Probabilities of multilocus genotypes in SIB recombinant inbred lines

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1 Introduction

Set the working directory where the source functions file exists. You can download it from https://github.com/Kamel20/PMGISRIL, and call the functions:

```
setwd("the directory")
source("sibFun.R")
```

Then load the required packages:

```
library(eply)
library(rlist)
library(rmarkdown)
```

2 Input

This code just needs to input the locus number and the recombination rates values. For example for L=3

```
L = 3
recRates = c(0.4, 0.2, 0.3)
```

3 The variables names

To gain time we found the variables that contribute in the system before.

```
allvar = list.load("allVarTillL=10.rds")
SCHPE = list.load("allContrVarTillL=10.rds")
varNom = allvar[[L]]$symQs
nonSymQs = allvar[[L]]$nonsymQs
scEq = SCHPE[[L]]
```

or you can use their function to create it again

```
allVar = systemVar(L)
##
```

Find the variables that are contributing to this system, please wait \dots

```
So, the variables names are

varNom = allVar$symQs

varNom
```

```
## [1] "000" "001" "002" "010" "011" "012" "020" "021" "022" "023" and the nonsymmetric variables are
```

```
nonSymQs = allVar$nonsymQs
 nonSymQs
## [1] "000" "001" "002" "002" "010" "011" "012" "012" "020" "021" "022"
## [12] "023" "020" "021" "023" "022" "011" "010" "012" "012" "001" "000"
## [23] "002" "002" "021" "020" "022" "023" "021" "020" "023" "022" "022"
## [34] "023" "020" "021" "023" "022" "020" "021" "002" "002" "000" "001"
## [45] "012" "012" "010" "011" "022" "023" "021" "020" "023" "022" "021"
## [56] "020" "012" "012" "011" "010" "002" "002" "001" "000"
Note that the first equation of the linear system constructed from here (\sum (Qs) = 1)
SQ = table(nonSymQs)
SQ
## nonSymQs
## 000 001 002 010 011 012 020 021 022 023
   4 4 8 4 4 8 8
This means that
\#\# 4Q(000) + 4Q(001) + 8Q(002) + 4Q(010) + 4Q(011) + 8Q(012) + 8Q(020) + 8Q(021) + 8Q(022) + 8Q(023) = 1
The all possible crossover for each contributed variable is
 scEq = allCrossOver(varNom = varNom)
## 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
 scEq
## [[1]]
## [1] "020" "022" "002" "000"
##
## [[2]]
## [1] "020" "022" "002" "000"
##
## [[3]]
## [1] "021" "023" "001" "002"
## [[4]]
## [1] "020" "022" "002" "000"
##
## [[5]]
## [1] "020" "022" "002" "000"
## [[6]]
## [1] "021" "023" "001" "002"
## [[7]]
## [1] "010" "012" "020" "023"
##
## [[8]]
## [1] "010" "012" "020" "023"
##
## [[9]]
## [1] "012" "011" "021" "022"
##
```

```
## [[10]]
## [1] "012" "011" "021" "022"
```

4 Find the system AQ = B

The system required to compute all the self-consistent equations except one that will replace by the equation of $\sum (Qs) = 1$.

```
res = twoWayRILsib(L, varNom, nonSymQs, scEq)
    # ===== 3 - Loci ===== #
##
##
## Computing the Self-consistent equations: ...
## 1 of 9
2 of 9
3 of 9
4 of 9
5 of 9
6 of 9
7 of 9
8 of 9
9 of 9
## done
Hence, the matrix A is
  A = res$A
  A[1:3,1:2]
##
       000
                                    001
## SQ "4"
                                     "4"
## 000 "2*(0.5)*(1-r12)*(1-r23)-1" "0"
## 001 "2*(0.5)*(1-r12)*(r23)"
and, the matrix B is
  B = res\$B
B[1:3]
## [1] 1 0 0
To solve this linear system you should evaluate this symbolic matrix
  AA = evalMatrix(A = A, recRates = recRates)
## r12 = 0.4
                 r23 = 0.2
                                r34 = 0.3
                                               r13 = 0.44
                                                               r24 = 0.38
                                                                              r14 = 0.476
For example,
  AA[1:3,1:2]
         [,1] [,2]
## [1,]
        4.00
## [2,] -0.52
                 0
## [3,]
        0.12
Hence,
```

```
sol = solve(AA, B)
 names(sol) = varNom
 sol
##
         000
                    001
                               002
                                          010
                                                     011
                                                                012
## 0.03413811 0.01322519 0.01308306 0.01125016 0.02666891 0.01045223
                               022
                                          023
         020
                    021
## 0.01164653 0.01027273 0.02641467 0.01048960
This means,
   Q(0,0,0) = 0.03413811, Q(0,0,1) = 0.01322519, Q(0,0,2) = 0.01308306
## Q(0,1,0) = 0.01125016, Q(0,1,1) = 0.02666891, Q(0,1,2) = 0.01045223
## Q(0,2,0) = 0.01164653, Q(0,2,1) = 0.01027273, Q(0,2,2) = 0.02641467
## Q(0,2,3) = 0.0104896
```

5 Verification by simulation

Note that the sum of all Q's equal to 1

```
QsProbs = rbind(sol, table(nonSymQs))
sum(QsProbs[1,] * QsProbs[2,])
## [1] 1
```

5.1 Convert Qs to Frequencies

We should first convert the Qs to genotypes frequencies:

```
Fexp = QsToFreq(L, sol)

## 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%

Fexp

## [1] 0.17056473 0.09414115 0.08767703 0.14761709 0.14761709 0.08767703

## [7] 0.09414115 0.17056473
```

5.2 Compute the frequencies by simulation

To arrive more stabilty for your results you should choose hight number RIL generation, we choose nRILS = 50000 RIL.

```
nRILS = 1000
```

Then, define the binary hetrzgouse F_2 generation:

```
childGenotype = matrix(c(rep(0, L), rep(1, L), rep(0, L), rep(1, L)), ncol = L, byrow = TRUE)
childGenotype
```

```
##
         [,1] [,2] [,3]
## [1,]
            0
                  0
## [2,]
            1
                  1
                        1
## [3,]
            0
                  0
                        0
## [4,]
            1
                        1
```

Now, run the simulation over nRILS

```
f = rep(0, 2^L)
for (i in 1:nRILS){
    child = Get_One_RIL(L, recRates, childGenotype, type = "sib")
    f[binTodec(child[1,])+1] = f[binTodec(child[1,])+1]+1
    }#EndFor
Fsim = f /nRILS
Fsim
```

```
## [1] 0.171 0.101 0.088 0.133 0.167 0.081 0.102 0.157
```

5.3 Simulation Accuercy

You can compare the analytics results with the simulation one and compute the mean square error (MSE) for that

```
mean((Fexp - Fsim)^2)

## [1] 0.00011588

For more detalis see (Jebreen et al., 2019).
```

Bibliography

Jebreen, K., Marianyela., P., and Martin, O. C. (2019). Probabilities of multilocus genotypes in sib recombinant inbred lines. Statistical Genetics and Methodology, a section of the journal Frontiers in Genetics (under submission).