Tarea 3

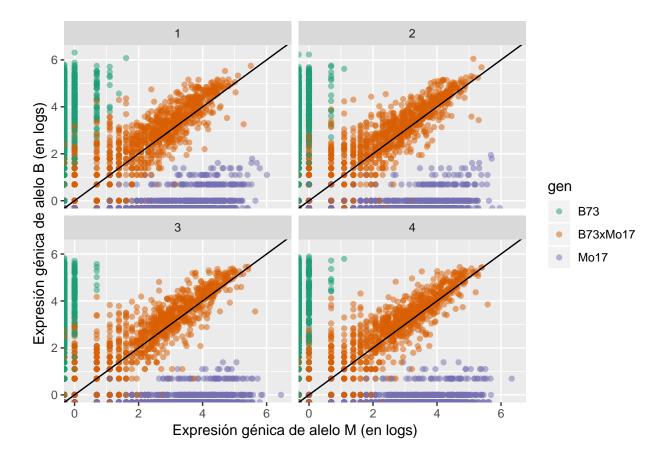
Daniel

16 de noviembre de 2018

Ejercicio 1: Ordenar los datos

a)

```
dd <- read.table('datos_rna.txt', header = T)</pre>
dd.tidy <- NULL
dd.tidy <- dd %>%
  gather(rep, n, -GeneID) %>%
  separate(2, into = c('Rep', 'gen', 'cond')) %>%
  spread(cond,n) %>% mutate(Rep = recode(Rep, 'REP1' = 1, 'REP2' = 2, 'REP3' = 3, 'REP4' = 4) ) %>%
  mutate(gen = recode(gen, B = 'B73', M = 'Mo17', BM = 'B73xMo17', MB= 'Mo17xB73'))
dd.tidy %>% filter(GeneID == 'AC155377.1_FG001')
##
                GeneID Rep
                                gen
                                     b
                                          m total
## 1 AC155377.1_FG001
                        1
                                B73 140
                                            1758
## 2 AC155377.1_FG001
                        1 B73xMo17 106 67
                                             2057
## 3 AC155377.1_FG001
                                             2014
                        1
                               Mo17
                                      2 138
## 4 AC155377.1_FG001
                        1 Mo17xB73 128
                                        99
                                             2521
## 5 AC155377.1 FG001
                        2
                                B73 179
                                          0
                                             2050
## 6 AC155377.1_FG001
                        2 B73xMo17 134 84
                                             2454
## 7 AC155377.1_FG001
                        2
                               Mo17
                                      1 150
                                             2205
## 8 AC155377.1_FG001
                        2 Mo17xB73 119
                                        70
                                             2001
## 9 AC155377.1_FG001
                        3
                                B73
                                    34
                                              173
## 10 AC155377.1_FG001
                                              166
                        3 B73xMo17
                                    22
                                         8
## 11 AC155377.1_FG001
                        3
                               Mo17
                                     1
                                         35
                                              201
## 12 AC155377.1_FG001
                        3 Mo17xB73
                                   19
                                         17
                                              186
## 13 AC155377.1_FG001
                        4
                                B73
                                    30
                                              170
## 14 AC155377.1_FG001
                                              138
                        4 B73xMo17
                                    24
                                         11
## 15 AC155377.1_FG001
                        4
                                     1
                                         38
                                              258
                               Mo17
## 16 AC155377.1_FG001
                        4 Mo17xB73 27 22
                                              213
  b)
dd.tidyg <- dd.tidy %>% filter(gen != 'Mo17xB73')
ggplot(dd.tidyg, aes(x= log(m), y= log(b), colour=gen)) + geom_point(alpha=.5) + facet_wrap(~Rep) + g
```



c) Las cuatro repeticiones muestran que las plantas tienen en su mayoría el aleo correspondiente al gen orginal, y las mixtas parecen tener una proporción similar de cada aleo.

Ejercicio 2: Rcpp y benchmark

a)

```
c <- dd.tidy %>% filter(gen == 'B73xMo17') %>% filter(GeneID == 'AC155377.1_FG001')

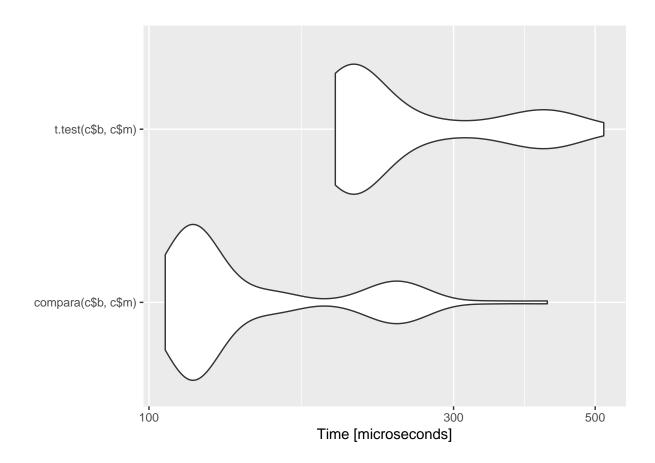
compara <- function(x, y) {
    m <- length(x)
    n <- length(y)
    # calculo el estadistico de la prueba
    sp <- sqrt(((m-1)*sd(x)^2 + (n-1)*sd(y)^2) / (m+n-2))
    tstat <- (mean(x) - mean(y)) / (sp*sqrt(1/m + 1/n))
    # calculo el p-valor
    2*(1 - pt( abs(tstat), df = n+m-2) )
}

compara(c$b,c$m)</pre>
```

[1] 0.433179

```
t.test(c$b, c$m)
##
##
    Welch Two Sample t-test
##
## data: c$b and c$m
## t = 0.83985, df = 5.2765, p-value = 0.4374
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -58.38153 116.38153
## sample estimates:
## mean of x mean of y
        71.5
##
                    42.5
-cppFunction('double comparaC(NumericVector x, NumericVector y) { int m = x.size(); int z = y.size();
int \ sp = sqrt(((m-1)pow(sd(x), \ 2.0) \ + \ (z-1)pow(sd(y), 2.0)) \ / \ (m+z-2)); \ int \ tstat = (mean(x) \ - \ mean(y))
/ (spsqrt(1/m + 1/z)); double n = z+m-2; double a = abs(tstat); double pvalue = 2(1 - pt(a,n)); return
pvalue; }')
comparaC(cb, cm)
  b)
mb <- microbenchmark(</pre>
  compara(c$b,c$m),
  t.test(c$b, c$m)
autoplot(mb)
```

Coordinate system already present. Adding new coordinate system, which will replace the existing one



- c) Compara va ser más eficiente que la función t-test.
- d) La baja cantidad de observaciones.