

Haplotypes, linkage disequilibrium, and the HapMap

Jeffrey Barrett

Boulder, 2009

Outline

- 1 Haplotypes
- 2 Linkage disequilibrium
- 3 HapMap
- 4 Tag SNPs

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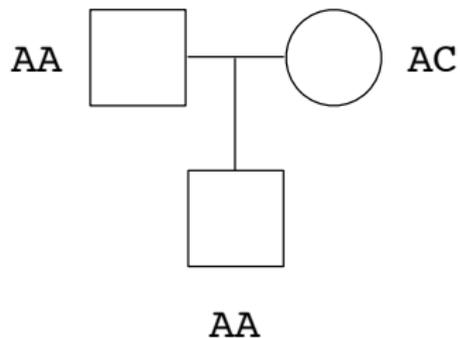
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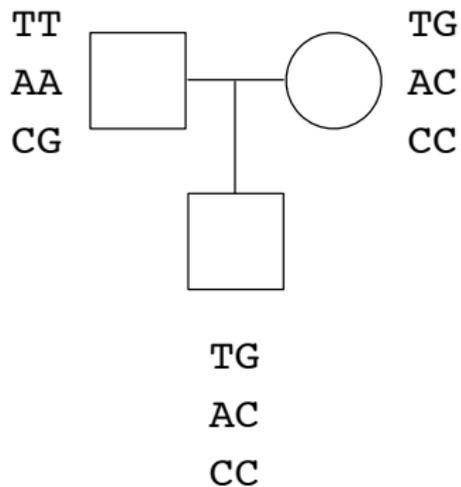
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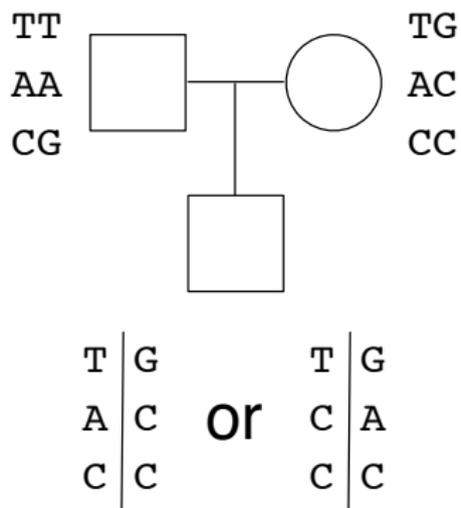
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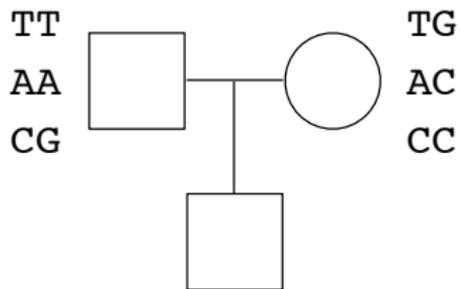
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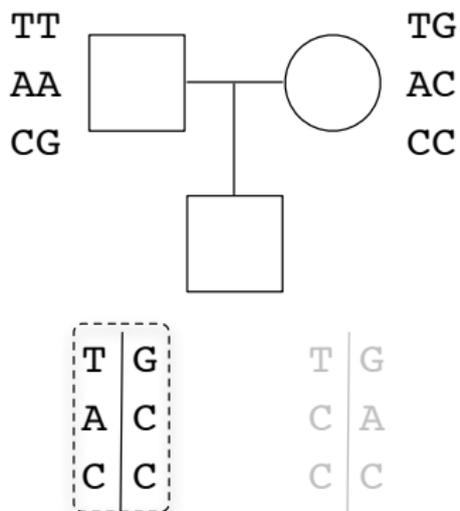
- Using experimental methods. (complicated and expensive)
- Using pedigree information.

How can we resolve phase?



T	G	or	T	G
A	C		C	A
C	C		C	C

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- Using a statistical algorithm applied to data from a population sample.

The EM algorithm

Consider a segment of the genome, with N haplotypes in the population, with frequencies $\theta_1 \dots \theta_N$. If we have a sample of M genotypes, g , we can represent the probability of those genotypes as a sum of the probabilities of each possible pair of haplotypes which could give rise to the observed genotypes:

$$2\theta_j\theta_k \quad j \neq k$$

$$\theta_j^2 \quad j = k$$

The probability of all the genotype data is then:

$$P(g|\theta) = P(g_1)P(g_2) \dots P(g_M)$$

The EM algorithm

Of course, we don't actually know the haplotype frequencies (θ), so we need to use a numerical algorithm to find a maximum likelihood estimate of those estimates, given the genotype data.

We do this in an iterative approach with two steps:

- 1 **Expectation**, where we compute the expected number of haplotypes each individual carries, given the haplotype frequencies.
- 2 **Maximization**, where we re-estimate the haplotype frequencies given the haplotype counts from the 'E' step.

Estimate for θ tends to converge relatively quickly. We can now make guesses about each individual's haplotypes based on these frequencies, but they're not guaranteed to be accurate!

Aspects of EM vs. alternatives

- Fast, and easy to implement.
- Works best when markers in question are relatively strongly correlated (which in practice is only in short genomic segments).
- Better for estimating population frequencies than individual haplotypes.
- Doesn't handle uncertainty in estimates very well.
- More complicated models, including that in the PHASE program, address some of these issues by incorporating knowledge of how haplotypes segregate in populations (the 'coalescent model')

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What creates genetic variation?

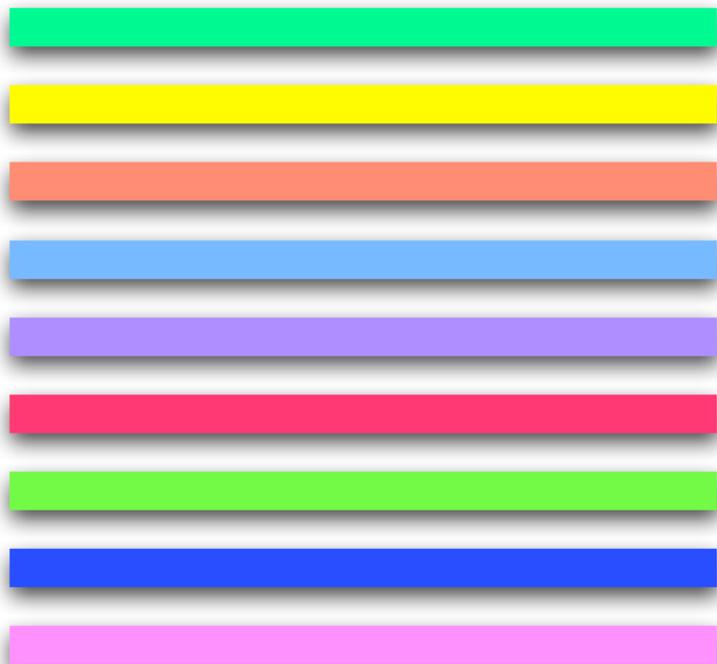
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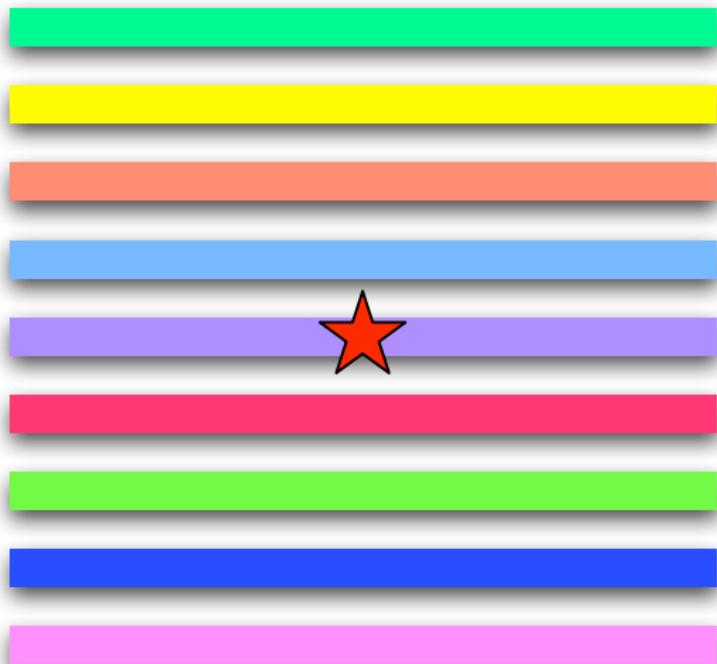
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The fate of new mutations is also affected by drift, selection, and population history. What we really care about is what patterns are left behind in genetic variation because of these forces, and how they affect disease studies.

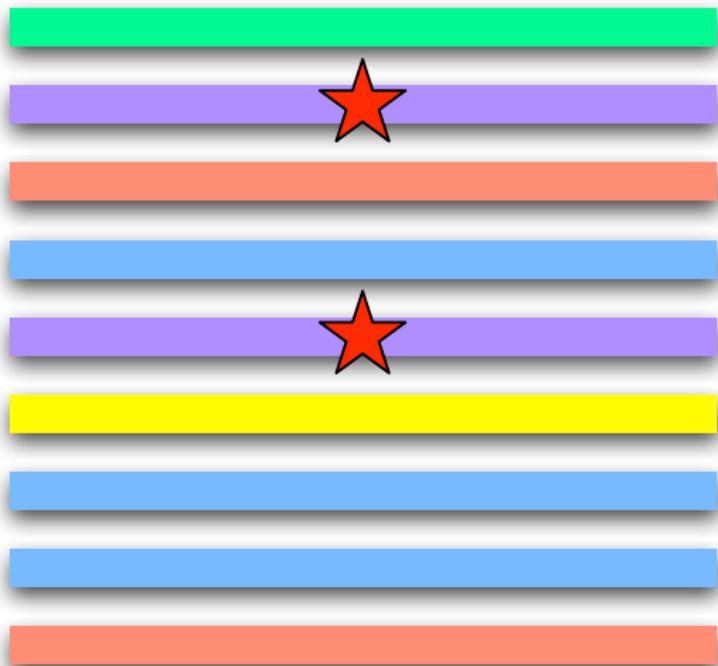
Mutation and recombination in a population



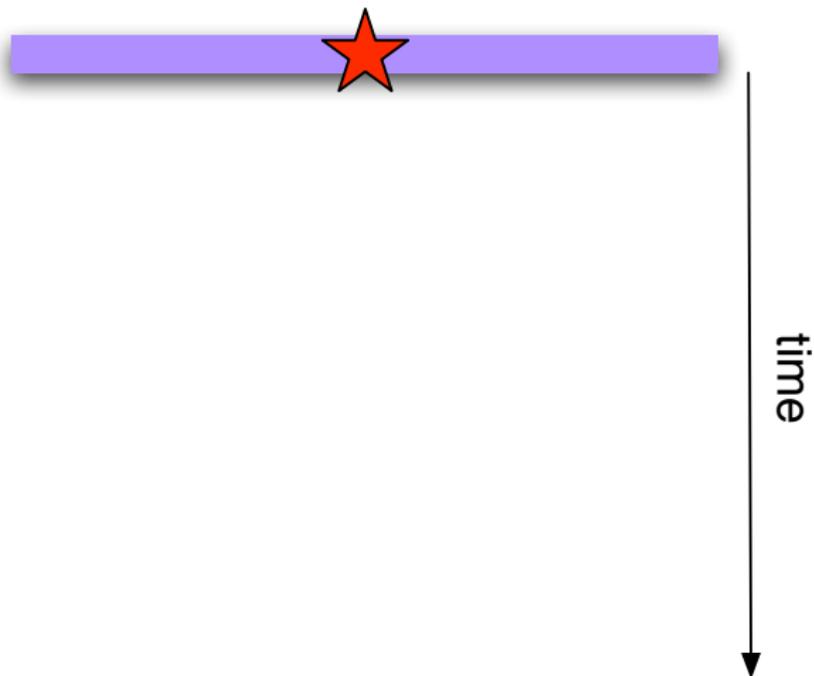
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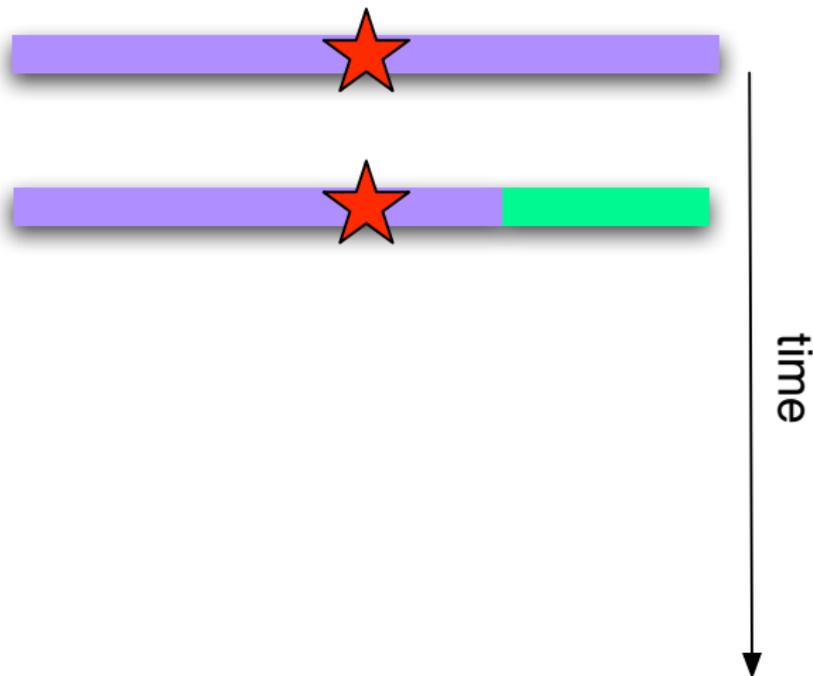
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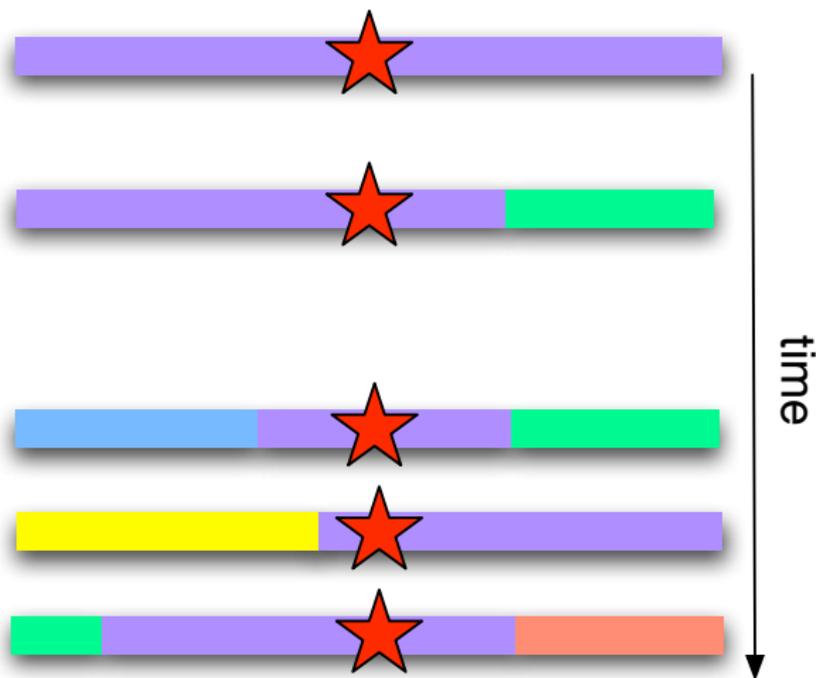
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Consequences of mutation and recombination

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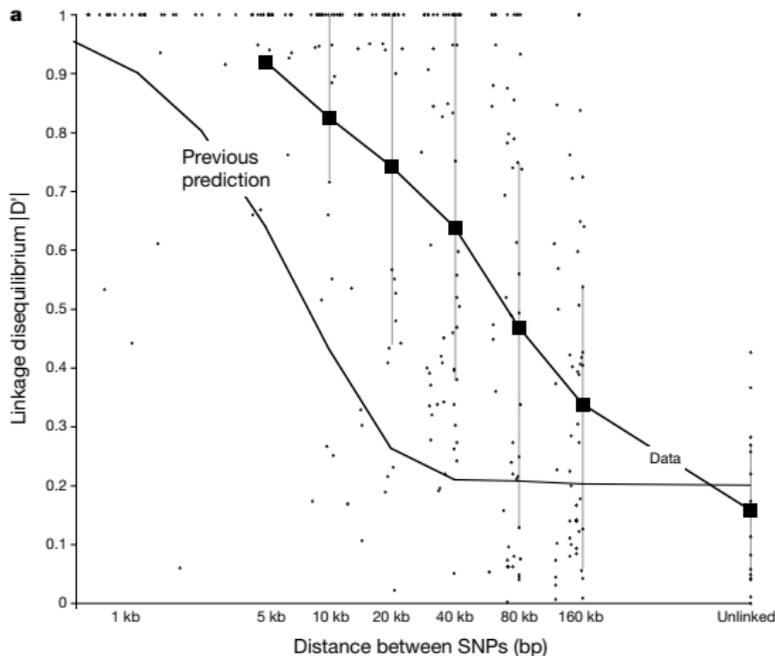
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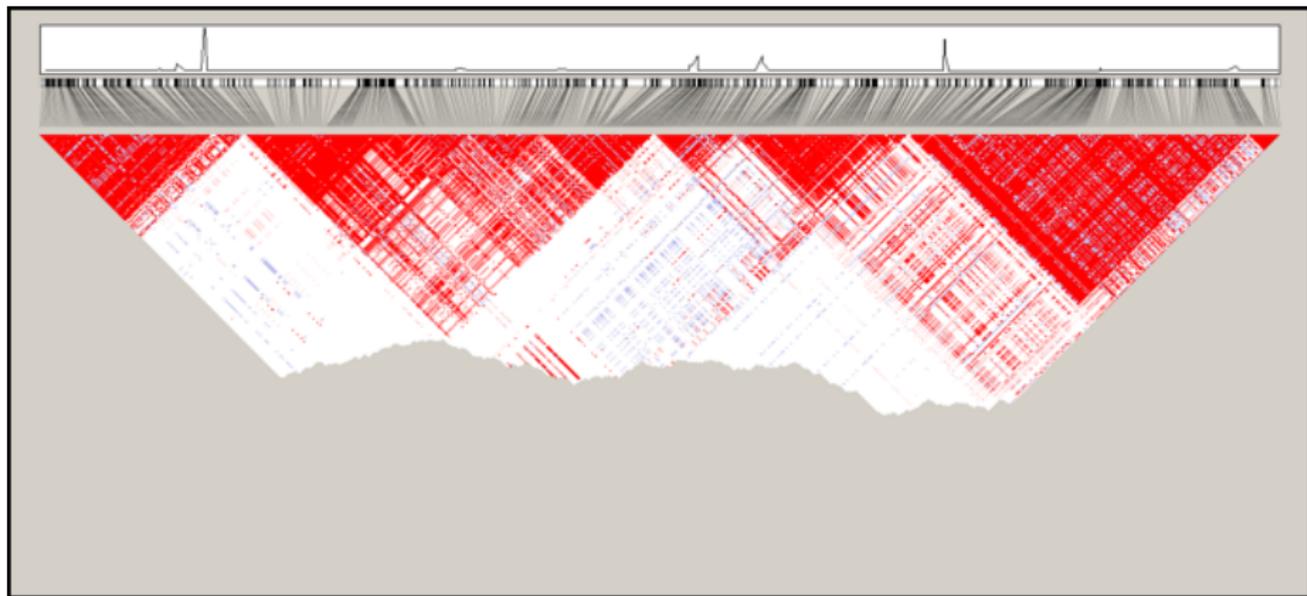
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- Under certain assumptions (neutral evolution, random mating, homogenous recombination), we can model exactly how far this correlation should extend.

Theoretical vs. empirical patterns of LD



Reich et al, *Nature*, 2001.

Heterogeneous recombination drives observed LD patterns



Quantifying LD

		SNP 1	
		p	1-p
SNP 2	q	pq	q(1-p)
	1-q	p(1-q)	(1-p)(1-q)

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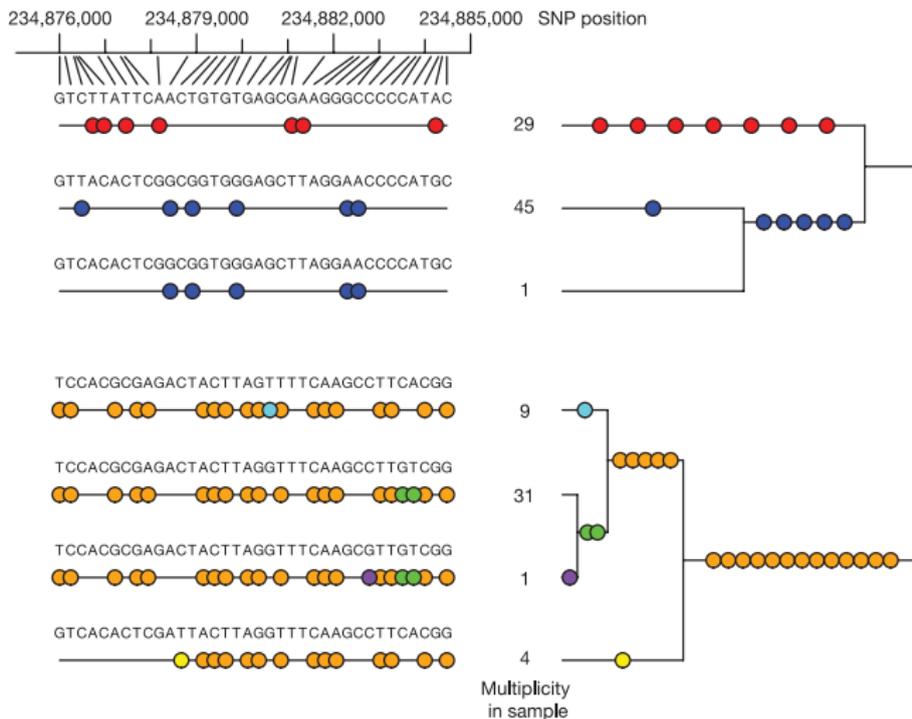
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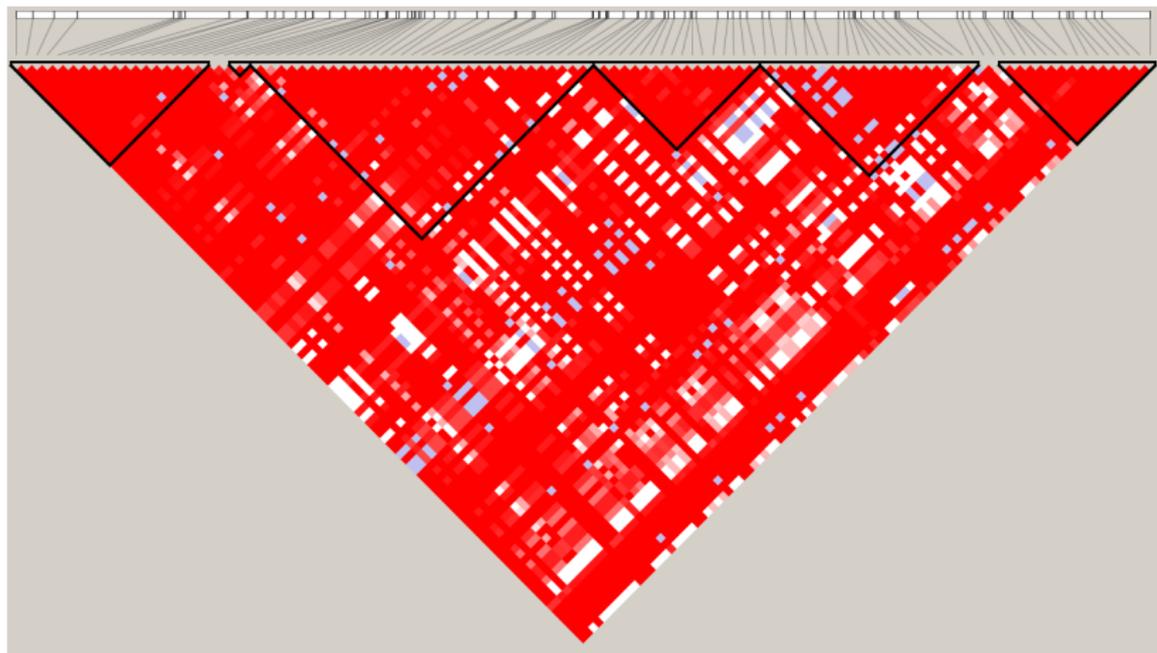
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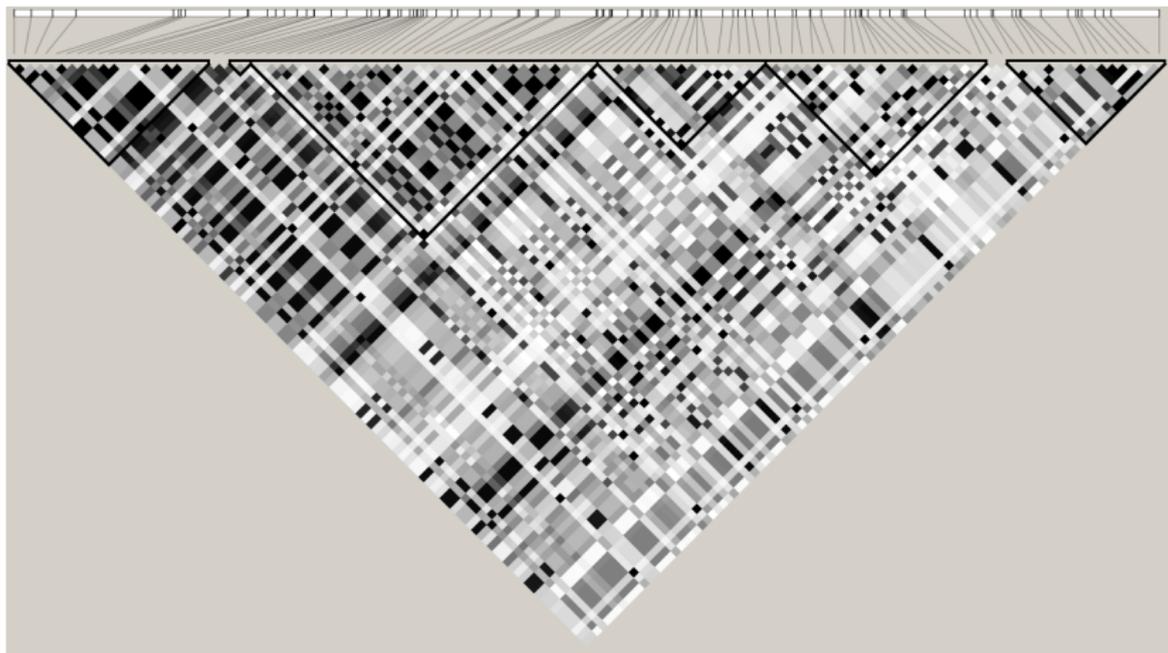
$$D' = D/D_{\max}$$

$$r^2 = D/p(1-p)q(1-q)$$

D' and r^2 

D' in a region of 100kb

D' for common SNPs in a region of 100kb

r^2 for common SNPs in a region of 100kb

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A haplotype map of the human genome



Project details (Phase I/II)

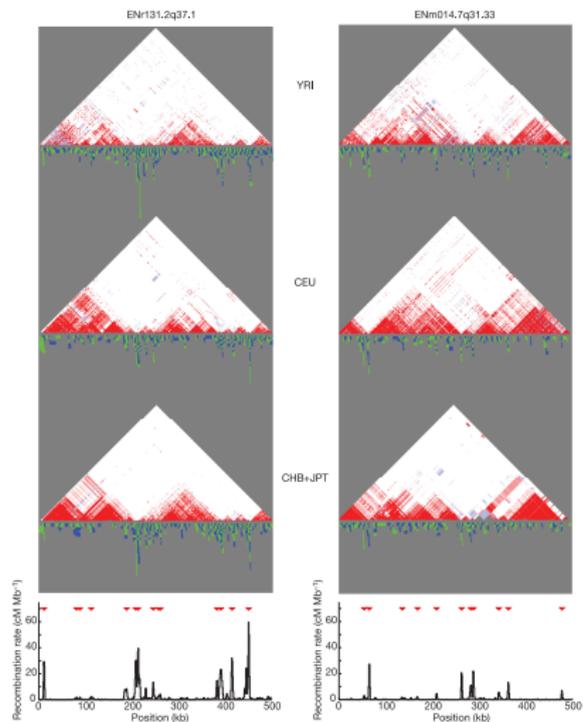
Samples:

- 90 Yoruba (30 parent-parent-offspring trios) from Ibadan, Nigeria (YRI)
- 90 CEPH samples (30 trios) of European descent from Utah (CEU)
- 45 Han Chinese from Beijing (CHB)
- 45 Japanese from Tokyo (JPT)

SNPs: Original goal was 1 SNP every 5kb, but as genotyping costs dropped, eventual catalogue included approximately 4 million polymorphic SNPs scattered across the genome.

Panel	% $r^2 > 0.8$	mean max r^2
YRI	81	0.90
CEU	94	0.97
CHB+JPT	94	0.97

Why multiple populations?



Project details (Phase III)

- African ancestry in Southwest USA (ASW)
- Chinese in Denver, CO (CHD)
- Gujarati Indians in Houston, TX (GIH)
- Luhya in Webuye, Kenya (LWK)
- Mexican ancestry in Los Angeles, CA (MEX)
- Maasai in Kinyawa, Kenya (MKK)
- Tuscans in Italy (TSI)
- (additional samples from CEU, YRI, JPT, CHB)

~ 1.5 million SNPs

Accessing HapMap data with Haploview

Welcome to HaploView

HapMap Download

Release: 21 Chromosome: 11 Analysis Panel: CEU

Start kb: 25975 End kb: 26025

Show HapMap info track

*Phased HapMap downloads require an active internet connection

Ignore pairwise comparisons of markers > 500 kb apart.

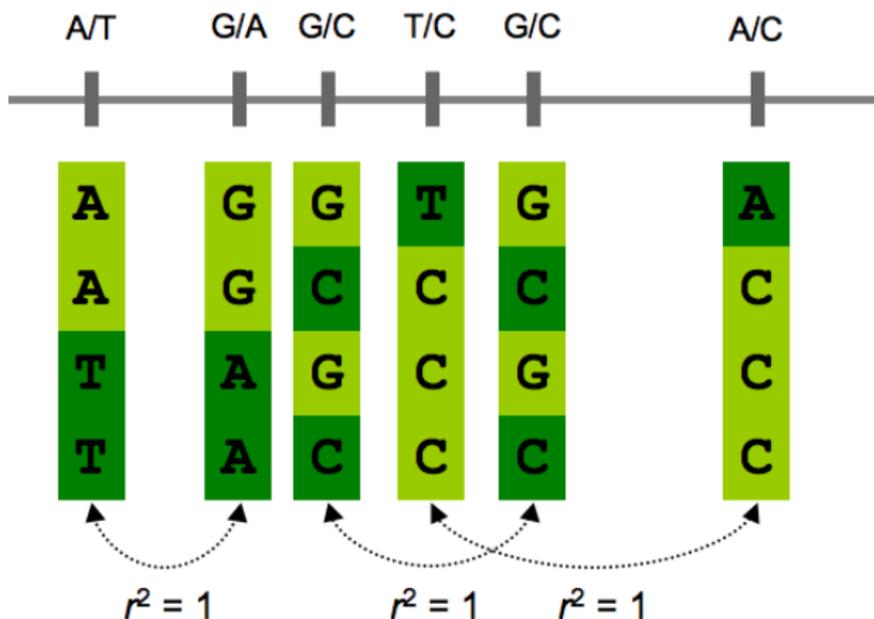
Exclude individuals with > 50 % missing genotypes.

OK Cancel Proxy Settings

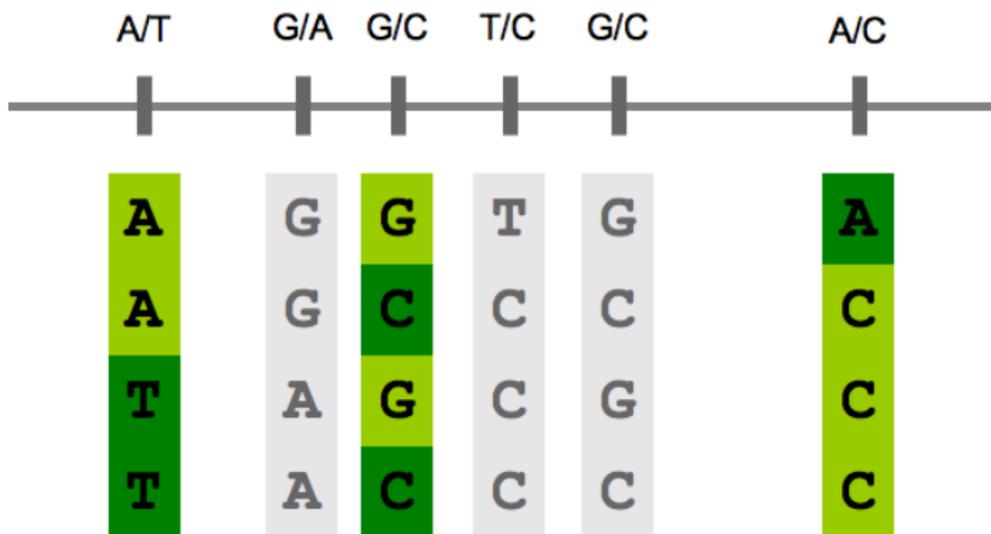
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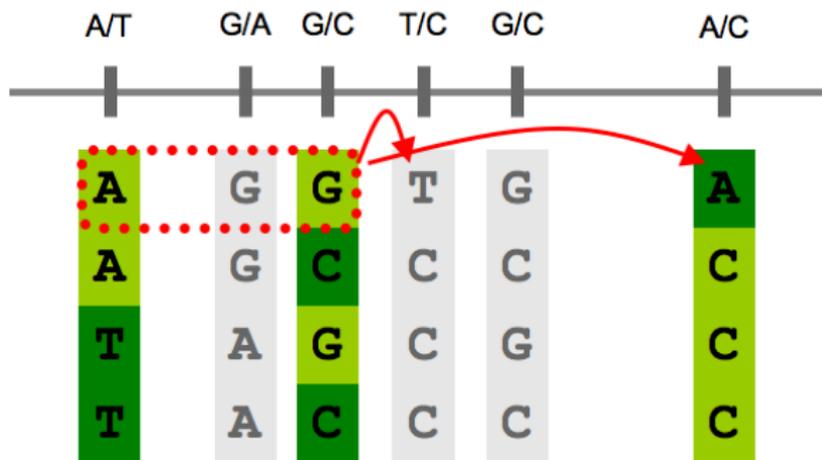
How can we use HapMap knowledge for disease studies?



Gain efficiency by removing redundant SNPs



Haplotypes can yield additional gains in efficiency



No need to genotype this SNP