

PacificBiosciences / pb-16S-nf

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Introduction

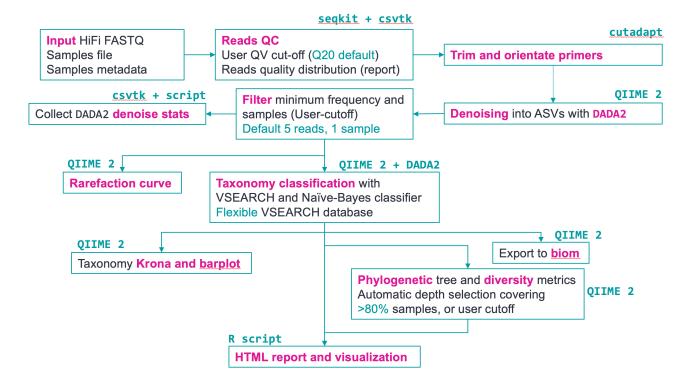
We describe here a new nextflow pipeline **pb-16S-nf** ¹ developed by **Khi Pin**, **Chua** (@proteinosome) as part of the Pacbio open code hosted on github and can be used to analyze data obtained with the Pacbio 16S method ². Khi Pin is actively developing this package further and was very helpful in deploying this code and correcting a few issues.

«This Nextflow pipeline is designed to process PacBio HiFi full-length 16S data into high quality amplicon sequence variants (ASVs) using **QIIME2** and **DADA2**. It provides a set of visualization through the QIIME 2 framework for interactive plotting. The pipeline generates a HTML report for the important statistics and top taxonomies» (taken from the github page).

The Nextflow pipeline depends on two text files and a matching folder of demultiplexed HiFi fastq files produced by the SMRTLink platform.

The pipeline performs a number pre-processing steps followed by DADA2 and Qiime2 commands. All of it integrated and standardized for ease of use.

The general workflow is shown in the next figure



¹https://github.com/PacificBiosciences/pb-16S-nf

 $^{^2} https://www.pacb.com/wp-content/uploads/Procedure-checklist-Amplification-of-bacterial-full-length-16S-rRNA-gene-with-barcoded-primers.pdf$



Theoretical composition of the Zymo mock community

In order to validate the workflow, we collected barcoded positive control samples from 5 Nucleomics Core 16S Sequel-IIe experiments and used them to compare pb-16S-nf results to the theoretical distribution present in the **Zymo mock community** ³

Avg. GC (%) Gram Stain gDNA Abun. (%) Pseudomonas aeruginosa 66.2 12 80% Escherichia coli 56.8 12 Salmonella enterica 52.2 12 Lactobacillus fermentum 52.8 12 Enterococcus faecalis 12 37.5 Staphylococcus aureus 32.7 12 Listeria monocytogenes 38.0 12 Bacillus subtilis 43.8 12 10% Saccharomyces cerevisiae 38.4 Yeast 2 ■ Cryptococcus neoformans 48.2 2 Yeast

Defined Microbial Community

The ZymoBIOMICS® Microbial Community Standard contains three easy-to-lyse bacteria, five toughto-lyse bacteria, and two tough-to-lyse yeasts.

Note: The copy number of rDNA genes in bacteria can vary a lot and contribute to 16S counts appear deferentially affected depending on the bacterial host. This is well documented for the mock community used in this experiment in the Zymo protocol document 4 .

Table	1:	Microbial	Composition

	Theoretical Composition (%)					
Species	Genomic DNA	16S Only ¹	16S & 18S¹	Genome Copy ²	Cell Number ³	
Pseudomonas aeruginosa	12	4.2	3.6	6.1	6.1	
Escherichia coli	12	10.1	8.9	8.5	8.5	
Salmonella enterica	12	10.4	9.1	8.7	8.8	
Lactobacillus fermentum	12	18.4	16.1	21.6	21.9	
Enterococcus faecalis	12	9.9	8.7	14.6	14.6	
Staphylococcus aureus	12	15.5	13.6	15.2	15.3	
Listeria monocytogenes	12	14.1	12.4	13.9	13.9	
Bacillus subtilis	12	17.4	15.3	10.3	10.3	
Saccharomyces cerevisiae	2	NA	9.3	0.57	0.29	
Cryptococcus neoformans	2	NA	3.3	0.37	0.18	

¹ The theoretical composition in terms of 16S (or 16S & 18S) rRNA gene abundance was calculated from theoretical genomic DNA composition with the following formula: 16S/18S copy number = total genomic DNA (g) × unit conversion constant (bp/g) / genome size (bp) × 16S/18S copy number per genome. Use this as reference when performing 16S targeted sequencing.

² The theoretical composition in terms of genome copy number was calculated from theoretical genomic DNA composition with the following formula: genome copy number = total genomic DNA (g) × unit conversion constant (bp/g) / genome size (bp). <u>Use this as reference when inferring microbial</u> abundance from shotgun sequencing data based on read depth.

³ The theoretical composition in terms of cell number was calculated from theoretical genomic DNA composition with the following formula: cell number = total genomic DNA (g) × unit conversion constant (bp/g) / genome size (bp)/ploidy.

 $^{^3} https://files.zymoresearch.com/datasheets/ds1706_zymobiomics_microbial_community_standards_data_sheet.pdf$

 $^{^4} https://files.zymoresearch.com/protocols/_d6305_d6306_zymobiomics_microbial_community_dna_standard.pdf$



Table 2: Strain Information

Species	NRRL Accession NO. ¹	Genome Size (Mb)	Ploidy	GC Content (%)	16/18S Copy Number	Gram Stain
Pseudomonas aeruginosa	B-3509	6.792	1	66.2	4	-
Escherichia coli	B-1109	4.875	1	46.7	7	-
Salmonella enterica	B-4212	4.760	1	52.2	7	-
Lactobacillus fermentum	B-1840	1.905	1	52.4	5	+
Enterococcus faecalis	B-537	2.845	1	37.5	4	+
Staphylococcus aureus	B-41012	2.730	1	32.9	6	+
Listeria monocytogenes	B-33116	2.992	1	38.0	6	+
Bacillus subtilis	B-354	4.045	1	43.9	10	+
Saccharomyces cerevisiae	Y-567	12.1	2	38.3	109²	Yeast
Cryptococcus neoformans	Y-2534	18.9	2	48.3	60 ²	Yeast

Notes:

- ¹ Several strains within the standard were replaced with similar strains beginning from Lot ZRC190633. This update will not affect the species composition of the standard. Refer to Appendix B to check if your product is from an older lot, and find the correct reference database to use accordingly.
- ² 18S rRNA gene copy numbers in a haploid genome of the two strains of Saccharomyces cerevisiae and Cryptococcus neoformans were estimated based on read depth information from mapping shotgun sequencing data.

Note that yeast have copy numbers one order of magnitude higher than bacteria, although not relevant here it will be in the case of mixed populations between yeast and bacteria and amplicons for both 16S and 18S (or ITS).



Nextflow install and setup

The nextflow pipeline is available from github and should be cloned locally on the analysis server. After download of the github repo, the first nextflow run gets 3 Docker images and downloads the classification databases.

During the first run, the docker components (N=3) will be downloaded and installed in the default Docker cache on the server

- kpinpb/pb-16s-nf-tools
- kpinpb/pb-16s-nf-qiime
- kpinpb/pb-16s-vis

```
git clone git@github.com:PacificBiosciences/pb-16S-nf.git
cd pb-16S-nf

nextflow run main.nf \
    --download_db \
    -profile docker

# a new 'databases' folder is added to the nextflow folder

# GTDB_bac120_arc53_ssu_r207_fullTaxo.fa.gz

# GTDB_ssu_all_r207.qza

# GTDB_ssu_all_r207.taxonomy.qza

# RefSeq_16S_6-11-20_RDPv16_fullTaxo.fa.gz

# silva-138-99-seqs.qza

# silva-138-99-tax.qza

# silva_nr99_v138.1_wSpecies_train_set.fa.gz
```

Nextflow test

A built-in test data can be used to validate the install as follows

```
# Create test_sample.tsv for testing
echo -e "sample-id\tabsolute-filepath\ntest_data\t$(readlink -f test_data/test_1000_reads.fastq.gz)" > test_data/test_sample.tsv

nextflow run main.nf \
    --input test_data/test_sample.tsv \
    --metadata test_data/test_metadata.tsv \
    --outdir test_results \
    -profile docker
```

The run should take only few minutes and produce a folder with intermediate data and results as discussed later in this report.



Nextflow Zymo run

Two text files need to be prepared based on the available read sets; a sample manifest and a metadata file (as standard in QIIME2):

- manifest: a sample.tsv file that relates the sample names and the full path to each fastq file
- metadata: a tsv file that relates the same sample names to sample groups or conditions used in the wet-lab experiment (info provided by the customer)

The two files used in this test run are reproduced below

• sample.tsv

sample-id	absolute-file-path
4170_bc1005-bc1096	/data/analyses/Zymo-SequelIIe-Hifi/reads/4170_bc1005-bc1096.fastq.gz
4285_bc1022-bc1107	/data/analyses/Zymo-SequelIIe-Hifi/reads/4285_bc1022-bc1107.fastq.gz
4296_bc1022-bc1060	/data/analyses/Zymo-SequelIIe-Hifi/reads/4296_bc1022-bc1060.fastq.gz
4112_bc1008-bc1075	/data/analyses/Zymo-SequelIIe-Hifi/reads/4112_bc1008-bc1075.fastq.gz
4128_bc1005-bc1107	/data/analyses/Zymo-SequelIIe-Hifi/reads/4128_bc1005-bc1107.fastq.gz

metadata.tsv

sample_name	condition
4170_bc1005-bc1096	control
4285_bc1022-bc1107	control
$4296_bc1022-bc1060$	control
4112_bc1008-bc1075	control
$4128_bc1005-bc1107$	control

the condition column should be of type 'categorical' (not numeric!)

Note that in a real experiment, the conditions will describe more sample groups than just 'control'

The following code was run to start the analysis:

```
# the full list of full path fastq can be obtained with:
\# find fastq_folder -name "*.fastq.gz" -exec readlink -f {} \;
# use >= 32 cpu for good performance
cpu=32
infolder=<path-to-indata>
sample_file=${infoder}/sample.tsv
metadata_file=${infoder}/metadata.tsv
outfolder=<path-to-outdata>
nextflow run main.nf \
  --input ${sample_file} \
  --metadata ${metadata_file} \
  --outdir ${outfolder} \
  --dada2_cpu ${cpu} \
  --vsearch_cpu ${cpu} \
  --cutadapt_cpu ${cpu} \
  -profile docker
```

The nextflow pipeline produces live output and stores all log files for inspection as well as trouble-shooting



Zymo run results

After running the 5 Zymo samples, the standard output can be inspected and part of it shared with the customer The main output folder has the following standard structure:

cutadapt_summary
dada2
filtered_input_FASTQ
import_qiime
nb_tax
parameters.txt
results
summary_demux
trimmed_primers_FASTQ

The results folder contains symbolic links to all final key files and can be forwarded to the customer as-is

alpha-rarefaction-curves.gzv best_tax_merged_freq_tax.tsv best_taxonomy.tsv best_taxonomy_withDB.tsv best_tax.qza dada2_qc.tsv dada2_stats.qzv dada2_table.qzv feature-table-tax.biom feature-table-tax_vsearch.biom krona.gzv merged_freq_tax.qzv phylogeny_diversity rarefaction_depth_suggested.txt reads_QC samplefile.txt stats.tsv tax export taxonomy_barplot_nb.qzv ${\tt taxonomy_barplot_vsearch.qzv}$ taxonomy.vsearch.qza visualize_biom.html vsearch_merged_freq_tax.tsv

The results of this run are shared next to this report in the **Zymo-SequelIIe-Hifi_results_local** folder to allow more exploration of this typical data.

All files ending with .qzv are QIIME2 visualization files that can be fed to the online QIIME2-Viewer (https://view.qiime2.org/) to create and customize plots or tables.

Files with extension .qza are QIIME2 objects that can be reloaded in QIIME2 to proceed in the analysis while files with extension .tsv are data files that can be used for further analysis (eg. in R).

The main file present in the results folder is the RMarkdown converted document **visualize_biom.html** reporting all QC metrics and key findings through the user browser. Tables in that file are live and can be filtered.

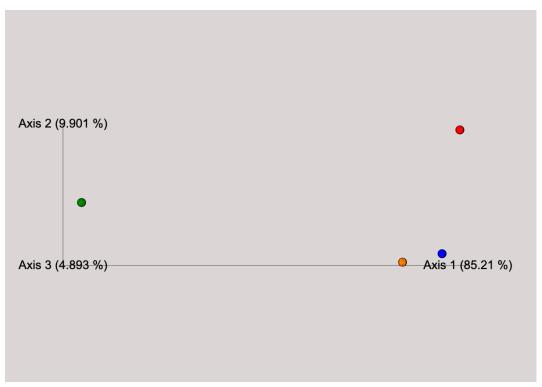


Examples of QIIME-View outputs

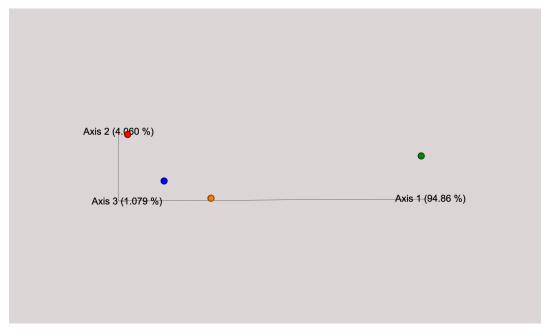
The .qzv files present in the results folder allow plotting using the **Qiime viewer site** 5

As illustration we show below two of the plots produced by the pipeline and showing multi dimentional principal component analyses results

• Bray Curtis



• Weighted unifrac



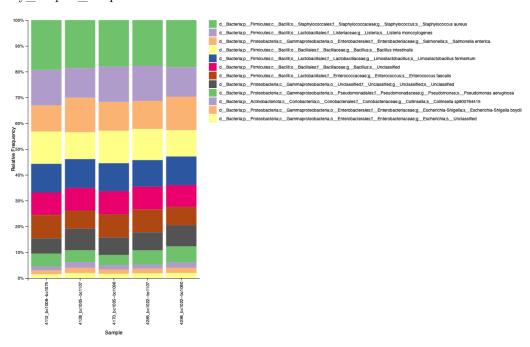
The viewer can also produce pictures for other results present in the results folder among which the two classification

 $^{^5 \}mathrm{https://view.qiime2.org/}$

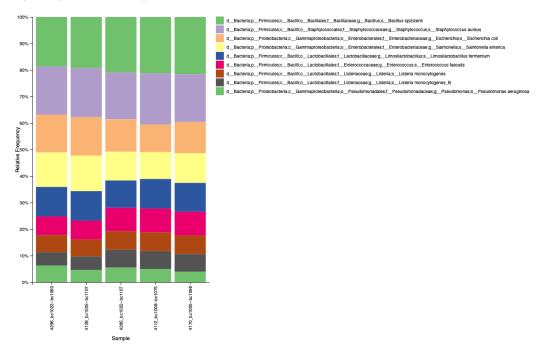


outcomes produced by the workflow for our 5 Zymo samples. The *Vsearch* classification is sometimes more accurate and is based only on Vsearch best hits while the *Naive Bayes Classifier* is based on multiple search results and may be more complete but may sometimes include absent species (see doc on github)).

• taxonomy_barplot_nb.qzv



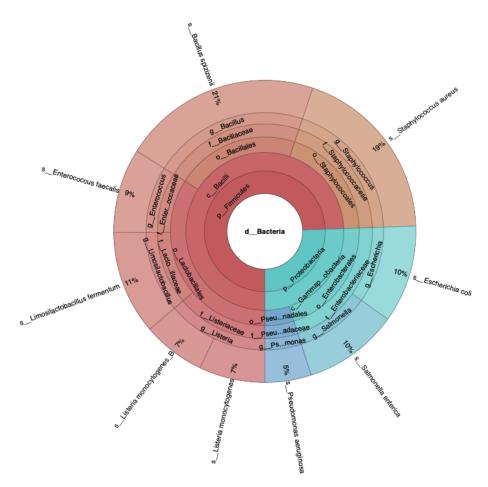
taxonomy_barplot_vsearch.qzv



When compared to the theoretical distribution shown in the first part of this report, experimental results are nicely concordant.

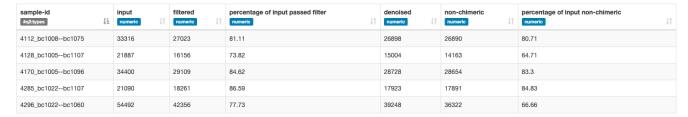
Additional plots can be produced like the Krona classification shown next (sample bc1008-bc1075)





Various tables (.qzv) can be converted to pretty tables in the viewer as well.

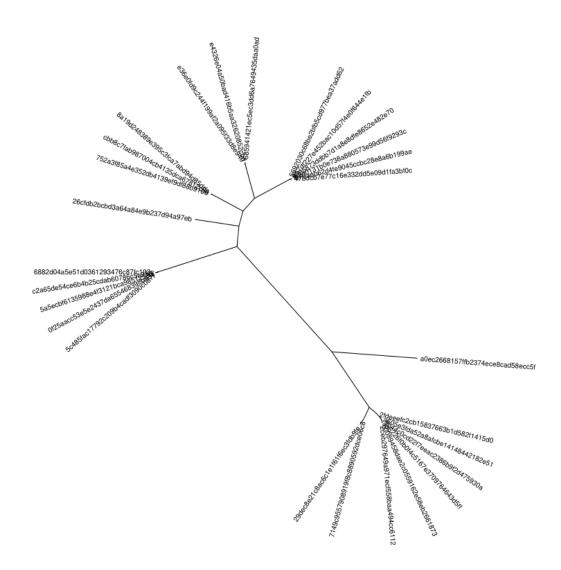
As example, the content of dada2_stats.qzv (this table is also present in the html report)



A phylogeny tree can be converted to a picture using a tree viewer 6 and the file $phylotree_mafft_rooted.nwk$ (note that, in the current version of the pipeline, the tree shows the ASV labels rather than the stain names)

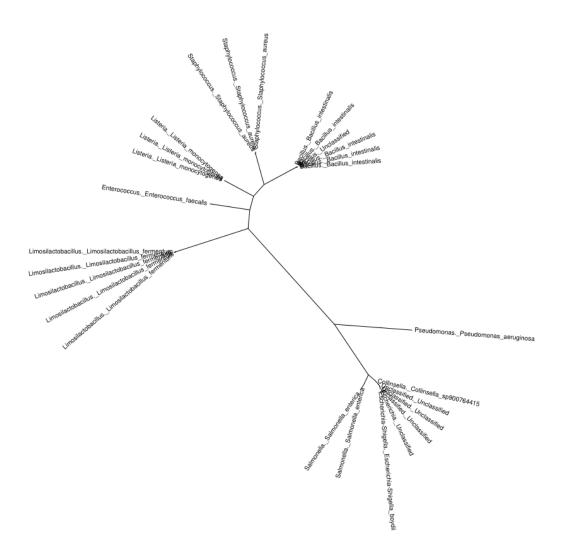
 $^{^6 \}rm https://github.com/arklumpus/TreeViewer$





A little R magic can replace the Feature.ID used for tip labels with genus and species extracted from the taxonomy results.





Note: In this particular case (synthetic community), multiple ASV of the same genus+species suggest different 16S copies with different sequences present in the same bacterium present in the mock community.

The height expected species are found back, although with slightly different names as in the Zymo doc. Some with a unique ASV (Pseudomonas.aeruginosa and Enterococcus.faecalis) while others are represented by multiple ASVs (eg. Escherichia Coli and Bacilus.subtilis both with 6 ASV detected).

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