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 ...
```

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   8   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)"     "-1"
```

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    4
## [2,] -0.52    0
## [3,]  0.12   -1
```

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
##        020        021        022        023
## 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 = 50000
```

Then, define the binary hetrzgouse F_2 genertaion:

```
  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    0
## [2,]    1    1    1
## [3,]    0    0    0
## [4,]    1    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] 0.17256 0.09388 0.08814 0.14710 0.14538 0.08642 0.09740 0.16912
```

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] 2.97788e-06
```

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.*