

MAGERI benchmark using reference standard DNA library

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Load metadata

```
library(stringr)
library(pROC)
```

```
## Type 'citation("pROC")' for a citation.
```

```
##
```

```
## Attaching package: 'pROC'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      cov, smooth, var
```

```
library(ggplot2)
```

```
library(ggbeeswarm)
```

```
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
df.vmeta <- read.table("hd734_variant_metadata.txt", sep="\t", header=T) # variants observed in HD734 a
```

```
df.smeta <- read.table("sample_metadata.txt", sep="\t", header=T) %>% filter(type == "blank") # metadat
```

VCF parsing function

```
read_vcf <- function(file_name) {
  .vcf <- read.table(file_name, header = F, sep = "\t", stringsAsFactors = F)
  colnames(.vcf) <- c("chromosome", "position", "skip1", "from", "to", "qual", "skip2", "info", "skip3")
  .vcf$skip1 <- NULL
  .vcf$skip2 <- NULL
  .vcf$skip3 <- NULL
  .vcf$skip4 <- NULL

  .vcf$qual <- as.integer(.vcf$qual)
  .vcf <- subset(.vcf, nchar(from) == 1 &
                 nchar(to) == 1 & !is.na(qual)) # no indels

  .infosplit <- str_split_fixed(.vcf$info, regex("[=;]"), 9)[,c(2, 4, 8)]

  .vcf$coverage <- as.numeric(.infosplit[,1])
}
```

```

.vcf$frequency <- as.numeric(.infosplit[,2])
.vcf$cqs <- as.numeric(.infosplit[,3])
.vcf$info <- NULL

.vcf$count <- as.integer(round(.vcf$coverage * .vcf$frequency))

.vcf
}

head(read_vcf("p126.h4_2_ballast_m1.vcf"))

## Warning in read_vcf("p126.h4_2_ballast_m1.vcf"): NAs introduced by coercion

##   chromosome position from to qual coverage   frequency   cqs count
## 1      chr2 212295704   C A    8      2265 0.0004415011 40.00000    1
## 2      chr2 212295705   C A    8      2265 0.0004415011 40.00000    1
## 3      chr2 212295713   A G    7      2265 0.0004415011 40.00000    1
## 4      chr2 212295718   G A    6      2265 0.0004415011 40.00000    1
## 5      chr2 212295725   C A    8      2264 0.0004416961 40.00000    1
## 6      chr2 212295732   C T   24      2264 0.0013250883 39.66667    3

Read samples with HD734 standard DNA and control human DNA, append metadata
df <- data.frame()

read_vcf_with_metadata <- function(file_name, primer_set, replica, ratio, type) {
  .vcf <- read_vcf(file_name)
  .vcf <- merge(.vcf, df.vmeta, all.x = type != "standard", all.y = F)
  .vcf$known.frequency <- .vcf$known.frequency * ratio
  .vcf$known.frequency[is.na(.vcf$known.frequency)] <- 0
  .vcf$primer_set <- primer_set
  .vcf$replica <- primer_set
  .vcf$type <- type

  .vcf <- subset(.vcf, frequency < 0.4 & count > 0) # remove alleles in control

  .vcf
}

for (i in 1:nrow(df.smeta)) {
  df <- with(df.smeta, rbind(df,
                             read_vcf_with_metadata(paste(prefix[i], "vcf", sep="."),
                                                         primer_set[i],
                                                         replica[i],
                                                         ratio[i],
                                                         type[i])))
}

## Warning in read_vcf(file_name): NAs introduced by coercion

## Warning in read_vcf(file_name): NAs introduced by coercion

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## Warning in read_vcf(file_name): NAs introduced by coercion

```

```
## Warning in read_vcf(file_name): NAs introduced by coercion
```

```
## Warning in read_vcf(file_name): NAs introduced by coercion
```

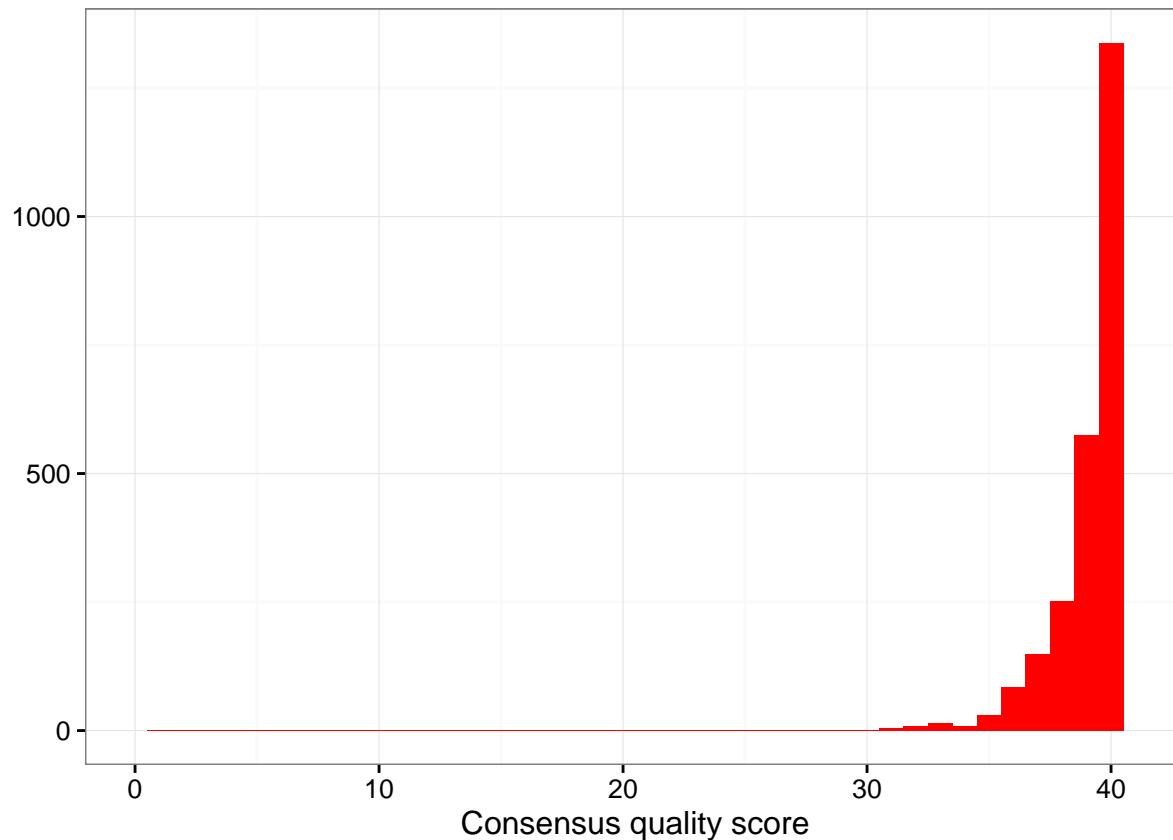
```
## Warning in read_vcf(file_name): NAs introduced by coercion
```

Histogram of CQS scores

```
fig13 <- ggplot(df, aes(cqs)) +  
  geom_histogram(fill="red", binwidth=1) +  
  ylab("") + scale_x_continuous("Consensus quality score", limits=c(0, 41)) +  
  theme_bw()
```

```
save(file = "../figures/fig13.Rda", fig13)
```

fig13



Expected HTS errors and observed error count

```
p_err <- 1-pbinom(3, size = 5, p=1e-2) + 0.5*dbinom(3, size = 5, p=1e-2)  
n_migs <- 100000  
df.1 <- data.frame(count = df$frequency * n_migs, type = "Observed")  
lambda <- p_err * n_migs  
df.1 <- rbind(df.1, data.frame(count = rpois(nrow(df), lambda), type="Expected"))
```

```
fig14 <- ggplot(df.1, aes(x=count, fill=type)) +  
  geom_histogram() +  
  ylab("") + scale_x_log10("Error count per 100,000 MIGs") +
```

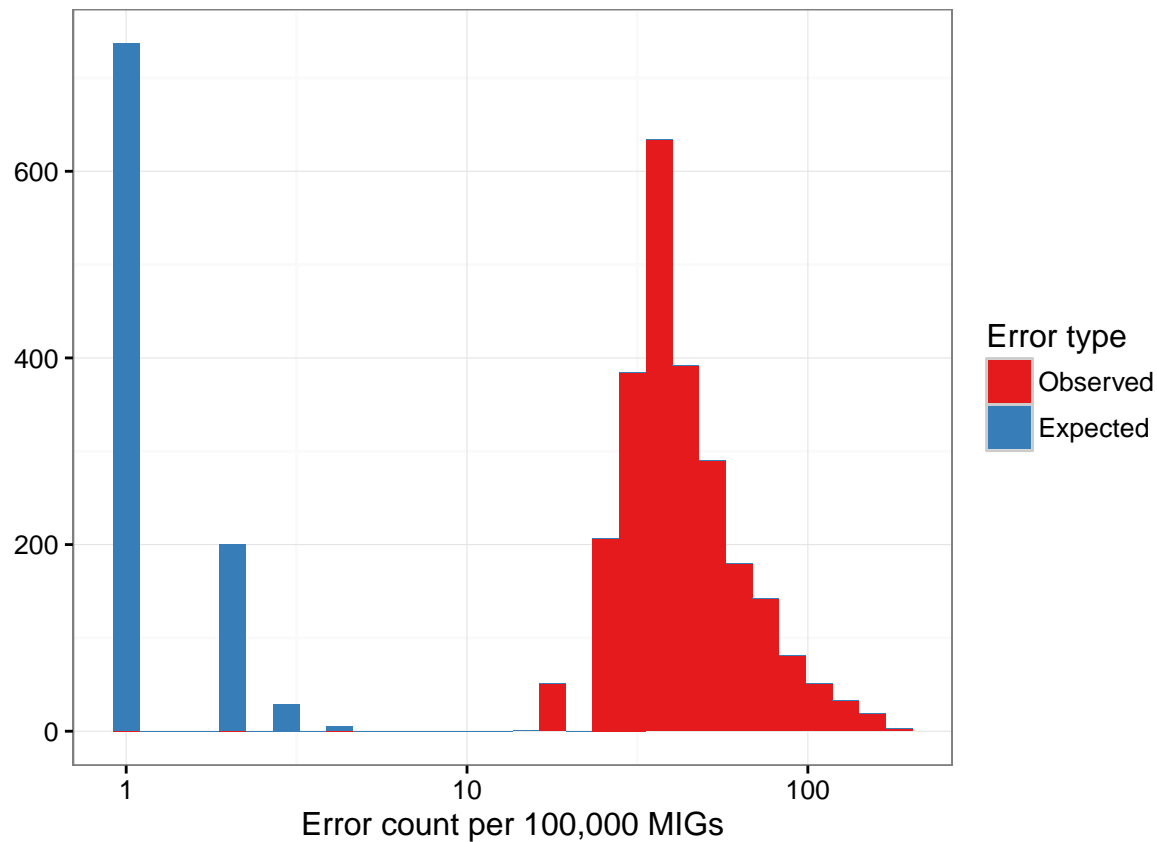
```
scale_fill_brewer("Error type", palette = "Set1") +
theme_bw()
```

```
save(file = "../figures/fig14.Rda", fig14)
```

```
fig14
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
## Warning: Removed 1497 rows containing non-finite values (stat_bin).
```



UMI coverage histogram

```
df.smeta.2 <- read.table("sample_metadata.txt", sep="\t", header=T)
```

```
df.1 <- data.frame()
```

```
for (i in 1:nrow(df.smeta.2)) {
  prefix <- df.smeta.2$prefix[i]
  .df <- read.table(paste0("proc_stat/", prefix, ".umi.histogram.txt"), header = T)
  .df$sample <- with(df.smeta.2, paste(type[i], primer_set[i], ratio[i]))
  .df$replica <- df.smeta.2$replica[i]
  df.1 <- rbind(df.1, .df)
}
```

```
fig12 <- ggplot(df.1, aes(x=mig.size.bin, y = read.count, color=sample, linetype=as.factor(replica))) +
  annotate(geom="rect", xmin=0,xmax=5, fill="grey", ymin=-Inf,ymax=Inf) +
  geom_line() +
```

```

ylab("Number of reads") +
scale_x_log10("MIG size", breaks = 2^(seq(0, 20, by=2)), limits=c(1,100000)) +
scale_color_brewer(palette = "Paired", guide=F) +
scale_linetype_discrete(guide=F)+
theme_bw()

save(file = "../figures/fig12.Rda", fig12)

```

fig12

Warning: Removed 72 rows containing missing values (geom_path).

