MAGERI benchmark using reference standard DNA library

Mikhail Shugay

December 1, 2016

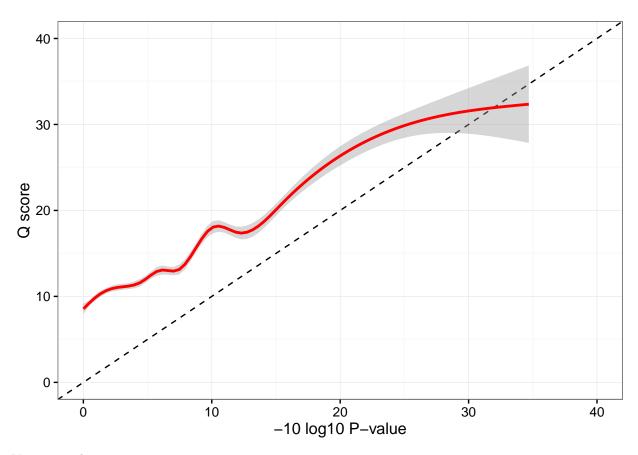
```
Load metadata
df.vmeta <- read.table("hd734_variant_metadata.txt", sep="\t", header=T) # variants observed in HD734 a
df.smeta <- read.table("sample_metadata.txt", sep="\t", header=T) # metadata for amplicon sequencing sa
VCF parsing function
library(stringr)
library(pROC)
## Type 'citation("pROC")' 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
read_vcf <- function(file_name) {</pre>
  .vcf <- read.table(file_name, header = F, sep = "\t", stringsAsFactors = F)</pre>
  colnames(.vcf) <- c("chromosome", "position", "skip1", "from", "to", "qual", "skip2", "info", "skip3"</pre>
  .vcf$skip1 <- NULL</pre>
  .vcf$skip2 <- NULL
  .vcf$skip3 <- NULL
  .vcf$skip4 <- NULL
  .vcf$qual <- as.integer(.vcf$qual)</pre>
  .vcf <- subset(.vcf, nchar(from) == 1 &</pre>
                    nchar(to) == 1 & !is.na(qual)) # no indels
  .infosplit <- str_split_fixed(.vcf$info, regex("[=;]"), 14)[,c(2, 4, 12, 14)]</pre>
  .vcf$coverage <- as.numeric(.infosplit[,1])</pre>
```

```
.vcf$frequency <- as.numeric(.infosplit[,2])</pre>
  .vcf$bb.a <- as.numeric(.infosplit[,3])</pre>
  .vcf$bb.b <- as.numeric(.infosplit[,4])</pre>
  .vcf$info <- NULL</pre>
  .vcf$count <- as.integer(.vcf$coverage * .vcf$frequency)</pre>
  .vcf$qual <- as.integer(.vcf$qual)</pre>
  subset(.vcf, nchar(from) == 1 & nchar(to) == 1 & !is.na(qual))
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
                                                                     bb.a
## 1
           chr2 212295704
                              C A
                                      8
                                             2265 0.0004415011 0.9957058
## 2
                              C A
                                             2265 0.0004415011 0.9957058
           chr2 212295705
                                      8
## 3
           chr2 212295713
                              A G
                                      7
                                             2265 0.0004415011 1.0295689
## 4
           chr2 212295718
                            G A 6
                                             2265 0.0004415011 2.1314828
           chr2 212295725
                             C A 8
                                             2264 0.0004416961 0.9957058
## 5
                             СТ
## 6
           chr2 212295732
                                     24
                                             2264 0.0013250883 2.1314828
##
         bb.b count
## 1 21862.23
## 2 21862.23
                  0
## 3 16394.12
## 4 24344.82
                  Λ
## 5 21862.23
## 6 24344.82
Read samples with HD734 standard DNA and control human DNA, append metadata
library(TailRank) # For betabinom.
## Loading required package: oompaBase
# installing:
# source("https://bioconductor.org/biocLite.R")
# biocLite("Biobase")
# install.packages("TailRank", repos="http://R-Forge.R-project.org")
df <- data.frame()</pre>
read_vcf_with_metadata <- function(file_name, primer_set, replica, ratio, type) {</pre>
  .vcf <- read_vcf(file_name)</pre>
  .vcf <- merge(.vcf, df.vmeta, all.x = type != "standard", all.y = F)</pre>
  .vcf$known.frequency <- .vcf$known.frequency * ratio</pre>
  .vcf$known.frequency[is.na(.vcf$known.frequency)] <- 0</pre>
  .vcf$primer_set <- primer_set</pre>
  .vcf$replica <- primer_set</pre>
  .vcf$type <- type</pre>
  .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,</pre>
                             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
# Compute scores
df$qual <- with(df,</pre>
  mapply(function(x,y,a,b) 1 - pbb(x,y,a,b) + 0.5 * dbb(x,y,a,b),
         count, coverage, bb.a, bb.b)
```

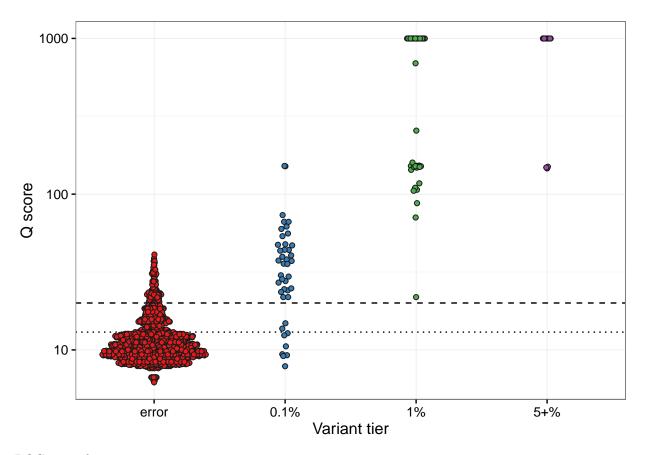
```
df$qual <- ifelse(df$qual <= 1e-100, 999, -10*log10(df$qual))</pre>
## Warning in ifelse(df$qual <= 1e-100, 999, -10 * log10(df$qual)): NaNs
## produced
Group observed variants into tiers
dftier <- cut(dfknown.frequency, c(-1, 0, 0.009, 0.02, 1))
levels(df$tier) <- c("error", "0.1%", "1%", "5+%")</pre>
summary(df$tier)
## error 0.1%
                  1%
                       5+%
## 1471
                  46
            42
                        12
Quality score and error P-values
df.e <- df %>%
 filter(type == "blank") %>%
 mutate(true.p.value = -10 * log10(1 - (rank(frequency) + 0.5) / n()))
## Warning in eval(substitute(expr), envir, enclos): NaNs produced
ggplot(df.e, aes(true.p.value, qual)) +
 geom_abline(slope = 1, intercept = 0, linetype = "dashed") +
 geom smooth(color="red") +
 scale_x_continuous("-10 log10 P-value", limits = c(0, 40)) +
 scale_y_continuous("Q score", limits = c(0, 40)) +
 theme_bw()
```

Warning: Removed 1 rows containing non-finite values (stat_smooth).



Variant quality scores

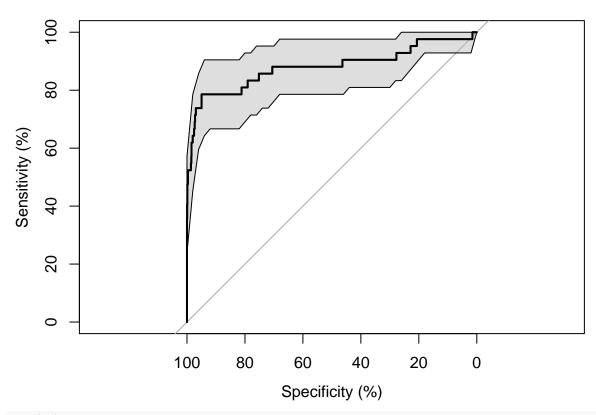
```
ggplot(df,aes(tier, qual, fill=tier)) +
  geom_hline(yintercept = 13.0103, linetype ="dotted") +
  geom_hline(yintercept = 20.0103, linetype ="dashed") +
  geom_quasirandom(varwidth = T, shape=21, color="grey10") +
  scale_fill_brewer("", palette = "Set1", guide=F) +
  xlab("Variant tier") + scale_y_log10("Q score") +
  theme_bw()
```



ROC curve for rare variants

```
df.1 <- subset(df, tier %in% c("error", "0.1%"))
df.1$type <- ifelse(df.1$tier == "error", 0, 1)
rocobj1 <- plot.roc(df.1$type, df.1$qual, percent=T, col="black", ci=T)

sens.ci <- ci.se(rocobj1, specificities=seq(0, 100, 2))
plot(sens.ci, type="shape", col = alpha("grey", 0.5))</pre>
```



rocobj1

```
##
## Call:
## plot.roc.default(x = df.1$type, predictor = df.1$qual, percent = T, col = "black", ci = T)
##
## Data: df.1$qual in 1471 controls (df.1$type 0) < 42 cases (df.1$type 1).
## Area under the curve: 87.97%
## 95% CI: 80.3%-95.65% (DeLong)</pre>
```