

# Introduction to GWAS Imputation of Missing Genotypes

## **Christian Werner**

(Quantitative geneticist and biostatistician) EiB, CIMMYT, Texcoco (Mexico)

# Filippo Biscarini



(Biostatistician, bioinformatician and quantitative geneticist) CNR-IBBA, Milan (Italy)

## Oscar González-Recio



(Computer biologist and quantitative geneticist) INIA-UPM, Madrid (Spain)



# Imputation of missing genotypes – why?

#### Imputation - the process of replacing missing data with substituted values

#### Preliminary step for a wide range of genetic analyses

Most models and software for population genetics, genomic selection (GS) and genome-wide association studies (GWAS) do not handle missing data by default and require complete datasets

- 1. Genotyping techniques generate a proportion of missing data (uncalled genotypes)
  - SNP arrays ~5%
  - RAD-Seq (e.g. GBS) ~**50%**
- 2. Optimization/efficiency of genotyping strategies (low → high density data) scaling-up: low → high density (mixed genotyping strategies) whole-genome sequence imputation



# Imputation of missing genotypes – methods

#### 1. General methods for the imputation of any type of data

- mean substitution (replacing missing values with the mean of the SNP across the population), median imputation
- K-Nearest Neighbour Imputation (KNNI)
- many more ...

#### 2. Methods specific for the imputation of missing genotypes

Two groups (and combinations of them):

- based on pedigree information
- based on LD and allele frequency



# Imputation of missing genotypes

## Pedigree imputation uses linkage

- Family statistic
- Correlation between adjacent markers within a family
- Fast and simple, but limitations when inheritance is unclear

## Haplotype library imputation uses LD

- Population statistic
- Correlation between adjacent markers within a population
- Very powerful, but computationally demanding



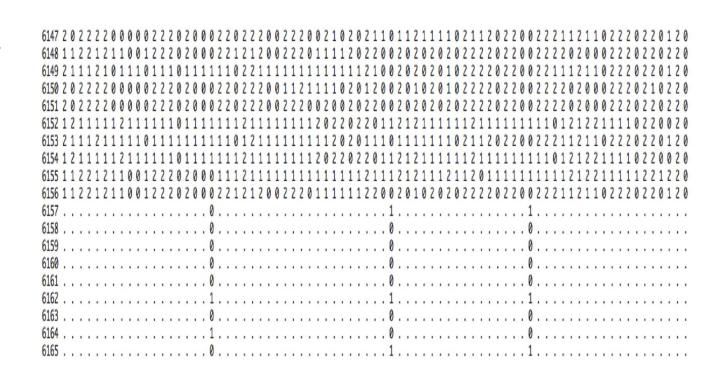
# Imputation of missing genotypes – filling the gaps

#### **Sequenced genotypes**

- full profile
- no missing data

#### **SNP** genotypes

- landmarks along the genome
- empty slots to be filled in





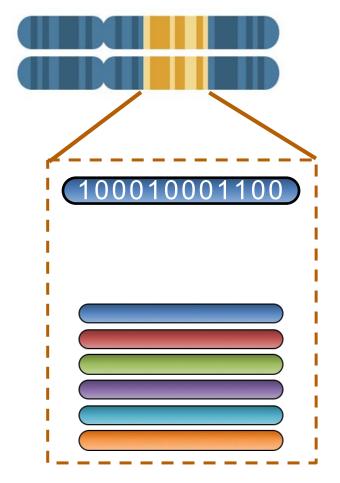


## **Haplotype**

A (section of) a single chromosome with known sequence (phased)

## **Haplotype Library**

A collection of haplotypes







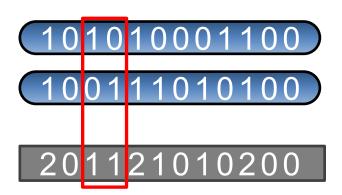
# Allele dosage – prerequisite for imputation and regression

- Diploid genomes (or diploid-like meiotic behaviour)
- A single locus can exhibit four allelic combinations
- Label a=0 and A=1

### Thus the dosage is:

AA = 2 Aa = 1 aA = 1

aa = 0



Maternal chromosome

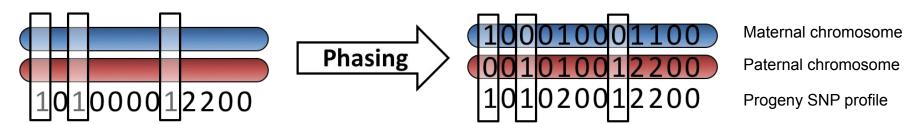
Paternal chromosome

Progeny SNP profile



## Haplotype phasing

- Phasing
  - Determining the haplotype of origin for heterozygotic loci





# Inheritance of genotypes – Pedigree

### **Father**

Chromosome 1 Chromosome 2 **SNP** array

10100111011100111001110011 1111022211111111111111121021

#### Mother

00010011110010101100110011 10111121211121212211221121

Chromosome 1 Chromosome 2 **SNP** array



## Progeny

Paternal chromosome Maternal chromosome **SNP** array

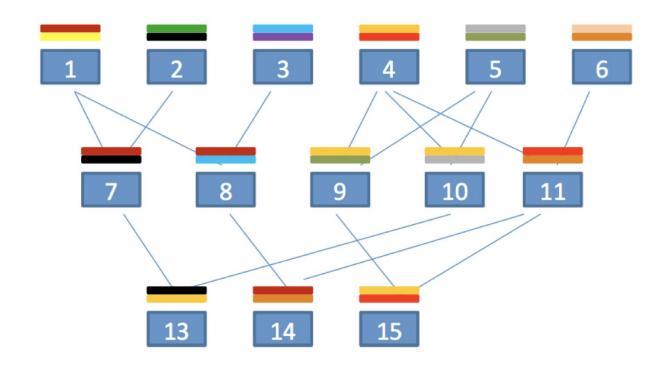
10100111011100111001110011 00010011110010101100110011

1**X**1101221211**XX**212101**X**20022



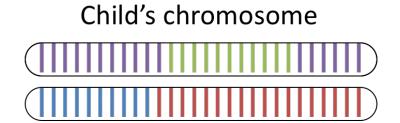


# Inheritance of genotypes – Pedigree

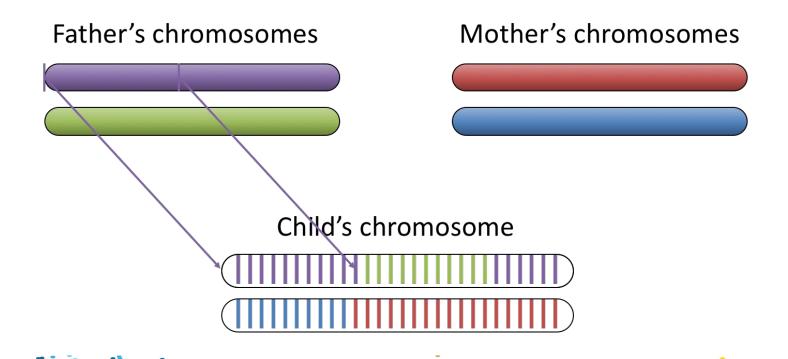




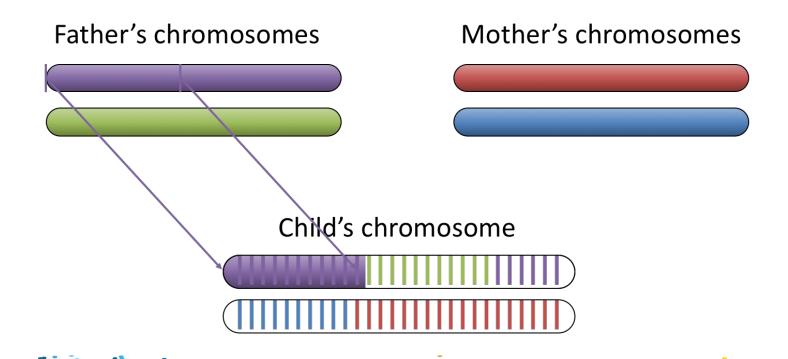




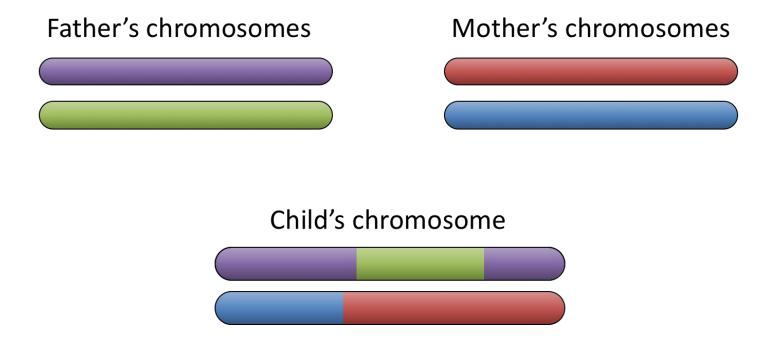








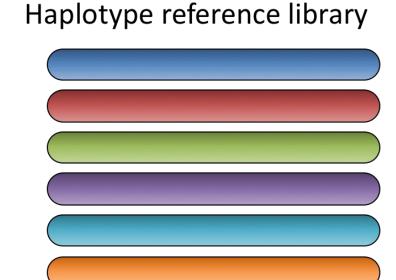






## Imputing from sequenced parents using haplotype libraries

Individual's Chromosome

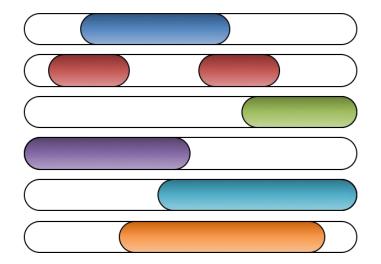




## Imputing from sequenced parents using haplotype libraries

Individual's Chromosome

## Haplotype reference library



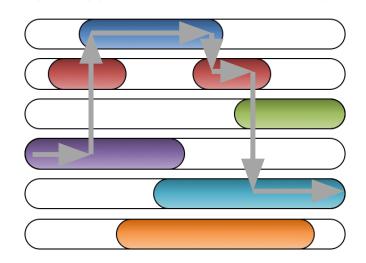


## Imputing from sequenced parents using haplotype libraries

Individual's Chromosome

An individual's haplotype is a mosaic of haplotypes from a reference library.

## Haplotype reference library







## Imputation of missing genotypes - Which approach?

Beagle uses an LD-based approach (Hidden Markov Model; HMM) which in general does a good job using default settings.

- relatively user-friendly
- widely used in the literature
- also does phasing of your data

There are other HMM-based algorithms which show comparable imputation accuracies and computational efficiency. Some of them, however, might not phase your genotypes.

• In this case you need another software to perform phasing before imputation.

A stand-alone pedigree imputation approach should not be used as it is less accurate than algorithms using an HMM.

 However, some algorithms combine pedigree information and an HMM. This might increase accuracy and / or computational efficiency.



# **Imputation with BEAGLE**



## Localised haplotype clustering imputation – LHCI

- **popular method** for the imputation of missing genotypes
- developed originally for humans, has since found wide application also in animals and plants
- makes use solely of genomic information (LD, allele frequency etc.) no pedigree!
- haplotypes are inferred (reconstructed), their frequency estimated, and are clustered "locally"

#### **Detailed introduction how BEAGLE works**

https://www.youtube.com/watch?v=-oUvXXg6tl8



## Localised haplotype clustering imputation – LHCI

- Hidden Markov Model (HMM)
- Find the most likely haplotype pair for each individual given the genotype data for that individual and the haplotype frequency model
- genotypes are then **imputed** based on probabilities from the last fitted model (iterative algorithm)
- **LHCI** is implemented in the software "**BEAGLE**" (Browning and Browning 2007: <a href="https://faculty.washington.edu/browning/beagle">https://faculty.washington.edu/browning/beagle</a>)
- LHCl is the method
- Beagle is the software that implements it



# Genotype imputation – measuring accuracy

Imputation accuracy of all genotype classes (total, AA, AB, BB)

## Why is this important?

- Data are usually unbalanced (major/minor alleles)
- Rare allele (1%) → a naive classifier that always predicts the major allele would be correct 99 times out of 100

99% accuracy overall100% accuracy for the major allelebut 0% accuracy in the minor allele!



# Genotype imputation – measuring accuracy

Imputation accuracy of all genotype classes (total, AA, AB, BB)

## Why is this important?

- Data are usually unbalanced (major/minor alleles)
- Rare allele (1%) → a naive classifier that always predicts the major allele would be correct 99 times out of 100

## Key message

Check the accuracy in the different genotype classes, not the total accuracy