greengenes, and RDP databases is shown for all OTUs.

Supplementary Material 3. MS Excel file compiling the output information retrieved from the SINA server (https://www.arb-silva.de/aligner/) for taxonomy identification of 18 OTUs derived from analysis of D6305 with the Illumina MiSeq approach. Information regarding sequence quality, identity percentage, mapping coordinates against the *E.* coli reference, length and taxonomy based on SILVA, Greengenes, and RDP databases is shown for all OTUs.

Abbreviations

ITS: internal transcribed spacer; ONT: Oxford Nanopore Technologies; PCR: polymerase chain reaction; rRNA: Ribosomal RNA; *rrn*: the DNA region containing the 16S and 23S bacterial rRNA genes and its respective ITS region; SINA: SILVA Incremental Aligner.

Acknowledgements

This work and the contract to A.B.P. were supported by the European Union's Seventh Framework Program under the grant agreement number 613979 (MyNewGut).

Availability of supporting data

Accessions for the *rrn* database containing the reference sequences for alignments and taxonomic annotation are available at https://github.com/alfbenpa/rrn_db. The code source of the original split_barcodes.pl perl script is available at https://github.com/nanoporetech/barcoding/releases/tag/1.0.0 with ONT copyright. Other supporting data are also available from the *Giga*-Science database, GigaDB [38].

Competing interests

ABP is part of the MinIONTM Access Programme (MAP).

Authors' contributions

A.B.P. and Y.S. designed the study and managed the project. A.B.P. performed the experiments and analysed and managed the data. A.B.P. drafted the manuscript. Both authors read and approved the final manuscript.

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