

DNA metabarcoding diet analysis in reindeer is quantitative and integrates feeding over several weeks

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Abstract

Filtering of the SPER01 DNA metabarcoding raw data.

1. Setting up the R environment

1.1. Install missing packages

```
packages <- c("igraph", "tidyverse", "devtools",  
             "R.utils", "vegan", "vctrs", "magrittr")  
install.packages(setdiff(packages,  
                        rownames(installed.packages())),  
                dependencies = TRUE  
                )
```

1.2. Loading of the R libraries

- ROBITools package is used to read result files produced by OBITools.
- ROBITaxonomy package provides function allowing to query OBITools formatted taxonomy.

```
if (!"ROBITools" %in% rownames(installed.packages())) {  
  # ROBITools are not available on CRAN and have to be installed  
  # from http://git.metabarcoding.org using devtools  
  
  metabarcoding_git <- "https://git.metabarcoding.org/obitools"
```

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```

devtools::install_git(paste(metabarcoding_git,
                             "ROBIUtils.git",
                             sep="/"))

devtools::install_git(paste(metabarcoding_git,
                             "ROBITaxonomy.git",
                             sep="/"))

devtools::install_git(paste(metabarcoding_git,
                             "ROBITools.git",
                             sep="/"))
}

library(ROBITools)
library(ROBITaxonomy)

```

- [tidyverse](#)¹ provides various method for efficient data manipulation and plotting via [ggplot2](#)²

```
library(tidyverse)
```

```
library(R.utils)
```

```
library(vegan)
```

```
library(magrittr)
```

```
source("methods.R")
```

Attaching package: 'matrixStats'

The following object is masked from 'package:dplyr':

count

Attaching package: 'vctrs'

The following object is masked from 'package:dplyr':

data_frame

The following object is masked from 'package:tibble':

data_frame

2. Loading the data

2.1. Load the NCBI taxonomy

```
if (! file.exists("Data/ncbi20210212.adx")) {  
  gunzip("Data/ncbi20210212.adx.gz",remove=FALSE)  
  gunzip("Data/ncbi20210212.ndx.gz",remove=FALSE)  
  gunzip("Data/ncbi20210212.rdx.gz",remove=FALSE)  
  gunzip("Data/ncbi20210212.tdx.gz",remove=FALSE)  
}
```

```
taxo <- read.taxonomy("Data/ncbi20210212")
```

2.2. Loading the metabarcoding data

```
if (! file.exists("Data/Rawdata/SPER01_all_paired.ali.assigned.ann.diag.uniq.ann.c1.l10.clean.EMBL.tag.ann"))  
  gunzip("Data/Rawdata/SPER01_all_paired.ali.assigned.ann.diag.uniq.ann.c1.l10.clean.EMBL.tag.ann.gz")
```

```
SPER01.raw = import.metabarcoding.data("Data/Rawdata/SPER01_all_paired.ali.assigned.ann.diag.uniq.ann.c1.l10.clean.EMBL.tag.ann.gz")
```

2.3. Loading the metadata

```
samples.metadata = read_csv("Data/Faeces/sampling_dates.csv",  
                             show_col_types = FALSE)
```

3. Sample description

Normalization of samples names

Extract information relative to PCR replicates and sample names.

```
sample_names_split = strsplit(as.character(sample_names), "_R")  
  
replicate = sapply(sample_names_split, function(x) x[length(x)])  
sample_id = sapply(sample_names_split, function(x) x[1])  
  
samples_desc = data.frame(name = samples(SPER01.raw)$sample, replicate = replicate, sample_id = sample_id)  
  
SPER01.raw@samples = samples_desc
```

3.1. Categorize MOTUs

DNA Sequence of the 6 synthetic sequences used as SPER01 positive controls.

```
Standard1 = "taagtctcgactagttgtgacctaacgaatagagaattctataagacgtgtgtcccat"  
Standard2 = "gtgtatggtatatttgaataatattaaatagaatttaaatcaatctttacatcgcttaata"  
Standard3 = "cacaatgctcggtactagaagcatttgta"  
Standard4 = "attgaatgaaaagattattcgatatagaat"  
Standard5 = "agaacgctagaatctaagatgggggggggatgagtaagatatttatcagtaacatatga"
```

```
Standard6 = "atTTTtGtaactcattaacaatTTTTTTTTTgatgtatcataagTactaaactagttact"
```

- Identify which MOTUs are corresponding to these positive control sequences and associated them to their corresponding category.
- All the MOTUs exhibiting a similarity with one of the reference SPER01 database greater than 95% is tagged as SPER01
- The remaining sequences are tagged as Unknown

```
sequence_type = rep("Unknown", nrow(motus(SPER01.raw)))
sequence_type[which(motus(SPER01.raw)$`best_identity:db_GH` > 0.95)] = "SPER01"
sequence_type[which(motus(SPER01.raw)$sequence == Standard1)] = "standard1"
sequence_type[which(motus(SPER01.raw)$sequence == Standard2)] = "standard2"
sequence_type[which(motus(SPER01.raw)$sequence == Standard3)] = "standard3"
sequence_type[which(motus(SPER01.raw)$sequence == Standard4)] = "standard4"
sequence_type[which(motus(SPER01.raw)$sequence == Standard5)] = "standard5"
sequence_type[which(motus(SPER01.raw)$sequence == Standard6)] = "standard6"
```

```
SPER01.raw@motus$sequence_type = as.factor(sequence_type)
table(SPER01.raw@motus$sequence_type)
```

SPER01	standard1	standard2	standard3	standard4	standard5	standard6	Unknown
44125	1	1	1	1	1	1	36419

4. Curation procedure

4.1. Select motus occuring at least at 1% in at least one PCR

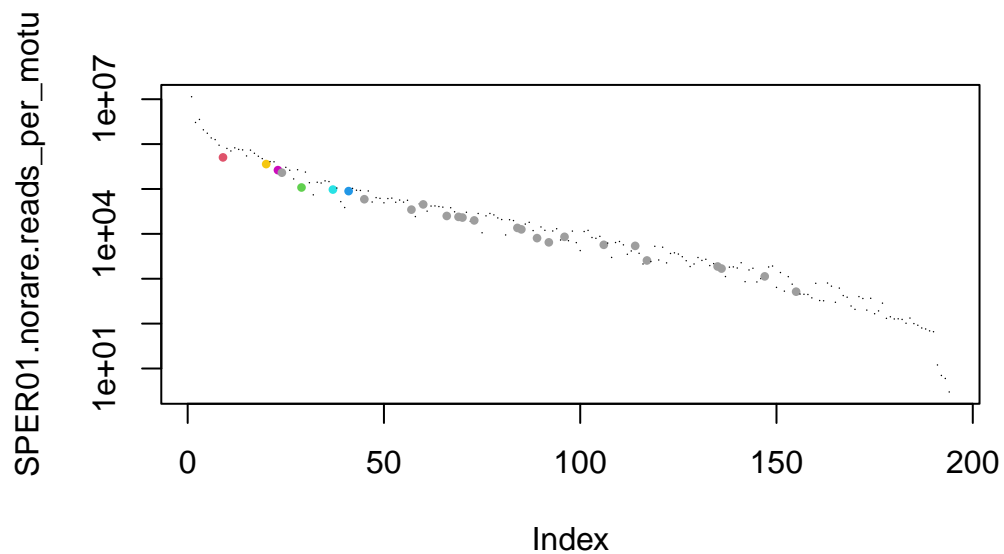
Only MOTUs occurring at least at one percent in at least one PCR are conserved. The others are discarded and correspond to few rare taxa, and many spurious MOTUs generated by PCR artefacts.

```
norare = apply(decostand(reads(SPER01.raw), method = "total"),
               MARGIN = 2,
               FUN = max) >= 0.01
table(norare)
```

```
norare
FALSE TRUE
80356  194
```

```
SPER01.norare <- SPER01.raw[,which(norare)]
```

```
SPER01.norare.reads_per_motus = colSums(reads(SPER01.norare))
plot(SPER01.norare.reads_per_motus, log="y",
     cex=0.1 + 0.5 * (SPER01.norare@motus$sequence_type!="SPER01"),
     col = as.integer(SPER01.norare@motus$sequence_type), pch=16)
```



4.2. Analysis of the PCR Positive controls

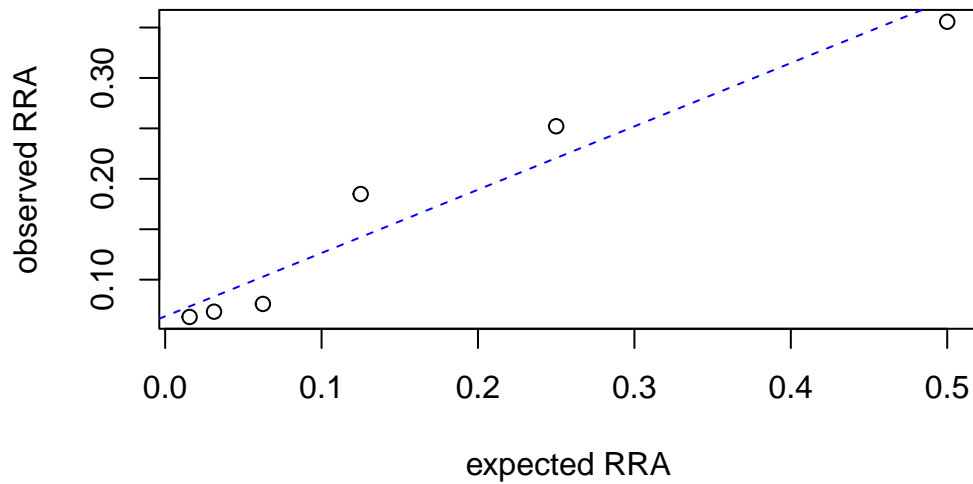
Synthetic MOTUs are extracted from the data set and their relative read abundances (RRA) is plotted as a function of their theoretical abundances to check the quality of the PCR.

```
rp <- SPER01.norare.reads_per_motus[! SPER01.norare@motus$sequence_type %in% c("SPER01","Unknown")]
rp/sum(rp)

GHP2_00000016 GHP3_00000014 GHP3_00000156 GHP2_00000010 GHP3_00000144
0.35574603    0.25211192    0.18491040    0.07594321    0.06826363
GHP2_00000144
0.06302481

expectedRRA <- 1/2^(1:6)
observedRRA <- rp/sum(rp)

plot(expectedRRA,observedRRA,
      xlab="expected RRA",
      ylab="observed RRA")
abline(lm(observedRRA ~ expectedRRA),
      col = "blue", lty = 2)
```



4.3. Filtering for PCR outliers

Only library 1 and 2 have individually tagged PCR replicates.

```
library_3.ids = read.csv("Data/samples_library_3.txt",
                        stringsAsFactors = FALSE,
                        header = FALSE)[,1]

library3.keep = gsub("_R.?$", "_R", rownames(SPER01.norare)) %in% library_3.ids
SPER01.lib3 = SPER01.norare[library3.keep,]
SPER01.lib12= SPER01.norare[!library3.keep,]

dim(SPER01.lib3)
```

```
[1] 65 194
```

```
dim(SPER01.lib12)
```

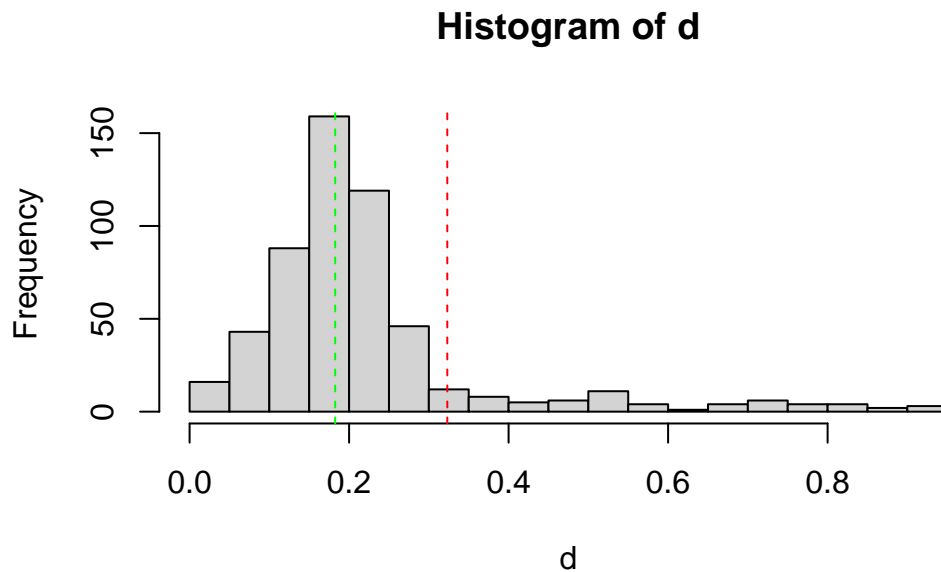
```
[1] 541 194
```

Load the script containing the selection procedure implemented in function `tag_bad_pcr`.

```
source("select_pcr.R")
```

4.3.1. First selection round

```
keep1 = tag_bad_pcr(samples = samples(SPER01.lib12)$sample_id,
                    counts = reads(SPER01.lib12),
                    plot = TRUE
                    )
```



Histogram shows the empirical distribution of the PCR replicate distances. The red vertical dashed line indicates the threshold used to discard outlier PCRs. The green vertical dashed line indicates the mode of the observed distribution.

```
table(keep1$keep)
```

```
FALSE  TRUE
   44   497
```

FALSE is the count of PCR to discard, TRUE the count of PCR conserved at the end of this selection round.

```
samples(SPER01.lib12)$name[!keep1$keep]
```

```
[1] "DNANC12_R3" "DNANC14_R3" "DNANC15_R3" "DNANC_10_R2" "DNANC_11_R3"
[6] "DNANC_13_R3" "DNANC_14_R3" "DNANC_15_R3" "DNANC_7_R2" "DNANC_8_R2"
[11] "DNANC_9_R2" "PCRNC_3_R3" "PCRNC_6_R3" "PCRPOS_3_R3" "PCRPOS_4_R2"
[16] "PCRPOS_5_R3" "PCRPOS_6_R3" "X_29_R1" "X_38_R3" "X_50_R2"
[21] "X_64_R1" "X_66_R1" "Y_24_R2" "Y_28_R2" "Y_2_R1"
[26] "Y_33_R2" "Y_36_R2" "Y_44_R1" "Y_45_R1" "Y_46_R3"
[31] "Y_47_R1" "Y_48_R1" "Y_48_R2" "Y_49_R2" "Y_51_R2"
[36] "Y_52_R1" "Y_56_R3" "Z_20_R1" "Z_30_R1" "Z_36_R1"
```

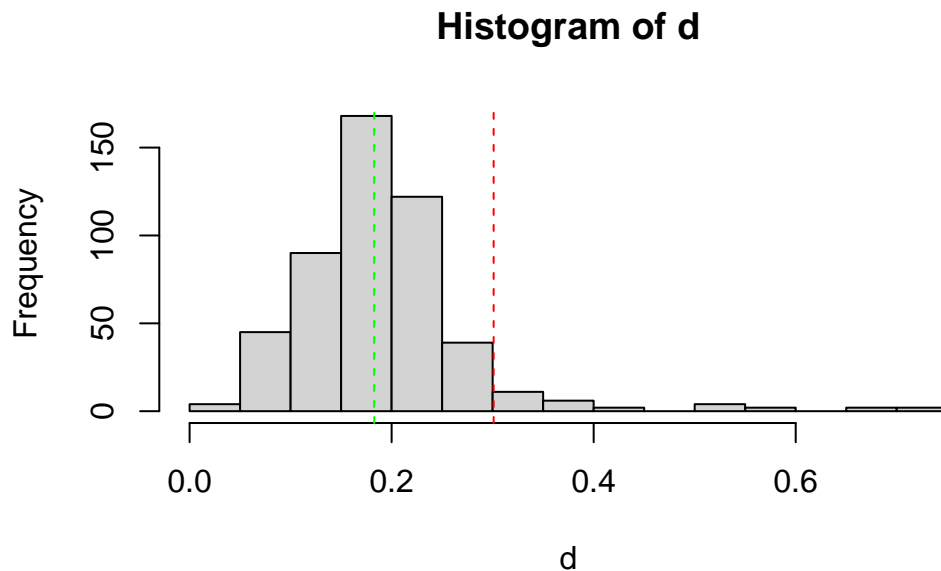
```
[41] "Z_45_R2"      "Z_53_R1"      "Z_5_R1"       "Z_61_R3"
```

Above is the list of the ids of the discarded PCRs.

```
SPER01.lib12.k1 = SPER01.lib12[keep1$keep,]
```

4.3.2. Second selection round

```
keep2 = tag_bad_pcr(samples = samples(SPER01.lib12.k1)$sample_id,
                    counts = reads(SPER01.lib12.k1),
                    plot = TRUE
                    )
```



```
table(keep2$keep)
```

```
FALSE  TRUE
   24   473
```

```
samples(SPER01.lib12.k1)$name[!keep2$keep]
```

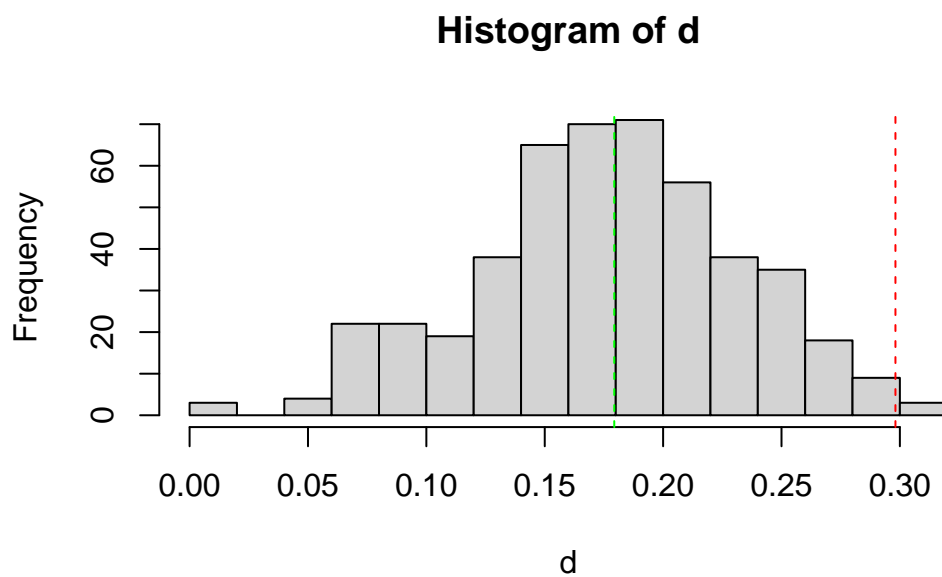
```
[1] "PCRPOS_3_R2" "X_50_R1"      "X_50_R3"      "Y_18_R1"      "Y_2_R2"
[6] "Y_2_R3"      "Y_31_R1"      "Y_44_R2"      "Y_46_R1"      "Y_46_R2"
[11] "Y_47_R2"     "Y_49_R1"      "Y_49_R3"      "Y_56_R1"      "Y_56_R2"
[16] "Y_71_R1"     "Z_30_R2"      "Z_30_R3"      "Z_45_R1"      "Z_49_R1"
[21] "Z_53_R2"     "Z_53_R3"      "Z_61_R1"      "Z_61_R2"
```



```
SPER01.lib12.k2 = SPER01.lib12.k1[keep2$keep,]
```

4.3.3. Third selection round

```
keep3 = tag_bad_pcr(samples = samples(SPER01.lib12.k2)$sample_id,
                    counts = reads(SPER01.lib12.k2),
                    plot = TRUE
                    )
```



```
table(keep3$keep)
```

```
FALSE  TRUE
    7   466
```

```
keep3[!keep3$keep,]
```

	samples	distance	maximum	repeats	keep
X_14_R3	X_14	0.2982691	0.2982691	3	FALSE
Y_44_R3	Y_44	0.0000000	0.0000000	1	FALSE
Y_47_R3	Y_47	0.0000000	0.0000000	1	FALSE
Y_50_R3	Y_50	0.2986271	0.2986271	3	FALSE
Y_52_R3	Y_52	0.3004914	0.3004914	2	FALSE
Z_45_R3	Z_45	0.0000000	0.0000000	1	FALSE
Z_55_R3	Z_55	0.3004595	0.3004595	3	FALSE

```
SPER01.lib12.k3 = SPER01.lib12.k2[keep3$keep,]
```

4.3.4. Merge remaining PCR replicates

```
freq = decostand(reads(SPER01.lib12.k3),
                 method = "total")
SPER01.lib12.k3$count = reads(SPER01.lib12.k3)
SPER01.lib12.k3@reads = freq

SPER01.merged = aggregate(SPER01.lib12.k3, MARGIN = 1, by = list(sample_id=samples(SPER01.lib12.k3))
```

Look for controls left in library 1 and 2

```
rownames(SPER01.merged)

[1] "X_10" "X_11" "X_12" "X_14" "X_15" "X_16" "X_17" "X_18" "X_19" "X_2"
[11] "X_20" "X_21" "X_22" "X_23" "X_24" "X_25" "X_26" "X_27" "X_28" "X_29"
[21] "X_3"  "X_30" "X_31" "X_33" "X_34" "X_35" "X_36" "X_37" "X_38" "X_39"
[31] "X_4"  "X_41" "X_42" "X_44" "X_51" "X_53" "X_54" "X_56" "X_57" "X_59"
[41] "X_60" "X_63" "X_64" "X_65" "X_66" "X_68" "X_70" "X_74" "X_75" "X_76"
[51] "X_77" "X_78" "X_79" "X_80" "X_9"  "Y_1"  "Y_11" "Y_13" "Y_14" "Y_18"
[61] "Y_21" "Y_23" "Y_24" "Y_25" "Y_26" "Y_28" "Y_29" "Y_3"  "Y_31" "Y_32"
[71] "Y_33" "Y_34" "Y_36" "Y_38" "Y_39" "Y_4"  "Y_40" "Y_41" "Y_42" "Y_43"
[81] "Y_45" "Y_5"  "Y_50" "Y_51" "Y_52" "Y_53" "Y_57" "Y_58" "Y_59" "Y_6"
[91] "Y_61" "Y_69" "Y_7"  "Y_70" "Y_71" "Y_72" "Y_74" "Y_8"  "Y_9"  "Z_1"
[101] "Z_10" "Z_11" "Z_12" "Z_13" "Z_14" "Z_15" "Z_16" "Z_17" "Z_18" "Z_19"
[111] "Z_21" "Z_22" "Z_23" "Z_24" "Z_25" "Z_27" "Z_28" "Z_3"  "Z_31" "Z_32"
[121] "Z_33" "Z_34" "Z_35" "Z_36" "Z_37" "Z_38" "Z_4"  "Z_40" "Z_42" "Z_43"
[131] "Z_44" "Z_46" "Z_48" "Z_49" "Z_5"  "Z_51" "Z_52" "Z_54" "Z_55" "Z_56"
[141] "Z_59" "Z_6"  "Z_60" "Z_62" "Z_63" "Z_65" "Z_66" "Z_67" "Z_68" "Z_69"
[151] "Z_7"  "Z_70" "Z_71" "Z_72" "Z_73" "Z_74" "Z_75" "Z_76" "Z_77" "Z_78"
[161] "Z_79" "Z_8"  "Z_80"
```

4.4. Merge lib 1,2 and 3

4.4.1. Remove controls in library 3

```
rownames(SPER01.lib3)

[1] "DNANC_3_R1" "DNANC_4_R1" "DNANC_5_R1" "DNANC_6_R1" "DNANC_6_R2"
[6] "PCRCNC_1_R1" "PCRCNC_2_R1" "PCRCNC_2_R2" "PCRPOS_1_R1" "PCRPOS_2_R1"
[11] "PCRPOS_2_R2" "X_1_R1" "X_32_R1" "X_40_R1" "X_43_R1"
[16] "X_45_R1" "X_46_R1" "X_47_R1" "X_48_R1" "X_49_R1"
[21] "X_52_R1" "X_55_R1" "X_58_R1" "X_5_R1" "X_62_R1"
[26] "X_67_R1" "X_69_R1" "X_6_R1" "X_71_R1" "X_72_R1"
[31] "X_73_R1" "X_7_R1" "X_8_R1" "Y_10_R1" "Y_12_R1"
[36] "Y_15_R1" "Y_16_R1" "Y_17_R1" "Y_19_R1" "Y_20_R1"
[41] "Y_22_R1" "Y_27_R1" "Y_30_R1" "Y_35_R1" "Y_37_R1"
[46] "Y_54_R1" "Y_55_R1" "Y_60_R1" "Y_62_R1" "Y_63_R1"
[51] "Y_64_R1" "Y_65_R1" "Y_66_R1" "Y_67_R1" "Y_68_R1"
[56] "Y_73_R1" "Z_26_R1" "Z_29_R1" "Z_2_R1" "Z_41_R1"
[61] "Z_47_R1" "Z_50_R1" "Z_57_R1" "Z_58_R1" "Z_9_R1"
```

```
SPER01.lib3.samples = SPER01.lib3[-(1:11),]
rownames(SPER01.lib3.samples@reads) = sub("_R.?$","",rownames(SPER01.lib3.samples))
rownames(SPER01.lib3.samples)
```

```
[1] "X_1" "X_32" "X_40" "X_43" "X_45" "X_46" "X_47" "X_48" "X_49" "X_52"
[11] "X_55" "X_58" "X_5" "X_62" "X_67" "X_69" "X_6" "X_71" "X_72" "X_73"
[21] "X_7" "X_8" "Y_10" "Y_12" "Y_15" "Y_16" "Y_17" "Y_19" "Y_20" "Y_22"
[31] "Y_27" "Y_30" "Y_35" "Y_37" "Y_54" "Y_55" "Y_60" "Y_62" "Y_63" "Y_64"
[41] "Y_65" "Y_66" "Y_67" "Y_68" "Y_73" "Z_26" "Z_29" "Z_2" "Z_41" "Z_47"
[51] "Z_50" "Z_57" "Z_58" "Z_9"
```

4.4.2. Merge library 1, 2 and 3

```
SPER01.lib123.reads = rbind(SPER01.merged@reads,
                           decostand(SPER01.lib3.samples@reads,method = "total"))

common = intersect(names(SPER01.merged@samples),
                   names(SPER01.lib3.samples@samples))

SPER01.lib123.samples = rbind(SPER01.merged@samples[,common],
                              SPER01.lib3.samples@samples[,common])

SPER01.lib123 = metabarcoding.data(reads = decostand(SPER01.lib123.reads,method = "total"),
                                   samples = SPER01.lib123.samples,
                                   motus = SPER01.merged@motus)

dim(SPER01.lib123)
```

```
[1] 217 194
```

```
SPER01.lib123@samples$animal_id = sapply(SPER01.lib123@samples$sample_id,
                                          function(x) strsplit(as.character(x),"_")[[1]][1])
```

4.4.3. Check for empty MOTUs

Look at MOTUs still present in the data matrix, but represented by no more reads because of the filtering procedure.

```
zero = colSums(reads(SPER01.lib123)) == 0
table(zero)
```

```
zero
FALSE TRUE
184    10
```

```
SPER01.nozero = SPER01.lib123[,!zero]
```

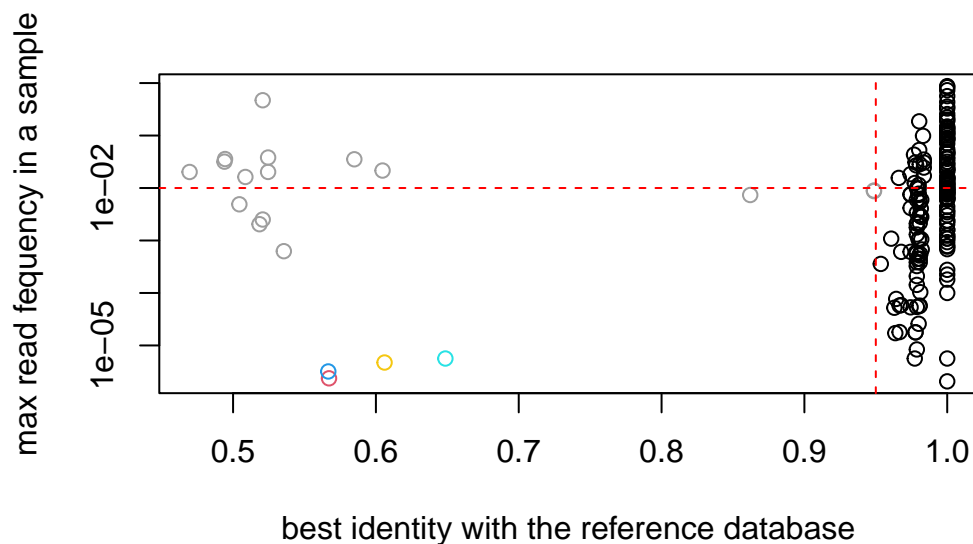
```
table(SPER01.nozero@motus$sequence_type)
```

```
SPER01 standard1 standard2 standard3 standard4 standard5 standard6 Unknown
```

164 1 0 1 1 0 1 16

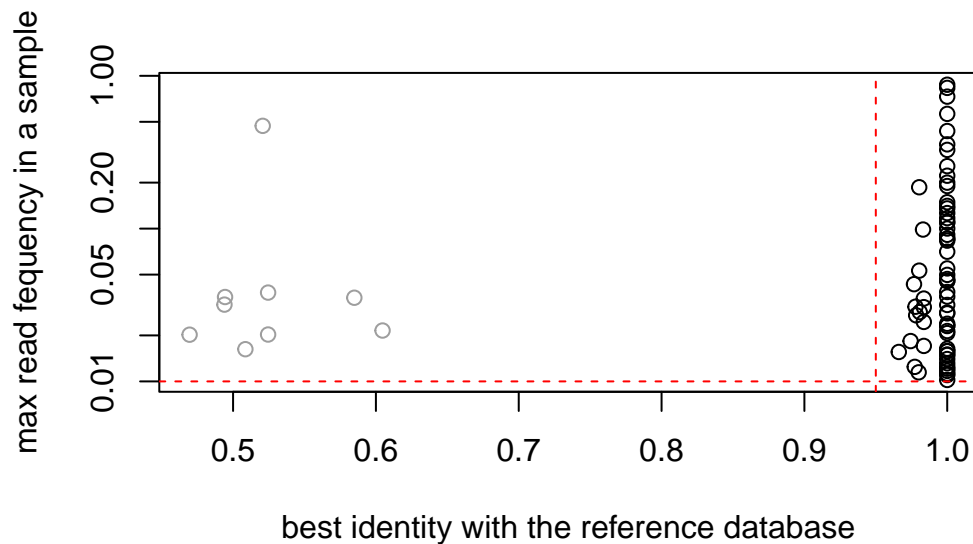
5. Filter out rare species

```
plot(SPER01.nozero@motus$`best_identity:db_GH`,
     apply(reads(SPER01.nozero),2,max),
     col=as.factor(SPER01.nozero@motus$sequence_type),
     log="y",
     ylab="max read frequency in a sample",
     xlab="best identity with the reference database")
abline(h=0.01,col="red",lty=2)
abline(v=0.95,col="red",lty=2)
```



```
SPER01.merged3 = SPER01.lib123[, apply(reads(SPER01.lib123),2,max) > 0.01]
```

```
plot(SPER01.merged3$motus$`best_identity:db_GH`,
     apply(reads(SPER01.merged3),2,max),
     col=as.factor(SPER01.merged3$motus$sequence_type),
     log="y",
     ylab="max read frequency in a sample",
     xlab="best identity with the reference database")
abline(h=0.01,col="red",lty=2)
abline(v=0.95,col="red",lty=2)
```



```
which(SPER01.merged3@motus$sequence_type == "Unknown" & apply(reads(SPER01.merged3),2,max) > 0.4)
```

```
GHP3_00000038
21
```

```
SPER01.merged3@motus[21,"sequence"]
```

```
[1] "tatagggttttcttggtgtatttcacaccgaaccaggatgggcatgcaaacaggttggtctcggtagttcagccctcgccatcggcaggggatttt"
```

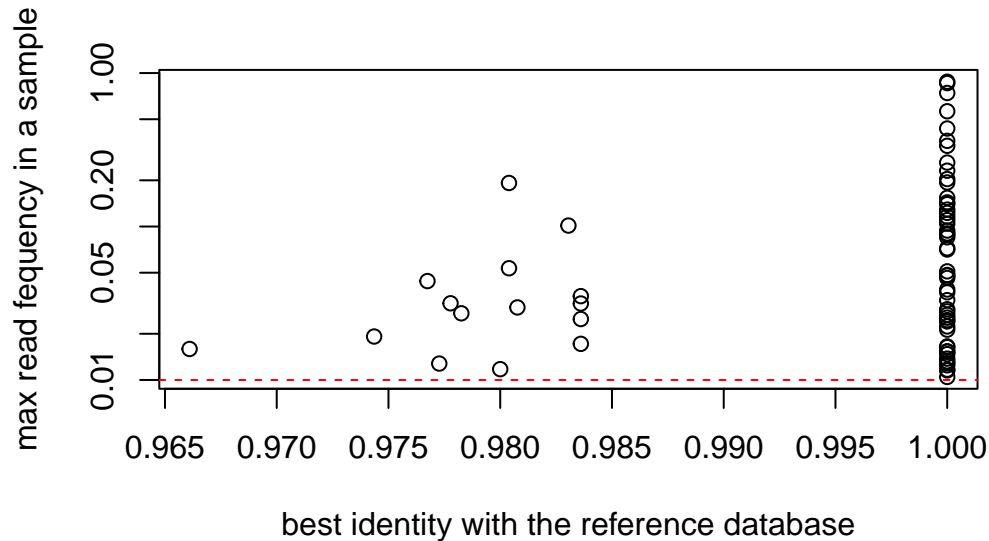
Looks like nothing at embl by blast

5.1. Keep only MOTUs Strictly identical to one of the reference sequence

5.1.1. First level stringency filter (95% identity)

```
SPER01.merged4 = SPER01.merged3[,SPER01.merged3@motus$`best_identity:db_GH` > 0.95]
SPER01.merged4@reads = decostand(SPER01.merged4@reads,method = "total")
SPER01.merged4@motus <- SPER01.merged4@motus %>% select(-starts_with("obiclean_status:"))
```

```
plot(SPER01.merged4$motus$`best_identity:db_GH`,
     apply(reads(SPER01.merged4),2,max),
     col=as.factor(SPER01.merged4$motus$sequence_type),
     log="y",
     ylab="max read frequency in a sample",
     xlab="best identity with the reference database")
abline(h=0.01,col="red",lty=2)
```



5.1.2. High stringency filtering (100% identity)

```
spermatophyta.taxid <- ecofind(taxo,patterns = "~Spermatophyta$")
SPER01.merged4@motus$is_spermatophyta <- is.subcladeof(taxo,SPER01.merged4@motus$taxid,spermatophyta)
table(SPER01.merged4@motus$is_spermatophyta)
```

```
FALSE  TRUE
      6    60
```

```
SPER01.merged4@motus %>% filter(!is_spermatophyta)
```

	id	best_identity:db_GH	best_match:db_GH	count	family
GHP1_00000082	GHP1_00000082	1	KY427333	137903	1203520
GHP1_00000361	GHP1_00000361	1	AF515231	49169	1203520
GHP3_00000569	GHP3_00000569	1	AF192562	56685	13803
GHP3_00000419	GHP3_00000419	1	HE993635	22646	1203500
GHP1_00000874	GHP1_00000874	1	AJ133265	13008	3250
GHP1_00008815	GHP1_00008815	1	AF023727	2987	52989

	family_name	genus	genus_name	match_count:db_GH	rank
GHP1_00000082	Athyriaceae	32109	Athyrium	1	species
GHP1_00000361	Athyriaceae	32109	Athyrium	1	genus
GHP3_00000569	Sphagnaceae	13804	Sphagnum	1	genus
GHP3_00000419	Cystopteridaceae	32115	Gymnocarpium	1	species
GHP1_00000874	Lycopodiaceae	NA	<NA>	1	subfamily
GHP1_00008815	Orthotrichaceae	NA	<NA>	1	family

	scientific_name	species	species_list:db_GH
GHP1_00000082	Athyrium sinense	672195	['Athyrium sinense']
GHP1_00000361	Athyrium	NA	[]

GHP3_00000569	Sphagnum	NA	[]
GHP3_00000419	Gymnocarpium dryopteris	32116	['Gymnocarpium dryopteris']
GHP1_00000874	Lycopodioideae	NA	[]
GHP1_00008815	Orthotrichaceae	NA	[]
	species_name	taxid	
GHP1_00000082	Athyrium sinense	672195	
GHP1_00000361	<NA>	32109	
GHP3_00000569	<NA>	13804	
GHP3_00000419	Gymnocarpium dryopteris	32116	
GHP1_00000874	<NA>	1965347	
GHP1_00008815	<NA>	52989	
	sequence	sequence_type	
GHP1_00000082	atcttggtattattcggatgaatttcgggcatgaggcga	SPER01	
GHP1_00000361	atcttggtattattcagatgaatttcgggcatgaggcga	SPER01	
GHP3_00000569	atcttggtttcataacataaatgg	SPER01	
GHP3_00000419	atcttggtattactcaaataatgaatttcgggcaatgaggcaa	SPER01	
GHP1_00000874	atcctgttttagcaaataatggcgg	SPER01	
GHP1_00008815	atattattttattttaaaaaataa	SPER01	
	is_spermatophyta		
GHP1_00000082	FALSE		
GHP1_00000361	FALSE		
GHP3_00000569	FALSE		
GHP3_00000419	FALSE		
GHP1_00000874	FALSE		
GHP1_00008815	FALSE		

```

musaceae.taxid <- ecofind(taxo,patterns = "^Musaceae$")
to_keep <- ! (is.subcladeof(taxo,SPER01.merged4@motus$taxid,musaceae.taxid) | SPER01.merged4@motus$
is.subcladeof(taxo,SPER01.merged4@motus$taxid,spermatophyta.taxid) &
SPER01.merged4@motus$`best_identity:db_GH` == 1
table(to_keep)

to_keep
FALSE TRUE
22 45

```

```

SPER01.merged4@motus %>% filter(!to_keep)

id best_identity:db_GH best_match:db_GH count
GHP2_00000044 GHP2_00000044 0.9803922 AB817362 2080755
GHP2_00000355 GHP2_00000355 0.9807692 AF098856 303606
GHP2_00000136 GHP2_00000136 0.9767442 DQ359689 261620
GHP1_00000082 GHP1_00000082 1.0000000 KY427333 137903
GHP2_00000549 GHP2_00000549 0.9830508 KX872610 135866
GHP1_00000014 GHP1_00000014 0.9803922 AB817372 52144
GHP2_00000171 GHP2_00000171 0.9743590 AC183493 90296
GHP2_00000260 GHP2_00000260 0.9777778 EF440558 51279
GHP2_00000820 GHP2_00000820 0.9772727 AJ505541 61937
GHP1_00000361 GHP1_00000361 1.0000000 AF515231 49169
GHP2_00000554 GHP2_00000554 0.9782609 AJ430966 62557
GHP3_00000569 GHP3_00000569 1.0000000 AF192562 56685

```

GHP1_00000681	GHP1_00000681	0.9836066	AB979732	24268
GHP3_00000153	GHP3_00000153	0.9836066	AB979732	19101
GHP1_00003018	GHP1_00003018	0.9836066	AB979732	10572
GHP1_00000030	GHP1_00000030	0.9836066	AB979732	25179
GHP3_00000419	GHP3_00000419	1.0000000	HE993635	22646
GHP1_00000874	GHP1_00000874	1.0000000	AJ133265	13008
GHP2_00008966	GHP2_00008966	0.9661017	KX872610	7043
GHP2_00009529	GHP2_00009529	0.9800000	AY344156	6375
GHP1_00008815	GHP1_00008815	1.0000000	AF023727	2987
GHP3_00027523	GHP3_00027523	1.0000000	AB817687	2256

	family	family_name	genus	genus_name	match_count:db_GH
GHP2_00000044	NA	<NA>	NA	<NA>	33
GHP2_00000355	4210	Asteraceae	NA	<NA>	6
GHP2_00000136	3440	Ranunculaceae	3445	Ranunculus	1
GHP1_00000082	1203520	Athyriaceae	32109	Athyrium	1
GHP2_00000549	4479	Poaceae	NA	<NA>	NA
GHP1_00000014	3745	Rosaceae	NA	<NA>	14
GHP2_00000171	3700	Brassicaceae	NA	<NA>	1
GHP2_00000260	3318	Pinaceae	3337	Pinus	2
GHP2_00000820	4136	Lamiaceae	NA	<NA>	2
GHP1_00000361	1203520	Athyriaceae	32109	Athyrium	1
GHP2_00000554	NA	<NA>	NA	<NA>	15
GHP3_00000569	13803	Sphagnaceae	13804	Sphagnum	1
GHP1_00000681	3514	Betulaceae	NA	<NA>	1
GHP3_00000153	3514	Betulaceae	NA	<NA>	1
GHP1_00003018	3514	Betulaceae	NA	<NA>	1
GHP1_00000030	3514	Betulaceae	NA	<NA>	1
GHP3_00000419	1203500	Cystopteridaceae	32115	Gymnocarpium	1
GHP1_00000874	3250	Lycopodiaceae	NA	<NA>	1
GHP2_00008966	4479	Poaceae	NA	<NA>	11
GHP2_00009529	14101	Juncaceae	13578	Juncus	2
GHP1_00008815	52989	Orthotrichaceae	NA	<NA>	1
GHP3_00027523	4637	Musaceae	NA	<NA>	1

	rank	scientific_name	species
GHP2_00000044	order	Asterales	NA
GHP2_00000355	subfamily	Asteroideae	NA
GHP2_00000136	genus	Ranunculus	NA
GHP1_00000082	species	Athyrium sinense	672195
GHP2_00000549	<NA>	<NA>	NA
GHP1_00000014	tribe	Maleae	NA
GHP2_00000171	tribe	Brassicaceae	NA
GHP2_00000260	subgenus	Pinus	NA
GHP2_00000820	tribe	Mentheae	NA
GHP1_00000361	genus	Athyrium	NA
GHP2_00000554	order	Asterales	NA
GHP3_00000569	genus	Sphagnum	NA
GHP1_00000681	family	Betulaceae	NA
GHP3_00000153	family	Betulaceae	NA
GHP1_00003018	family	Betulaceae	NA
GHP1_00000030	family	Betulaceae	NA
GHP3_00000419	species	Gymnocarpium dryopteris	32116
GHP1_00000874	subfamily	Lycopodioideae	NA

GHP2_00008966	no rank	Poeae Chloroplast Group 1 (Aveneae type)	NA
GHP2_00009529	genus	Juncus	NA
GHP1_00008815	family	Orthotrichaceae	NA
GHP3_00027523	family	Musaceae	NA
GHP2_00000044			
GHP2_00000355			
GHP2_00000136			
GHP1_00000082			
GHP2_00000549			
GHP1_00000014			['Photinia loriformis', 'Docynia
GHP2_00000171			
GHP2_00000260			
GHP2_00000820			
GHP1_00000361			
GHP2_00000554		['Cymbonotus lawsonianus', 'Saussurea nematolepis', 'Helichrysum glumaceum', 'Villarsia	
GHP3_00000569			
GHP1_00000681			
GHP3_00000153			
GHP1_00003018			
GHP1_00000030			
GHP3_00000419			
GHP1_00000874			
GHP2_00008966			['Chascolytrum itatiaiae', ']
GHP2_00009529			
GHP1_00008815			
GHP3_00027523			
	species_name	taxid	
GHP2_00000044	<NA>	4209	
GHP2_00000355	<NA>	102804	
GHP2_00000136	<NA>	3445	
GHP1_00000082	Athyrium sinense	672195	
GHP2_00000549	<NA>	NA	
GHP1_00000014	<NA>	721813	
GHP2_00000171	<NA>	981071	
GHP2_00000260	<NA>	139271	
GHP2_00000820	<NA>	216718	
GHP1_00000361	<NA>	32109	
GHP2_00000554	<NA>	4209	
GHP3_00000569	<NA>	13804	
GHP1_00000681	<NA>	3514	
GHP3_00000153	<NA>	3514	
GHP1_00003018	<NA>	3514	
GHP1_00000030	<NA>	3514	
GHP3_00000419	Gymnocarpium dryopteris	32116	
GHP1_00000874	<NA>	1965347	
GHP2_00008966	<NA>	1652080	
GHP2_00009529	<NA>	13578	
GHP1_00008815	<NA>	52989	
GHP3_00027523	<NA>	4637	
		sequence	
GHP2_00000044		atcacgttttccgaaaacaacaaagggttcagaaagcgaaaataaaaaag	

```

GHP2_00000355      atcacgttttccgaaaacaaacaaagggttcagaaagcgaagaaaaaa
GHP2_00000136      atcctgttttcagaaaacaaaagggttcagaaagcaaagg
GHP1_00000082      atcttgattatttcggatgaatttcgggcgatgaggcga
GHP2_00000549      atccgtgttttgagaaaacaaaggggttctcgaatcgaactataatacaaaggaaaag
GHP1_00000014      atcctgttttatgaaaataaacaagggttcataaacgaaaataaaaaag
GHP2_00000171      atcatgggttacgcgaacaaacaaagtttagaaagcgg
GHP2_00000260      atccggttcataagacaatgtttcttctcctaagataggaagg
GHP2_00000820      atcctgttttccaaaacaaagggtttcaaaaacgaaaaaaag
GHP1_00000361      atcttgattatttcagatgaatttcgggcgatgaggcga
GHP2_00000554      atcacgttttccgaaaacaaagggttcagaaagcgaagatcaaaaag
GHP3_00000569      atcttgttttcataacataaatgg
GHP1_00000681      gtcctgttttccgaaaacaaataaaacaaatttaagggttcataaagtgagaataaaaaag
GHP3_00000153      ctctgttttccgaaaacaaataaaacaaatttaagggttcataaagtgagaataaaaaag
GHP1_00003018      ttctgttttccgaaaacaaataaaacaaatttaagggttcataaagtgagaataaaaaag
GHP1_00000030      tcctgttttccgaaaacaaataaaacaaatttaagggttcataaagtgagaataaaaaag
GHP3_00000419      atcttgattactcaaatgaatttcgggcaatgaggcaa
GHP1_00000874      atcctgttttagcaaatggcgg
GHP2_00008966      atccgtgttttgagaaaacaaaggggttctcaaactgaactataatacaaaggaaaag
GHP2_00009529      gtctttattttgataaaattgtttttatagaaaaattcaaatcaaaaaa
GHP1_00008815      atattattttatttaaaaaataa
GHP3_00027523      atccttattttgagaaaacaaagggtttataaaactagaattttaaag

```

	sequence_type	is_spermatophyta
GHP2_00000044	SPER01	TRUE
GHP2_00000355	SPER01	TRUE
GHP2_00000136	SPER01	TRUE
GHP1_00000082	SPER01	FALSE
GHP2_00000549	SPER01	NA
GHP1_00000014	SPER01	TRUE
GHP2_00000171	SPER01	TRUE
GHP2_00000260	SPER01	TRUE
GHP2_00000820	SPER01	TRUE
GHP1_00000361	SPER01	FALSE
GHP2_00000554	SPER01	TRUE
GHP3_00000569	SPER01	FALSE
GHP1_00000681	SPER01	TRUE
GHP3_00000153	SPER01	TRUE
GHP1_00003018	SPER01	TRUE
GHP1_00000030	SPER01	TRUE
GHP3_00000419	SPER01	FALSE
GHP1_00000874	SPER01	FALSE
GHP2_00008966	SPER01	TRUE
GHP2_00009529	SPER01	TRUE
GHP1_00008815	SPER01	FALSE
GHP3_00027523	SPER01	TRUE

```

SPER01.final <- SPER01.merged4[,which(to_keep)]
SPER01.final@reads <- decostand(SPER01.final@reads,method = "total")

```

6. Saving the filtered dataset

6.1. Updating the sample metadata

6.1.1. Adding samples metadata

```
metadata <- read_csv("Data/Faeces/metadata.csv",
                     show_col_types = FALSE)

SPERO1.final@samples %<>%
  select(sample_id, animal_id) %>%
  left_join(metadata, by = "sample_id") %>%
  mutate(id = sample_id) %>%
  column_to_rownames("id") %>%
  select(sample_id, animal_id, Sample_number, Date, Sample_time, times_from_birch, Fed_biomass)
```

6.1.2. Homogenize time from burch

Adds : - 6 hours to animal X, - 3 hours to animal Y, - 4 hours to animal 2

```
SPERO1.final@samples %<>%
  mutate(times_from_birch = times_from_birch +
         ifelse(animal_id == "X", 6,
               ifelse(animal_id == "Y", 3, 4)))

SPERO1.final@samples %<>%
  mutate(Animal_id = ifelse(animal_id == "X", "9/10",
                           ifelse(animal_id == "Y", "10/10", "12/10")))
```

6.1.3. Adds pellets consumption data

```
pellets <- read_tsv("Data/pellet_weigth.txt", show_col_types = FALSE) %>%
  mutate(Date = str_replace(Date, "2018", "18")) %>%
  separate(Date, c("d", "m", "y"), sep = "/") %>%
  mutate(d = as.integer(d)+1,
         m = as.integer(m),
         m = ifelse(d==32, m+1, m),
         d = ifelse(d==32, 1, d),
         d = sprintf("%02d", d),
         m = sprintf("%02d", m)) %>%
  unite(col="Date", d, m, y, sep="/") %>%
  pivot_longer(-Date, names_to = "Animal_id", values_to = "pellets")

SPERO1.final@samples %<>%
  left_join(pellets)
```

Joining with `by = join_by(Date, Animal_id)`

6.2. Only keep samples

```
SPER01.final <- SPER01.final[which(str_detect(SPER01.final@samples$sample_id,"^[XYZ]")),]
```

6.3. Updating count statistics

```
SPER01.final %<>%  
  update_motus_count() %>%  
  update_samples_count() %>%  
  clean_empty()
```

6.4. Add MOTUs Metadata

```
SPER01.final@motus %<>%  
  mutate(category = ifelse(is.subcladeof(taxo,taxid,spermatophyta.taxid),  
                           "Plant",  
                           "Lichen"))
```

6.5. Write CSV files

```
write_csv(SPER01.final@samples,  
          file = "Data/Faeces/FE.Spermatophyta.samples.samples.csv")  
write_csv(SPER01.final@motus,  
          file = "Data/Faeces/FE.Spermatophyta.samples.motus.csv")  
write_csv(SPER01.final@reads %>%  
          decostand(method = "total") %>%  
          as.data.frame() %>%  
          rownames_to_column("id"),  
          file = "Data/Faeces/FE.Spermatophyta.samples.reads.csv")
```

References

- [1] H. Wickham, M. Averick, J. Bryan, W. Chang, L. McGowan, R. François, G. Golemund, A. Hayes, L. Henry, J. Hester, M. Kuhn, T. Pedersen, E. Miller, S. Bache, K. Müller, J. Ooms, D. Robinson, D. Seidel, V. Spinu, K. Takahashi, D. Vaughan, C. Wilke, K. Woo, H. Yutani, [Welcome to the tidyverse](https://www.tidyverse.org), Journal of open source software 4 (43) (2019) 1686. doi: [10.21105/joss.01686](https://doi.org/10.21105/joss.01686).
URL <https://joss.theoj.org/papers/10.21105/joss.01686>
- [2] H. Wickham, [ggplot2: Elegant Graphics for Data Analysis](https://ggplot2.tidyverse.org), Springer-Verlag New York, 2016.
URL <https://ggplot2.tidyverse.org>