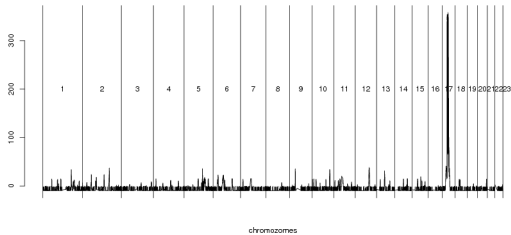


Local IBD inference

Ida Moltke, Naples, May 2017



Outline

1. Introduction

- Goal
- Motivation

2. Inferring local IBD sharing between pairs of individuals

- Current solutions
- An HMM based solution
- Example of use: disease mapping

3. Exercises

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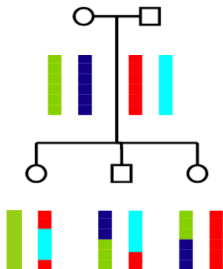
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Goal

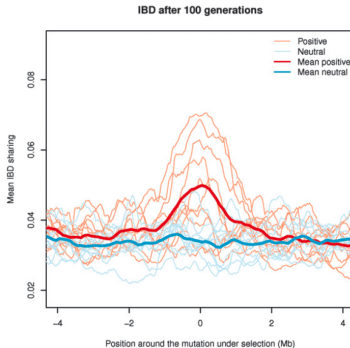
- ▶ We want to infer tracts of IBD sharing along the genome
- ▶ For example:



- ▶ For unphased data: to infer if individuals share 0, 1 or 2 alleles IBD locally

Motivation

- Can be used for population genetic analyses, e.g. detection of selection:



- Can be used for disease mapping (will return to this later if time allows)

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Current solutions

There are several methods for doing this:

- ▶ Non-probabilistic, including but not limited to:
 - ▶ GERMLINE (Gusev et al. 2009)
Finds stretches of identity above a certain (user specified) length
- ▶ Probabilistic, including but not limited to:
 - ▶ BEAGLE (Browning et al. 2010-2016)
 - ▶ Relate (Albrechtsen et al. 2009)

We will look at a fairly simple probabilistic solution
(a simple version of Relate)

Intuition behind

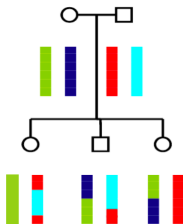
Info source 1: genotypes and allele frequencies (as before!)

- ▶ **If alleles are not identical they cannot be IBD**
- ▶ If alleles are identical it could be due to either IBD or chance
- ▶ If an allele is frequent, identity will occur often simply by chance
- ▶ If an allele is rare, identity will occur rarely by chance
- ▶ **So the rarer the allele, the more identity makes us believe the individuals are IBD**

Intuition behind (cont')

Info source 2: The length of the stretches of identity (new!)

- ▶ IBD is broken up recombination and thus comes in consecutive tracts, hence so will IBD 0, 1 and 2

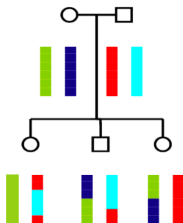


- ▶ Identity that occurs by chance does not have this property.
- ▶ In fact, often very unlikely to see long tracts of identity by chance

Intuition behind (cont')

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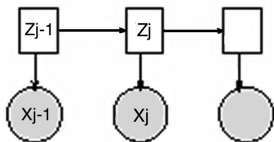
- ▶ IBD is broken up recombination and thus comes in consecutive tracts, hence so will IBD 0, 1 and 2



- ▶ Identity that occurs by chance does not have this property.
- ▶ In fact, often very unlikely to see long tracts of identity by chance
- ▶ The more distantly related, the shorter IBD tracts are expected to be

Using an Hidden Markov Model (HMM)

- ▶ We use these observations for inference based on a HMM
- ▶ Similar to the model we just looked at! E.g. has 2 variables per locus j :
 1. an observed variable, G_j (here genotype data, e.g. (AA,aa)).
 2. a hidden variable, Z_j , that indicates #alleles shared IBD in locus j .
- ▶ What is different: we no longer assume the loci are independent
- ▶ Instead Z_j depends on Z_{j-1}



- ▶ Mathematical formulation for L loci:

$$P(G, Z) = P(Z)P(G|Z) = P(Z_1) \left(\prod_{j=2}^L P(Z_j|Z_{j-1}) \right) \left(\prod_{j=1}^L P(G_j|Z_j) \right)$$

with $G = (G_1, G_2, \dots, G_L)$ and $Z = (Z_1, Z_2, \dots, Z_L)$.

Emission probabilities ($P(G_j|Z_j)$)

- $P(G_j|Z_j)$ is exactly the same as before:

| G_j | $Z_j=0$ | $Z_j=1$ | $Z_j=2$ | |
|-------|----------------|-------------------------|------------|-------------|
| AA,AA | f_A^4 | f_A^3 | p_A^2 | $\forall A$ |
| AA,aa | $2f_A^2 f_a^2$ | 0 | 0 | $A \neq a$ |
| AA,Aa | $4f_A^3 f_a$ | $2f_A^2 f_a$ | 0 | $A \neq a$ |
| Aa,Aa | $2f_A^2 f_a^2$ | $f_A^2 f_a + f_A f_a^2$ | $2f_A f_a$ | $A \neq a$ |

- Captures connection between IBD, genotypes and allele frequencies:
 - that alleles have to be identical to be IBD
 - that the rarer an allele, the more will observing IBS make us believe the individuals are IBD

Transition probabilities ($P(Z_j|Z_{j-1})$)

The probability of a change in IBD state from locus $j-1$ to j :

- ▶ We let these depend on the distance between markers, d , and R
- ▶ E.g. we set

$$P(Z_j = 0|Z_{j-1} = 1) = (1 - e^{-ad})k_0$$

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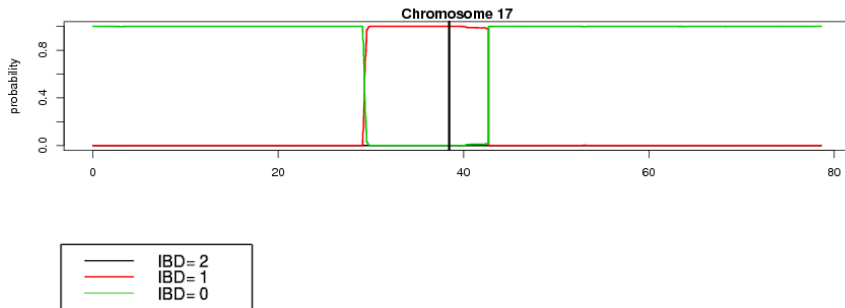
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- ▶ Dependence on the distance d :
 - ▶ The larger d is, the larger is the probability of a change
 - ▶ Reflects that the probability of recombination increases with d
 - ▶ Captures that IBD occurs in consecutive tracts that are broken up by recombination
- ▶ Dependence on R (in this case k_0)
 - ▶ The higher k_0 is (so the more distantly related), the higher is the probability of a change to no IBD sharing ($Z_j = 0$)
 - ▶ Captures that more distantly related tend to have shorter IBD tracts

The inference

- ▶ We can now use standard inference methods for HMMs
- ▶ We can e.g. (using the forward-backward algorithm) get the posterior probability of the 3 possible IBD states $P(Z_j|G)$:

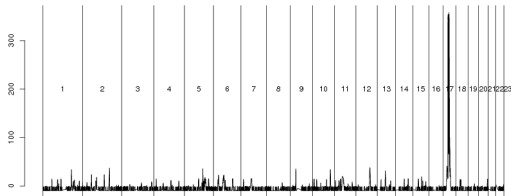


Example of use: disease mapping

- ▶ Individuals with a disease caused by the same mutation are IBD in a region harbouring the mutation
- ▶ Can identify such mutations by finding regions where cases tend to be IBD

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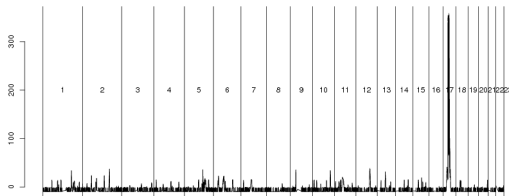
- ▶ Individuals with a disease caused by the same mutation are IBD in a region harbouring the mutation
- ▶ Can identify such mutations by finding regions where cases tend to be IBD
- ▶ E.g. in Albrechtsen et al. 2009:
 - ▶ 7 seemingly unrelated Danish individuals with breast cancer
 - ▶ All genotyped for 225,000 SNPs across the genome
 - ▶ Used an HMM to infer IBD 0,1,2 along the genome
 - ▶ Looked for excess IBD sharing in cases (vs. controls)



- ▶ Only 1 region, where all cases carried same cancer causing mutation

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- ▶ Only 1 region, where all cases carried same cancer causing mutation
- ▶ So mapping can be done w. very few individuals (and SNPs) (\ll GWAS)

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Exercises

Go to <http://popgen.dk/ida/EMBONaples2017/web/> and solve exercise 3 & 4
(run the exercises on the server logged in with `ssh -Y`)