Processing short nanopore reads with dada2

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Document utilises previously computed objects saved as rds files. Objects are reloaded from files. Original code can be executed by changing eval to TRUE or executing chunk inside Rstudio.

#### Preprocess Ion Torrent adapter reads

*Trim forward reads with Adapter A and trP1(rc) sequences*

cutadapt -g “CCATCTCATCCCTGCGTGTCTCCGACTCAG;o=30…ATCACCGACTGCCCATAGAGAGG;o=23” –trimmed-only -e 0.05 -o ev\_forward.fastq.gz ev\_reads\_hq.fastq.gz

*Trim reverse reads with trP1 and Adapter A(rc) sequences*

cutadapt -g ”CCTCTCTATGGGCAGTCGGTGAT;o=23…CTGAGTCGGAGACACGCAGGGATGAGATGG;o=30” –trimmed-only -e 0.05 -o ev\_reverse.fastq.gz ev\_reads\_hq.fastq.gz

*Reverse-complement reverse reads*

seqkit seq -rp -t DNA -o ev\_rcomp.fasta.gz ev\_reverce.fasta.gz

*Merge with forward reads*

cat ev\_forward.fasta.gz ev\_rcomp.fasta.gz > raw\_005.fasta.gz

*Import data to qiime*

qiime tools import –type MultiplexedSingleEndBarcodeInSequence –input-path raw\_005.fasta.gz –output-path raw\_005.qza

*Demultiplex*

qiime cutadapt demux-single –i-seqs raw\_005.qza –m-barcodes-file jt\_meta.tsv –m-barcodes-column Barcode\_seq –output-dir demuxed –p-error-rate 0 –p-anchor-barcode

*Trim pcr primers (519F and 926R)*

qiime cutadapt trim-single –i-demultiplexed-sequences per\_sample\_sequences.qza –p-overlap 15 –p-discard-untrimmed –p-front ACAGCMGCCGCGGTAATWC –o-trimmed-sequences trim1.qza

qiime cutadapt trim-single –i-demultiplexed-sequences trim1.qza –p-adapter AAACTCAAAKGAATTGACGG –o-trimmed-sequences trimmed-sequences.qza

*Decompress read files*

unzip trimmed-sequences.qza

**Note.** Parameters allow one error in sequencing adapters, no errors in barcode sequence and 1 and 2 errors in pcr primers, respectively.

**Note.** Some options in commands require double dash and are not displayed correctly in rendered document.

#### Load libraries

library(dada2);packageVersion("dada2")

[1] '1.32.0'

library(knitr);packageVersion("knitr")

[1] '1.48'

library(Biostrings);packageVersion("Biostrings")

[1] '2.72.1'

library(DECIPHER);packageVersion("DECIPHER")

[1] '3.0.0'

library(phyloseq);packageVersion("phyloseq")

[1] '1.48.0'

library(tidyverse);packageVersion("tidyverse")

[1] '2.0.0'

library(kableExtra);packageVersion("kableExtra")

[1] '1.4.0'

library(mia);packageVersion("mia")

[1] '1.12.0'

library(tidyr);packageVersion("tidyr")

[1] '1.3.1'

library(ggthemes);packageVersion("ggthemes")

[1] '5.1.0'

library(vegan);packageVersion("vegan")

[1] '2.6.6.1'

library(scater);packageVersion("scater")

[1] '1.32.1'

library(patchwork);packageVersion("patchwork")

[1] '1.2.0'

library(ggsci);packageVersion("ggsci")

[1] '3.2.0'

#### Set parameters

# Path variables  
path <- "data/reads/"  
training <- "~/feature\_classifiers/SILVA\_SSU\_r138\_2019.RData"  
meta\_file <- "data/jt\_meta.tsv"  
exportloc <- "results/"  
# Variables: truncation length, phix (Illumina)  
truncation <- 350  
#Creates results directory  
dir.create(exportloc)  
#metadata  
metadata <- data.frame(read\_tsv(meta\_file, show\_col\_types = FALSE))  
#set knitr cache path  
knitr::opts\_chunk$set(cache.path = "cache/")

#### Import reads

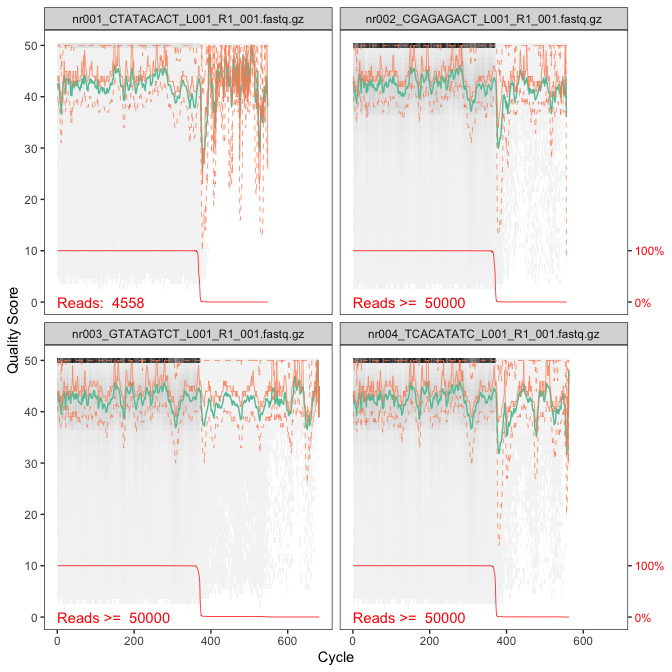
nr072 was removed from dataset (0 reads caused error in denoising).

# Forward fastq filenames have format: SAMPLENAME\_R1\_001.fastq  
fnFs <- sort(list.files(path, pattern="L001\_R1\_001.fastq.gz", full.names = TRUE))  
# Extract sample names, assuming filenames have format: SAMPLENAME\_XXX.fastq.gz  
sample.names <- sapply(strsplit(basename(fnFs), "\_"), `[`, 1)

Checking read quality of first samples

# Base quality plot  
p <- plotQualityProfile(fnFs[1:4], n = 50000)  
p

# Load base quality plot from saved object  
p <- readRDS("rds/qplot.rds")  
p



#### Filter and trim reads

# Filtered files are placed in filtered subdirectory  
filtFs <- file.path(path, "filtered", paste0(sample.names,  
 "\_F\_filt.fastq.gz"))  
# For single end data sets, maxEE default values  
names(filtFs) <- sample.names  
out <- filterAndTrim(fnFs, filtFs, truncLen=truncation,  
 maxN = 0, maxEE = 2 , truncQ = 2,  
 compress = TRUE, multithread = FALSE,  
 rm.phix = FALSE)  
saveRDS(out, file = "rds/out.rds")

# Filtered files are placed in filtered subdirectory  
filtFs <- file.path(path, "filtered", paste0(sample.names,  
 "\_F\_filt.fastq.gz"))  
# For single end data sets without phix control  
names(filtFs) <- sample.names  
# Load previously saved object  
out <-readRDS("rds/out.rds")

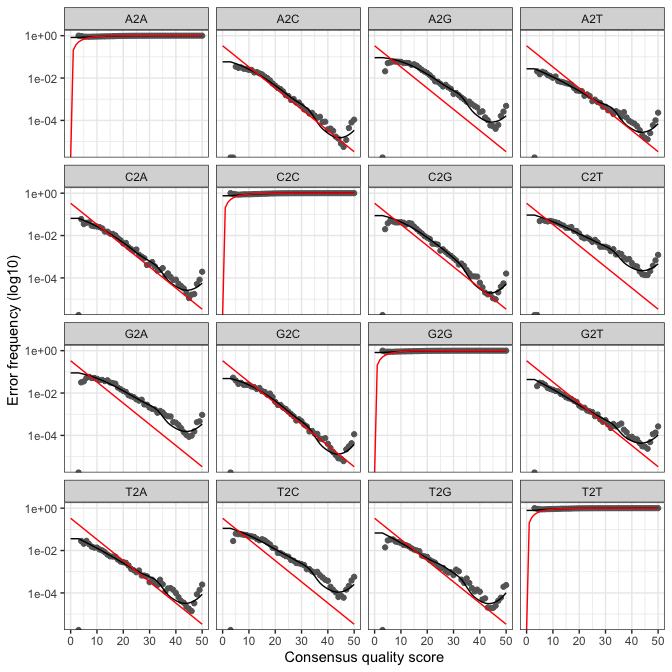
#### Learn and plot error profile

# Forward read error rate  
errF <- learnErrors(filtFs, multithread = TRUE)  
saveRDS(errF, file = "rds/errF.rds")

# Load previously saved object  
errF <- readRDS("rds/errF.rds")

Plot error profile

# Plotting error rate profile for forward reads  
plotErrors(errF, nominalQ = TRUE)



#### Denoise sequences

dadaFs <- dada(filtFs, err = errF, multithread = TRUE, verbose = FALSE)  
saveRDS(dadaFs, file = "rds/dadaFs.rds")

# Load previously saved object  
dadaFs <- readRDS("rds/dadaFs.rds")

#### Build ASV table

seqtab <- makeSequenceTable(dadaFs)  
# Dimensions of ASV table  
dim(seqtab)

[1] 131 8404

Remove chimeric variants

seqtab.nochim <- removeBimeraDenovo(seqtab, method = "consensus",  
 multithread = TRUE, verbose = TRUE)  
dim(seqtab.nochim)

[1] 131 7634

Amount of data remaining after chimera removal

sum(seqtab.nochim)/sum(seqtab)

[1] 0.9959133

#### Summary table

#If processing a single sample, remove the sapply calls  
getN <- function(x) sum(getUniques(x))  
track <- cbind(out, sapply(dadaFs, getN), rowSums(seqtab.nochim),  
 rowSums(seqtab.nochim != 0))  
colnames(track) <- c("Input", "Filtered", "DenoisedF", "Nonchimeric",  
 "N:o of variants")  
rownames(track) <- sample.names  
#table  
kable(track, caption="Summary table", booktabs = TRUE, longtable = TRUE) %>%  
 kable\_styling(latex\_options=c("striped", "HOLD\_position", "repeat\_header")) %>%  
 row\_spec(0,background = "teal", color = "ivory")

Summary table

|  | Input | Filtered | DenoisedF | Nonchimeric | N:o of variants |
| --- | --- | --- | --- | --- | --- |
| nr001 | 4558 | 3737 | 3589 | 3589 | 96 |
| nr002 | 101469 | 82272 | 82017 | 82017 | 449 |
| nr003 | 139054 | 113872 | 113654 | 113612 | 359 |
| nr004 | 256204 | 207763 | 207374 | 207160 | 361 |
| nr005 | 104757 | 85575 | 85340 | 85153 | 413 |
| nr006 | 125276 | 101867 | 101438 | 101135 | 407 |
| nr007 | 16776 | 13466 | 13352 | 13352 | 179 |
| nr008 | 60779 | 50372 | 50261 | 50261 | 259 |
| nr009 | 57496 | 46090 | 45844 | 45777 | 250 |
| nr010 | 58884 | 48417 | 48302 | 48287 | 239 |
| nr011 | 57804 | 46828 | 46648 | 46648 | 315 |
| nr012 | 63184 | 51624 | 51355 | 51279 | 291 |
| nr013 | 184184 | 151003 | 150506 | 150337 | 305 |
| nr014 | 37034 | 30255 | 30059 | 30059 | 167 |
| nr015 | 52921 | 42498 | 42295 | 42214 | 256 |
| nr016 | 57780 | 47377 | 47221 | 47198 | 184 |
| nr017 | 32249 | 27015 | 26761 | 26761 | 223 |
| nr018 | 32343 | 26457 | 26235 | 26204 | 238 |
| nr019 | 211 | 182 | 152 | 152 | 8 |
| nr020 | 17466 | 14548 | 14312 | 14309 | 271 |
| nr021 | 13454 | 10843 | 10407 | 10368 | 298 |
| nr022 | 36958 | 30203 | 29950 | 29916 | 330 |
| nr023 | 5863 | 4939 | 4760 | 4760 | 169 |
| nr024 | 39062 | 32944 | 32682 | 32675 | 287 |
| nr025 | 18358 | 15039 | 14839 | 14839 | 234 |
| nr026 | 54181 | 45020 | 44705 | 44689 | 461 |
| nr027 | 57462 | 46962 | 46734 | 46734 | 281 |
| nr028 | 6385 | 5190 | 4976 | 4976 | 135 |
| nr029 | 1149 | 915 | 809 | 809 | 57 |
| nr030 | 2978 | 2463 | 2280 | 2280 | 72 |
| nr031 | 55110 | 44610 | 44399 | 44191 | 303 |
| nr032 | 53814 | 42979 | 42625 | 42284 | 219 |
| nr033 | 41001 | 32923 | 32685 | 32664 | 287 |
| nr034 | 25124 | 20580 | 20298 | 20296 | 288 |
| nr035 | 1028 | 835 | 761 | 761 | 34 |
| nr036 | 55738 | 45478 | 45334 | 45302 | 332 |
| nr037 | 52126 | 42022 | 41630 | 41550 | 325 |
| nr038 | 63365 | 52054 | 51762 | 51391 | 404 |
| nr039 | 62495 | 50482 | 50215 | 50090 | 428 |
| nr040 | 66486 | 54623 | 54394 | 54325 | 364 |
| nr041 | 50049 | 41199 | 41048 | 41028 | 288 |
| nr042 | 56406 | 47601 | 47412 | 47302 | 201 |
| nr043 | 62973 | 52026 | 51849 | 51794 | 248 |
| nr044 | 51860 | 42460 | 42297 | 42212 | 259 |
| nr045 | 62931 | 51359 | 50996 | 50953 | 410 |
| nr046 | 57953 | 47598 | 47331 | 46941 | 259 |
| nr047 | 59138 | 49654 | 49488 | 49442 | 260 |
| nr048 | 55998 | 46407 | 46245 | 46231 | 284 |
| nr049 | 59015 | 49029 | 48942 | 48940 | 252 |
| nr050 | 58770 | 48128 | 47929 | 47881 | 367 |
| nr051 | 58199 | 48129 | 48011 | 47938 | 241 |
| nr052 | 59722 | 48648 | 48421 | 48387 | 456 |
| nr053 | 52295 | 42909 | 42812 | 42810 | 206 |
| nr054 | 54222 | 44000 | 43814 | 43761 | 255 |
| nr055 | 44896 | 36482 | 36250 | 36211 | 327 |
| nr056 | 56564 | 46213 | 46048 | 46037 | 340 |
| nr057 | 57220 | 44611 | 44384 | 44384 | 340 |
| nr058 | 29071 | 21680 | 21628 | 21628 | 101 |
| nr059 | 49673 | 37678 | 37585 | 37585 | 184 |
| nr060 | 56441 | 44838 | 44765 | 44720 | 204 |
| nr061 | 39464 | 31746 | 31592 | 31558 | 203 |
| nr062 | 57638 | 47417 | 47311 | 47311 | 121 |
| nr063 | 41622 | 34042 | 33991 | 33847 | 116 |
| nr064 | 57665 | 45751 | 45672 | 45672 | 151 |
| nr065 | 31057 | 24009 | 23923 | 23923 | 115 |
| nr066 | 46974 | 36323 | 36192 | 36192 | 160 |
| nr067 | 28068 | 23098 | 22969 | 22967 | 116 |
| nr068 | 29823 | 24238 | 24203 | 24203 | 86 |
| nr069 | 30168 | 24183 | 24129 | 24127 | 128 |
| nr070 | 60493 | 50010 | 49839 | 49598 | 200 |
| nr071 | 53242 | 41175 | 41058 | 40969 | 179 |
| nr073 | 49981 | 40478 | 40360 | 40357 | 179 |
| nr074 | 57395 | 46109 | 46003 | 46003 | 153 |
| nr075 | 56378 | 46525 | 46362 | 46362 | 181 |
| nr076 | 50105 | 39042 | 38965 | 38965 | 198 |
| nr077 | 53839 | 45055 | 44930 | 44903 | 272 |
| nr078 | 49264 | 40522 | 40432 | 40428 | 324 |
| nr079 | 1023 | 865 | 775 | 775 | 25 |
| nr080 | 76644 | 64132 | 63928 | 63928 | 348 |
| nr081 | 77967 | 62167 | 61521 | 61008 | 1391 |
| nr082 | 39055 | 32857 | 32731 | 32731 | 161 |
| nr083 | 33607 | 27830 | 27742 | 27719 | 227 |
| nr084 | 6891 | 5655 | 5515 | 5515 | 88 |
| nr085 | 63220 | 52028 | 51855 | 51853 | 218 |
| nr086 | 28118 | 21641 | 21539 | 21515 | 165 |
| nr087 | 1547 | 1240 | 1214 | 1214 | 44 |
| nr088 | 355 | 291 | 248 | 248 | 19 |
| nr089 | 94933 | 75244 | 74715 | 73698 | 62 |
| nr090 | 49193 | 39043 | 38697 | 38660 | 56 |
| nr091 | 43492 | 34421 | 34160 | 34137 | 35 |
| nr092 | 87487 | 69403 | 69153 | 69153 | 64 |
| nr093 | 170090 | 138581 | 137950 | 137467 | 90 |
| nr094 | 183908 | 136075 | 135427 | 135250 | 92 |
| nr095 | 187778 | 149971 | 149441 | 149364 | 86 |
| nr096 | 152388 | 120640 | 120247 | 120041 | 86 |
| nr097 | 151106 | 122605 | 122173 | 122018 | 69 |
| nr098 | 61905 | 49312 | 49062 | 49001 | 50 |
| nr099 | 6228 | 5126 | 5003 | 5003 | 32 |
| nr100 | 181 | 137 | 91 | 91 | 11 |
| nr101 | 118 | 86 | 47 | 47 | 7 |
| nr102 | 65 | 48 | 23 | 23 | 4 |
| nr103 | 196 | 145 | 78 | 78 | 9 |
| nr104 | 198 | 153 | 100 | 100 | 13 |
| nr105 | 179 | 145 | 82 | 82 | 11 |
| nr106 | 107 | 83 | 42 | 42 | 5 |
| nr107 | 176 | 144 | 68 | 68 | 7 |
| nr108 | 92 | 71 | 29 | 29 | 5 |
| nr109 | 139 | 113 | 67 | 67 | 9 |
| nr110 | 447 | 360 | 269 | 269 | 32 |
| nr111 | 122 | 96 | 52 | 52 | 4 |
| nr112 | 212 | 174 | 114 | 114 | 11 |
| nr113 | 144 | 116 | 74 | 74 | 5 |
| nr114 | 127 | 103 | 57 | 57 | 7 |
| nr115 | 502 | 393 | 378 | 378 | 11 |
| nr116 | 73 | 52 | 16 | 16 | 2 |
| nr117 | 127 | 103 | 46 | 46 | 6 |
| nr118 | 120 | 94 | 47 | 47 | 5 |
| nr119 | 170311 | 133480 | 132659 | 129636 | 87 |
| nr120 | 157 | 126 | 64 | 64 | 8 |
| nr121 | 161 | 131 | 72 | 72 | 11 |
| nr122 | 199 | 161 | 117 | 117 | 14 |
| nr123 | 60 | 48 | 14 | 14 | 2 |
| nr124 | 100 | 83 | 47 | 47 | 10 |
| nr125 | 84 | 66 | 35 | 35 | 6 |
| nr126 | 81 | 63 | 25 | 25 | 2 |
| nr127 | 208327 | 163285 | 162177 | 159439 | 112 |
| nr128 | 72407 | 59091 | 58754 | 58652 | 56 |
| nr129 | 45752 | 37244 | 37008 | 36922 | 44 |
| nr130 | 192968 | 154678 | 153444 | 145777 | 122 |
| nr131 | 19 | 17 | 13 | 13 | 2 |
| nr132 | 14 | 11 | 1 | 1 | 1 |

#### Assign taxonomy

#Create a DNAStringSet from the ASV sequences  
repseq <- DNAStringSet(getSequences(seqtab.nochim))  
# CHANGE TO THE PATH OF YOUR TRAINING SET  
load(training)  
ids <- IdTaxa(repseq, trainingSet, strand = "top",  
 processors = 3, verbose = FALSE,  
 threshold = 50)  
ranks <- c("domain", "phylum", "class", "order", "family",  
 "genus", "species")   
# Convert the output to a matrix analogous to the output from assignTaxonomy  
taxid <-t(sapply(ids, function(x) {  
 m <- match(ranks, x$rank)  
 taxa <- x$taxon[m]  
 taxa[startsWith(taxa, "unclassified\_")] <- NA  
 taxa  
}))  
colnames(taxid) <- ranks; rownames(taxid) <- getSequences(seqtab.nochim)  
saveRDS(taxid, file = "rds/taxid.rds")

# Load previously saved object  
taxid <- readRDS("rds/taxid.rds")

#### Build phyloseq object

pseq <- phyloseq(otu\_table(seqtab.nochim, taxa\_are\_rows = FALSE),  
 tax\_table(taxid))  
row.names(metadata) <- sample\_names(pseq)  
sample\_data(pseq) <- metadata  
pseq

phyloseq-class experiment-level object  
otu\_table() OTU Table: [ 7634 taxa and 131 samples ]  
sample\_data() Sample Data: [ 131 samples by 5 sample variables ]  
tax\_table() Taxonomy Table: [ 7634 taxa by 7 taxonomic ranks ]

Save repseq’s as refseq and rename variants

#create sequences from rownames to refseq   
seqs <- DNAStringSet(taxa\_names(pseq))  
names(seqs) <- taxa\_names(pseq)  
pseq <- merge\_phyloseq(pseq, seqs)  
#new variant names  
taxa\_names(pseq) <- paste0("ASV", seq(ntaxa(pseq)))  
#capitalise taxonomic ranks  
colnames(tax\_table(pseq)) <- c("Kingdom", "Phylum", "Class",   
 "Order", "Family", "Genus", "Species")

Remove non-bacterial variants

#remove non-bacterial taxa  
pseq <- subset\_taxa(pseq, Kingdom != is.na(Kingdom))  
pseq <- subset\_taxa(pseq, Kingdom != "Eukaryota")  
pseq

phyloseq-class experiment-level object  
otu\_table() OTU Table: [ 7280 taxa and 131 samples ]  
sample\_data() Sample Data: [ 131 samples by 5 sample variables ]  
tax\_table() Taxonomy Table: [ 7280 taxa by 7 taxonomic ranks ]  
refseq() DNAStringSet: [ 7280 reference sequences ]

#### Write results to files

Abundance table is transponed and written as tsv file

#variant names in rows  
ASV\_names <- taxa\_names(pseq)  
#sample names will be columns  
ASV\_counts <- t(otu\_table(pseq))  
ASVdf <- (data.frame(ASV\_names,ASV\_counts))  
#write  
write\_tsv(ASVdf, paste0(exportloc,"asvs.tsv"))

Likewise taxonomy table is saved as tsv

#variant names in rows  
ASV\_names <- taxa\_names(pseq)  
#taxonomy ranks in columns  
taxonomy <- (data.frame(ASV\_names, tax\_table(pseq)))  
#write  
write\_tsv(taxonomy,paste0(exportloc,"taxonomy.tsv"))

Variant sequences are saved into fasta file

pseq %>% refseq() %>% writeXStringSet(paste0(exportloc,"repseq.fasta"),  
 append = FALSE, compress = FALSE,  
 format = "fasta")

Compatible metadata file as tsv

sampleid <- sample\_names(pseq)  
metafile <- sample\_data(pseq)  
metadf <- data.frame(sampleid,metafile)  
write\_tsv(metadf, paste0(exportloc,"metadata.tsv"))

#### Vsearch denovo clustering (99%) data from Qiime

Import data from qiime qza files

#read otu table and meta file  
imported <- loadFromQIIME2(featureTableFile = "data/otu-table.qza",  
 sampleMetaFile = "data/jt\_meta2.tsv")  
#read decipher taxonomy  
taxid\_vs <- readRDS("data/taxid.rds")  
#convert tse object to phyloseq  
vsearch\_phylo <- makePhyloseqFromTreeSummarizedExperiment(imported)  
#add tax table  
tax\_table(vsearch\_phylo) <-taxid\_vs  
#change rank names  
colnames(tax\_table(vsearch\_phylo)) <- c("Kingdom", "Phylum", "Class",   
 "Order", "Family", "Genus", "Species")

Prune taxonomic data and convert back to TSE object

#remove non-bacterial taxa  
vsearch\_phylo <- subset\_taxa(vsearch\_phylo, Kingdom != is.na(Kingdom))  
vsearch\_phylo <- subset\_taxa(vsearch\_phylo, Kingdom != "Eukaryota")  
#vs\_phylo  
vsearch <- makeTreeSummarizedExperimentFromPhyloseq(vsearch\_phylo)  
#remove nr072 sample missing from dada dataset  
vsearch <- subsetSamples(vsearch, colData(vsearch)$SampleID != "nr072")  
#view object  
vsearch

class: TreeSummarizedExperiment   
dim: 124770 131   
metadata(0):  
assays(1): counts  
rownames(124770): efff17dee73a28bd87be7b58eb012da9f68fd1ff  
 2a7d89c5f313d7488ceb4d8d99f88de12f28221a ...  
 3b840c7f81f6ce484389ecd898e7eb4fb0f50027  
 3f53a184f4dccc6affaadf9cf2946a88041f60de  
rowData names(7): Kingdom Phylum ... Genus Species  
colnames(131): nr001 nr002 ... nr131 nr132  
colData names(5): SampleID Name Source Type Barcode\_seq  
reducedDimNames(0):  
mainExpName: NULL  
altExpNames(0):  
rowLinks: NULL  
rowTree: NULL  
colLinks: NULL  
colTree: NULL

Convert dada object to tse

dada <- makeTreeSummarizedExperimentFromPhyloseq(pseq)  
#create column for total counts  
colData(dada)$Sum <- colSums(assays(dada)$counts)  
colData(vsearch)$Sum <- colSums(assays(vsearch)$counts)  
  
#filter samples that have over 10 000 counts  
dada <- subsetSamples(dada, colData(dada)$Sum >= 10000)  
vsearch <- subsetSamples(vsearch, colData(vsearch)$Sum >= 10000)  
dim(dada)

[1] 7280 91

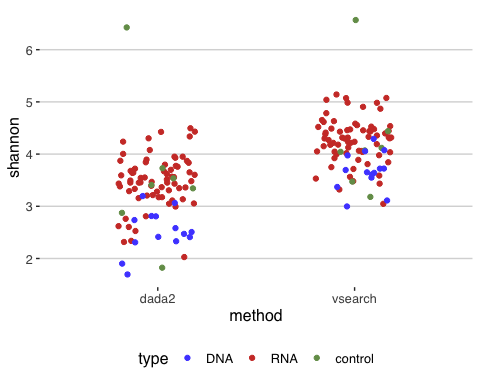
dim(vsearch)

[1] 124770 91

#create second copies  
dada\_2 <- dada  
vsearch\_2 <- vsearch

Calculate shannon index and compare denoising ja clustering results

#calculate shannon indexes  
dada <- estimateDiversity(dada, index = "shannon")  
vsearch <- estimateDiversity(vsearch, index = "shannon")  
#create df with values and sample type  
shannon <- data.frame(dada2 = colData(dada)$shannon, vsearch = colData(vsearch)$shannon, type = colData(dada)$Type)  
#pivot table  
long\_diversity <- shannon %>% pivot\_longer(col = c(dada2,vsearch), names\_to = "method", values\_to = "shannon")  
#plot data  
plot\_shannon <- ggplot(long\_diversity, aes(method,shannon, color=type)) + geom\_point(position = position\_jitter(width=0.2)) + theme\_hc() + scale\_color\_igv()  
plot\_shannon



Calculate bray-curtis and compare denoising ja clustering results

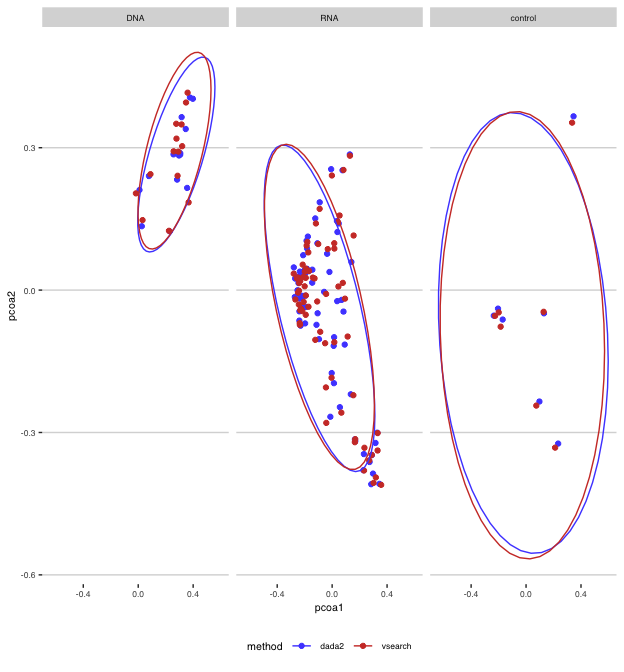
#transform dada counts to relabundance  
dada <- transformAssay(dada, assay.type = "counts", method = "relabundance",  
 name = "relabundance")  
#create bray-curtis distance matrix  
dada <- runMDS(dada, FUN = vegan::vegdist, method = "bray",  
 name="PCoA\_BC", exprs\_values = "relabundance")  
dada\_bray <- plotReducedDim(dada, "PCoA\_BC")  
#create df for plot  
bray\_dada\_df <- data.frame(pcoa1 = dada\_bray$data[,1],  
 pcoa2 = dada\_bray$data[,2],  
 type = colData(dada)$Type)  
#transform vsearch counts to relabundance  
vsearch <- transformAssay(vsearch, assay.type = "counts", method = "relabundance",  
 name = "relabundance")  
#create bray-curtis distance matrix  
vsearch <- runMDS(vsearch, FUN = vegan::vegdist, method = "bray",  
 name="PCoA\_BC", exprs\_values = "relabundance")  
vsearch\_bray <- plotReducedDim(vsearch, "PCoA\_BC")  
#create df for plot  
bray\_vsearch\_df <- data.frame(pcoa1 = vsearch\_bray$data[,1],  
 pcoa2 = vsearch\_bray$data[,2],  
 type = colData(vsearch)$Type)

Combine dataframes

#combined df  
bray\_both <- data.frame(dada1 = bray\_dada\_df$pcoa1,  
 dada2 = bray\_dada\_df$pcoa2,  
 vsearch1 = bray\_vsearch\_df$pcoa1,  
 vsearch2 = bray\_vsearch\_df$pcoa2,  
 type = bray\_dada\_df$type)  
bray\_both = data.frame(pcoa1=c(bray\_both$dada1, bray\_both$vsearch1),  
 pcoa2=c(bray\_both$dada2, bray\_both$vsearch2),  
 type=c(bray\_dada\_df$type,bray\_dada\_df$type))  
bray\_both$method <- "dada2"   
bray\_both$method[92:182] <- "vsearch"  
plot\_both <- ggplot(bray\_both, aes(pcoa1,pcoa2, color=method)) +  
 geom\_point() + facet\_wrap(~ type) + theme\_hc(base\_size=8) + scale\_color\_igv() + stat\_ellipse()

Plot

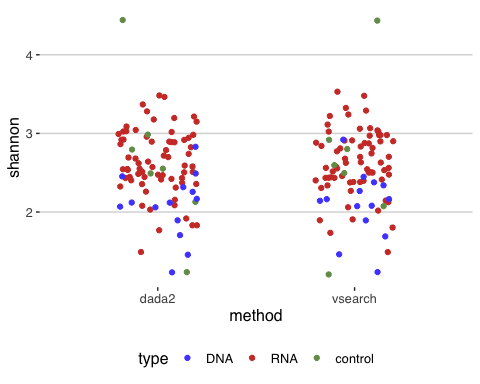
(plot\_both)



Differences are small

Recalculate indexes after taxonomy have been agglomerated to genus level

#calculate shannon indexes  
dada\_2 <- agglomerateByRank(dada\_2, rank = "Genus")  
dada\_2 <- estimateDiversity(dada\_2, index = "shannon")  
vsearch\_2 <- agglomerateByRank(vsearch\_2, rank = "Genus")  
vsearch\_2 <- estimateDiversity(vsearch\_2, index = "shannon")  
#create df with values and sample type  
shannon2 <- data.frame(dada2 = colData(dada\_2)$shannon,  
 vsearch = colData(vsearch\_2)$shannon,  
 type = colData(dada\_2)$Type)  
#pivot table  
long\_diversity2 <- shannon2 %>% pivot\_longer(col = c(dada2,vsearch),  
 names\_to = "method",  
 values\_to = "shannon")  
#plot data  
plot\_shannon2 <- ggplot(long\_diversity2, aes(method, shannon, color = type)) +  
 geom\_point(position = position\_jitter(width = 0.2)) + theme\_hc() +  
 scale\_color\_igv()  
plot\_shannon2

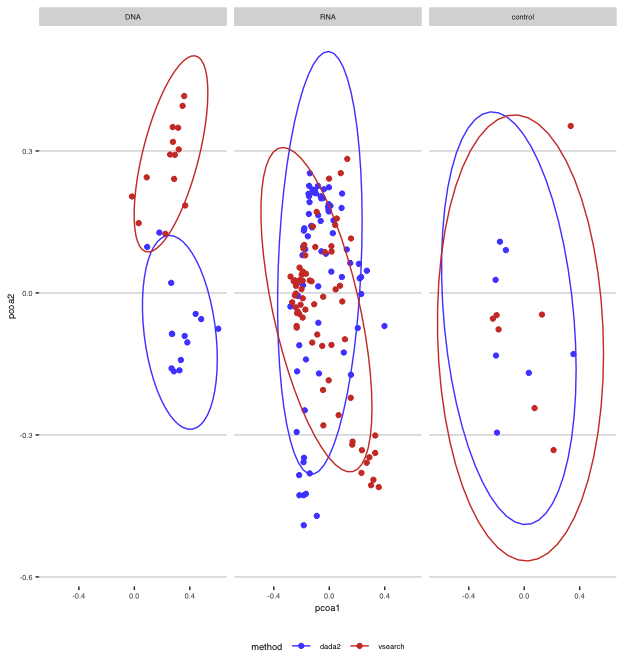


#transform dada counts to relabundance  
dada\_2 <- transformAssay(dada\_2, assay.type = "counts",  
 method = "relabundance", name = "relabundance")  
#create bray-curtis distance matrix  
dada\_2 <- runMDS(dada\_2, FUN = vegan::vegdist, method = "bray",  
 name="PCoA\_BC", exprs\_values = "relabundance")  
dada\_bray2 <- plotReducedDim(dada\_2, "PCoA\_BC")  
#create df for plot  
bray\_dada\_df <- data.frame(pcoa1 = dada\_bray2$data[,1],  
 pcoa2 = dada\_bray2$data[,2],  
 type = colData(dada\_2)$Type)  
#transform vsearch counts to relabundance  
vsearch\_2 <- transformAssay(vsearch\_2, assay.type = "counts", method = "relabundance",  
 name = "relabundance")  
#create bray-curtis distance matrix  
vsearch\_2 <- runMDS(vsearch, FUN = vegan::vegdist, method = "bray",  
 name="PCoA\_BC", exprs\_values = "relabundance")  
vsearch\_bray2 <- plotReducedDim(vsearch\_2, "PCoA\_BC")  
#create df for plot  
bray\_vsearch\_df <- data.frame(pcoa1 = vsearch\_bray2$data[,1],  
 pcoa2 = vsearch\_bray2$data[,2],  
 type = colData(vsearch\_2)$Type)

#combined df  
bray\_both <- data.frame(dada1 = bray\_dada\_df$pcoa1,  
 dada2 = bray\_dada\_df$pcoa2,  
 vsearch1 = bray\_vsearch\_df$pcoa1,  
 vsearch2 = bray\_vsearch\_df$pcoa2,  
 type = bray\_dada\_df$type)  
bray\_both = data.frame(pcoa1=c(bray\_both$dada1, bray\_both$vsearch1),  
 pcoa2=c(bray\_both$dada2, bray\_both$vsearch2),  
 type=c(bray\_dada\_df$type,bray\_dada\_df$type))  
bray\_both$method <- "dada2"   
bray\_both$method[92:182] <- "vsearch"  
plot\_both <- ggplot(bray\_both, aes(pcoa1,pcoa2, color=method)) + facet\_wrap(~ type) +  
 geom\_point() + theme\_hc(base\_size=7) + scale\_color\_igv() + stat\_ellipse()

Agglomerated pcoa plot

plot\_both



#### Observations

Customised sup basecalling produce high quality reads

Error profiles of short amplicon (truncated to 350 bp) follow expected frequency

Proportion of unique reads is much smaller when compared to long amplicons.

Thus, denoising seem to work normally on shorter read lengths

Shannon diversity results are higher when using vsearch. Clustering at 99 % identity level produces very high number of variants, so that might cause overestimation. Shannon values are brought much closer to each other, when data is agglomerated.

Bray-curtis plots are very similar in samples with more than 10 000 counts.

#### Advantages of nanopore

Read length is not limiting factor while designing amplicon targets

Base quality doesn’t decrease as a function of read length

Low diversity libraries are not problem for sequencing chemistry

Libraries prepared for other platforms can be conveniently converted to nanopore

Live basecalling allows controlling sequencing throughput and in some cases flow-cell can be reused

Cost per bp in amplicon sequencing is good compared to MiSeq

#### Disadvantages of nanopore

Homopolymer region accuracy is not quite as good as in Illumina

High accuracy basecalling is computationally quite intensive

Software tools are not at the same level as in other platforms and require more knowledge

Consistency of flow-cells (number of functional pores) and repeatibility of sequencing is so far unclear. As is shelf-life of flow-cells. Oxford Nanopore guarantee is 3 months from delivery date.

Pores might die if library preparation contains contaminants originating from sample (concerns mainly genomic or transcriptomic sequencing)