

imputeR Documentation

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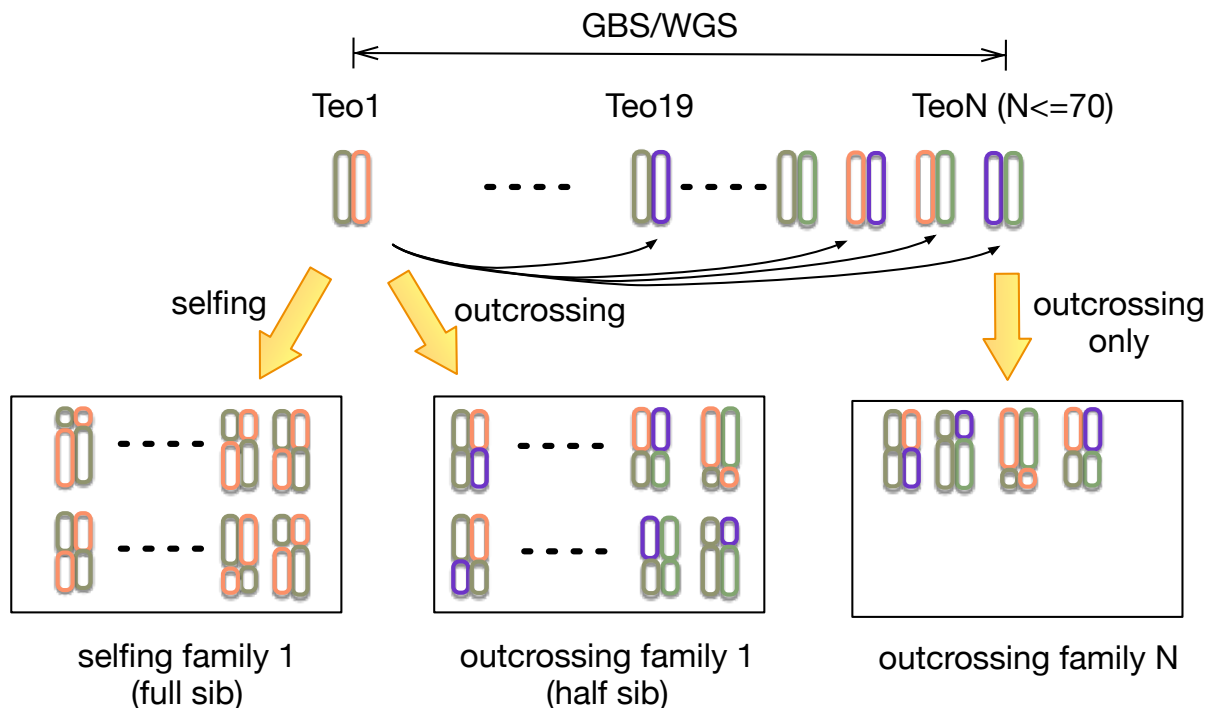
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1 Introduction

1.1 Crossing Scheme and Experimental Design



In this experiment, we selfed and outcrossed a set of ~70 teosinte landraces to get a progeny array composed of 4,875 individuals. The ~70 founders and all the progeny were genotyped using GBS. We also re-sequenced 20/70 founder lines (the others will be re-sequenced soon). Because of the high error rate of the GBS data, especially problematic for calling heterozygous sites, we employed a phasing and imputation strategy to infer the expected genotypes by combining parentage and GBS information.

This file is to document the R package we developed to solve the problems step by step.

1.2 Install and Usage

Install devtools first, and then use devtools to install imputeR from github.

```
# install and load devtools
devtools::install_github("hadley/devtools")
library(devtools)
# install and load imputeR
install_github("yangjl/imputeR")
library(imputeR)
```

1.3 How to find help.

Within "R" console, type `?impute_mom` or `help(impute_mom)` to find help information about the function.

```
?impute_mom
```

```
## No documentation for 'impute_mom' in specified packages and libraries:  
## you could try '??impute_mom'
```

2 Infer mom's genotype from GBS data

If we got mom's WGS data, this step can be skipped.

We have observed mom and observed (selfed) kids. We want to know $P(G|\theta)$, or the probability of mom's genotype given observed data θ . And according to [Bayes' theorem](#),

$$P(G|\theta) \propto P(\theta|G) \times P(G),$$

where $P(G)$ is the probability of the genotype according to the Hardy-Weinberg equilibrium estimated from the population. This consists of observed genotypes (G') of both mom and kids. So:

$$P(G|\theta) \propto \left(\prod_{i=1}^k P(G'_k|G) \right) \times P(G'_{mom}|G) \times P(G),$$

where $P(G'_{mom}|G)$ is the probability of mom's observed genotype given a true genotype G by considering error rates, i.e. GBS homozygote error = 0.02 and heterozygote error = 0.8.

And $P(G'_k|G)$ is the probability of the k th kid's observed genotype given genotype G by considering error rates and Mendelian segregation rate. The function `impute_mom` was implemented to compute mom's genotype probabilities.

2.1 Toy example

In the below toy example, we simulated a full sib family with 12 kids for 3 loci. Mom GBS genotype is 0 0 0 and kids are all 0 0 0. In the resulting table, the first three columns are the probabilities of genotype 0, 1, 2. The 4th column is the odd ratio of the highest divided by the 2nd highest probability. `gmax` mom's genotype with the highest probability. `gor` mom's genotype with the highest probability and OR bigger than your threshold.

```
library(devtools)
library(imputeR)
# mom's GBS data is a vector
obs_mom <- c(0, 0, 0)
# kids GBS data is a list of vectors
obs_kids <- list(c(2, 0, 0), c(0, 0, 0), c(2, 0, 0), c(0, 0, 0), c(2, 0, 0), c(1, 0, 0),
c(0, 0, 0), c(2, 0, 0), c(0, 0, 0), c(1, 1, 0), c(0, 0, 0), c(0, 0, 0))

# run impute_mom to get the probabilities of mom's genotype of all loci
geno <- impute_mom(obs_mom, obs_kids, hom.error=0.02, het.error=0.8, p=NULL)
```

```
## ###>>> impute mom's genotype using [ 12 ] selfed kids ...
## ###>>> no allele frequencies provided. generating random allele frequencies from a neutral SFS
```

```
# find the most likely genotype of mom
momgeno(geno, oddratio=0.5, returnall=TRUE)
```

##	g0	g1	g2	OR	gmax	gor
## 1	-20.10627	-16.30854	-20.16640	3.797734	1	1
## 2	-10.96087	-13.72975	-24.50188	2.768882	0	0
## 3	-10.91072	-13.13179	-23.70376	2.221068	0	0

2.2 Simulated Data

```
set.seed(123456)
sim <- SimSelfer(size.array=10, het.error=0.8, hom.error=0.002, numloci=100, rec=1.2, imiss=0.3)
```

2.3 Real Data

To load hdf5 file, you need to install Vince's [tasselr](#) and [ProgenyArray](#) packages. If you fail to install them, please follow the above links and install the required dependencies.

```
# install devtools and then install the developmental version of tasselr and ProgenyArray using devtools
devtools::install_github("hadley/devtools")
library(devtools)
install_github("vsbuffalo/tasselr")
install_github("vsbuffalo/ProgenyArray")
install_github("yangjl/imputeR")
```

Load the required packages. Note you have to specify the locations of the packages if they were not in your searching path.

```
# load packages
library(parallel)
library(devtools)
options(mc.cores=NULL)
# you need to specify the location where the packages were installed.
load_all("~/bin/tasselr")
load_all("~/bin/ProgenyArray")
load_all("~/Documents/Github/imputeR")
```

The following several lines help you to reformat the *.h5 HDF5 file into an R object. You need to specify the path of the HDF5 file.

```
# Note: at least 100G memory will be needed to load the hdf5 file
# load h5file
teo <- initTasselHDF5("largedata/teo.h5", version="5")
teo <- loadBiallelicGenotypes(teo, verbose = TRUE)
# reformat to imputeR object
source("R/my-TasselHDF5-method.R")
ob <- imputeRob(teo)
save(file="largedata/teo.RData", list="ob")
```

Then, for each mom, run as above in the toy example using only the selfed offspring. Note that unlike the toy example, you will need to supply a vector of allele frequencies at each locus estimated from the parents. If you do not supply this or leave `p=NULL`, a random allele frequency drawn from the neutral SFS will be used instead. Since parents are coded as 0,1, or 2 for N parents the allele frequency p at a locus can be calculated

as $\frac{\sum_{i=1}^N p_i}{2N}$.

3 Phasing Founder Genotypes

$$\begin{aligned} P(H|\theta) &\propto P(\theta|H) \times P(H) \\ P(H|\theta) &\propto \left(\prod_{i=1}^k P(H'_k|H) \right) \times P(H) \\ P(H|\theta) &\propto \left(\prod_{i=1}^k \prod_{l=1}^n P(G'_{i,l}|H) \right) \times P(H) \end{aligned}$$

- Where θ denotes observed data.
 - $P(H)$ is the probability of the haplotype for a given window size of n .
 - $P(G'_{i,l}|H)$ is the probability of kid i at locus l for a given haplotype H .
 - The prior $P(H)$ is that all possible haplotypes of a given window size are equally likely.
-

4 Imputing and Phasing Kids

$$\begin{aligned} P(H_k|\theta) &\propto P(\theta|H_k) \times P(H_k) \\ P(H_k|\theta) &\propto \left(\prod_{i=k} P(H'_k|H_k) \right) \times P(H_k) \\ P(H_k|\theta) &\propto \left(\prod_{i=k} \prod_{l=1}^n P(G'_{i,l}|H_k) \right) \times P(H_k) \end{aligned}$$

- Where θ denotes observed data.
- $P(H)$ is the probability of the haplotype for a given window size of n .
- $P(G'_{i,l}|H)$ is the probability of kid i at locus l for a given haplotype H .
- The prior $P(H)$ is that all possible haplotypes of a given window size are equally likely.

```
phase <- read.csv("../data/sim_phasing_res.csv")

hist(phase$er, breaks=30, main="Simulation (N=100)", col="#faebd7", xlab="Phasing Error Rate")
abline(v=mean(phase$er), col="red", lwd=2)
abline(v=median(phase$er), col="darkblue", lwd=2)
```

5 Phasing Dad of outcrossing progeny array

$$\begin{aligned} P(H_d|\theta) &\propto P(\theta|H_d) \times P(H_d) \\ P(H_d|\theta) &\propto \left(\prod_{i=1}^k P(H'_k|H_d, H_m) \right) \times P(H_m) \times P(H_d) \\ P(H|\theta) &\propto \left(\prod_{i=1}^k \prod_{l=1}^n P(G'_{i,l}|H) \right) \times P(H) \end{aligned}$$

- Where θ denotes observed data.

- $P(H_d)$ is the probability of the dad's haplotype for a given window size of n .
 - $P(G'_{i,l}|H)$ is the probability of kid i at locus l for a given haplotype H .
 - The prior $P(H)$ is that all possible haplotypes of a given window size are equally likely.
-