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(Matrix)
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 ...

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

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 8 4 4 8
                             8
                                 8 8
This means that
The all possible crossover for each contributed variable are
 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] 0.178 0.099 0.080 0.156 0.142 0.075 0.105 0.165
```

5.3 Simulation Accuercy

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

```
mean((Fexp - Fsim)^2)
## [1] 6.865514e-05
```

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

Bibliography

Jebreen, K., Petrizzelli, M., and Martin, O. C. (2019). Probabilities of multilocus genotypes in SIB recombinant inbred lines. (Submitted). Statistical Genetics and Methodology, a section of the journal Frontiers in Genetics.