

### 3. Linkage Disequilibrium (LD)

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# Contents

- 1 Introduction to LD
- 2 LD statistics
- 3 Estimation of LD
- 4 Computer exercises

## LD

- LD: an association between the alleles at different sites in the genome.
- Maybe a consequence of the physical closeness of the sites, but not necessarily so.
- LD is an important concept in disease-marker association studies.

## Linkage Disequilibrium (LD) and Hardy-Weinberg equilibrium (HWE)

- Both concepts refer to association between alleles
- HWE refers to association between alleles at the same locus (within one marker)
- LD refers to association between alleles at different loci (between markers)

# Measures of LD

- $D$  (deviation from independence)
- Lewontin's  $D' = \frac{D}{D_{max}}$
- $R^2$
- $\chi^2$  statistic of a contingency table
- p-value in a chi-square test or in an exact test
- ...

# Haplotype

- A **haplotype** is a combination of alleles at adjacent loci on a chromosome that are transmitted together to the next generation.
- In practice, a haplotype often refers to a set of SNPs on a single chromosome that are statistically associated.
- A haplotype map of the human genome has been constructed ([www.hapmap.org](http://www.hapmap.org)).

## LD

- Consider a population of  $n$  individuals
- Consider two sites (two bi-allelic markers)
- One marker with alleles  $A$  and  $a$ , and one marker with alleles  $B$  and  $b$ .
- Allele frequencies  $p_A$ ;  $p_a$ ;  $p_B$  and  $p_b$ .
- Expected probabilities of each haplotype under independence:

		SNP2		
		B	b	
SNP1	A	$p_A p_B$	$p_A p_b$	$p_A$
	a	$p_a p_B$	$p_a p_b$	$p_a$
		$p_B$	$p_b$	1

## LD

Observed probabilities of each haplotype in presence of LD

		SNP2		
		B	b	
SNP1	A	$p_{APB} + D$	$p_{APb} - D$	$p_A$
	a	$p_{aPB} - D$	$p_{apb} + D$	$p_a$
		$p_B$	$p_b$	1

where  $D$  can be:

$$D = p_{AB} - p_A p_B,$$

$$D = -p_{Ab} + p_A p_b,$$

$$D = -p_{aB} + p_a p_B,$$

$$D = p_{ab} - p_a p_b$$

if  $D \approx 0$  : no LD



# How to compute LD?

- $D = p_{AB} - p_A p_B$
- $p_A$  and  $p_B$  can be estimated by the sample allele frequencies  $\hat{p}_A$  and  $\hat{p}_B$ .
- However,  $p_{AB}$  is unknown.
- Because we have data at the genotype level, and  $p_{AB}$  is at the haplotype level.

# The data

Observed genotype data				
		SNP2		
		BB	Bb	bb
SNP1	AA	$n_{AABB}$	$n_{AABb}$	$n_{AAbb}$
	Aa	$n_{AaBB}$	$n_{AaBb}$	$n_{Aabb}$
	aa	$n_{aaBB}$	$n_{aaBb}$	$n_{aabb}$

- This data can be considered a sample from a multinomial distribution with 9 categories, where the probability of each of the 9 categories ultimately depends on the four haplotype probabilities  $p_{AB}$ ;  $p_{Ab}$ ;  $p_{aB}$  and  $p_{ab}$ .
- We will use a maximum likelihood (ML) approach

# ML estimation

$$\theta = (p_{AB}, p_{Ab}, p_{aB}, p_{ab}), \quad x = (n_{AABB}, n_{AABb}, \dots, n_{aabb})$$

$$L(\theta|x) = \frac{n!}{n_{AABB}! \cdot n_{aabb}!} \cdot (p_{AB}^2)^{n_{AABB}} \dots (p_{ab}^2)^{n_{aabb}}$$

$$l(\theta|x) = C + 2n_{AABB} \ln(p_{AB}) + \dots + 2n_{aabb} \ln(p_{ab})$$

- The problem can be reparametrized in terms of  $p_A$ ,  $p_B$  and  $p_{AB}$
- Because  $p_A = p_{AB} + p_{Ab}$ ,  $p_B = p_{AB} + p_{aB}$  and  $p_{AB} = 1 - p_{Ab} - p_{aB} - p_{ab}$
- Setting  $\frac{\partial l}{\partial \theta} = 0$ , no closed form solution can be found.
- We maximize the likelihood by a Newton-Raphson algorithm.
- Alternatively the expectation-maximization (EM) algorithm may be used.

# Example data set

- Data from the FAMuSS (Functional SNPs Associated with Muscle Size and Strength) study (Foulkes, 2009)
- $n = 1397$  individuals and 225 SNPs
- Muscle performance variables

# Computing LD in R

```

> fms <- read.delim("http://www.stat-gen.org/book.e1/data/FMS_data.txt",header=T,sep="\t")
> n <- nrow(fms)
> p <- ncol(fms)
> print(n)
[1] 1397
> print(p)
[1] 347
> attach(fms)
> actn3_r577x[1:10]
[1] CC CT CT CT CC CT TT CT CT CC
Levels: CC CT TT
> actn3_rs540874[1:10]
[1] GG GA GA GA GG GA AA GA GA GG
Levels: AA GA GG
> Actn3Snp1 <- genotype(actn3_r577x,sep="")
> Actn3Snp2 <- genotype(actn3_rs540874,sep="")
> out <- LD(Actn3Snp1,Actn3Snp2)
> class(out)
[1] "LD"
> attributes(out)
$names
[1] "call" "D" "D'" "r" "R^2" "n" "X^2"
[8] "P-value"
$class
[1] "LD"
> out$D
[1] 0.1945726
> out$"D'"
[1] 0.8858385

```

# ML estimation

It.	$l(P_{AB}, P_A, P_B   x)$	$P_{AB}$	$P_A$	$P_B$
0	-1471.8874	0.0100000	0.508276	0.434483
1	-1469.9878	0.0438867	0.503479	0.429587
2	-1460.8970	0.0375485	0.514644	0.441162
3	-1459.0183	0.0297541	0.514183	0.440727
4	-1458.2618	0.0288494	0.508727	0.435198
5	-1458.0022	0.0263196	0.509216	0.435692
6	-1457.9928	0.0257361	0.507443	0.433847
7	-1457.9840	0.0251530	0.509738	0.432716
8	-1457.9716	0.0253836	0.508019	0.434685
9	-1457.9709	0.0257321	0.507963	0.434594
10	-1457.9696	0.0256473	0.508296	0.434473
11	-1457.9696	0.0256113	0.508247	0.434500
12	-1457.9696	0.0256208	0.508278	0.434481
13	-1457.9696	0.0256212	0.508276	0.434483

After convergence:

$$p_{AB} = 0.0256212; \quad p_{Ab} = p_{A \cdot p_{AB}} = 0.4826544; \quad p_{aB} = p_B - p_{AB} = 0.408862;$$

$$p_{ab} = 1 - p_{Ab} - p_{aB} - p_{AB} = 0.08286239$$

$$D = -0.1952159$$

$D'$ 

- $-0.25 \leq D \leq +0.25$
- $D'$  is an attempt to standardize  $D$

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

$$D_{max} = \begin{cases} \min(p_A p_b, p_a p_B) & D > 0 \\ \min(p_A p_B, p_a p_b) & D < 0 \end{cases}$$

- $-1 \leq D' \leq 1$
- $D' \approx 0$ : low LD
- $|D'|$  close to 1 : high LD.

# $R^2$ and $\chi^2$ statistic

- The genotype data can be recoded as indicator data, creating indicators for the carriers of the A and B allele (AA=0, AB=1, BB=2).
- $R^2$  is the squared correlation between these indicators.
- $R^2$  is related to the  $\chi^2$  statistic of a 2x2 contingency table:  
 $R^2 = \chi^2 / (2n)$ .
- The  $\chi^2$  is related to  $D$ :

$$R^2 = \chi^2 / (2n) = \frac{D^2}{p_A p_B p_a p_b}$$



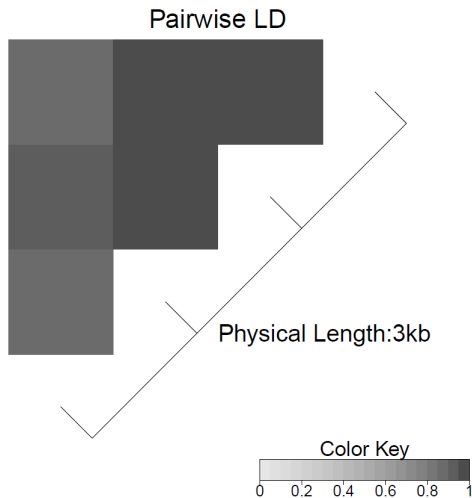
# LD heatmap: graphics for LD with many SNPs

```

> install.packages("LDheatmap")
> library(LDheatmap)
> Actn3Snp1 <- genotype(actn3_r577x, sep="")
> Actn3Snp2 <- genotype(actn3_rs540874, sep="")
> Actn3Snp3 <- genotype(actn3_rs1815739, sep="")
> Actn3Snp4 <- genotype(actn3_1671064, sep="")
> ActnAll <- data.frame(Actn3Snp1, Actn3Snp2, Actn3Snp3, Actn3Snp4)
> LD(ActnAll)$"D'"
      Actn3Snp1 Actn3Snp2 Actn3Snp3 Actn3Snp4
Actn3Snp1      NA 0.8858385 0.9266828 0.8932708
Actn3Snp2      NA      NA 0.9737162 0.9556019
Actn3Snp3      NA      NA      NA 0.9575870
Actn3Snp4      NA      NA      NA      NA
> LDheatmap(ActnAll, LDmeasure="D'")

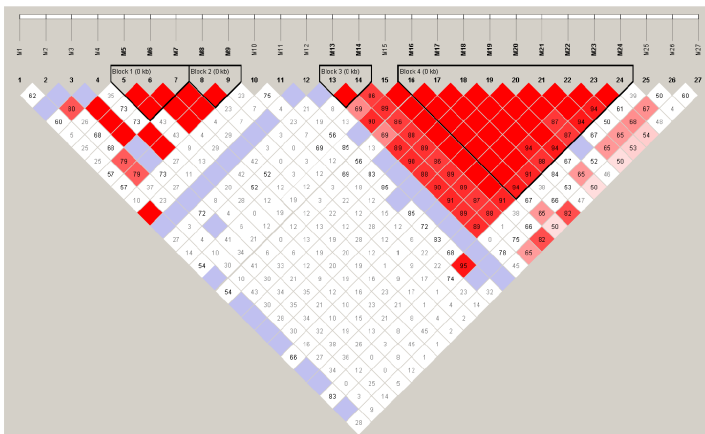
```

# LD heatmap

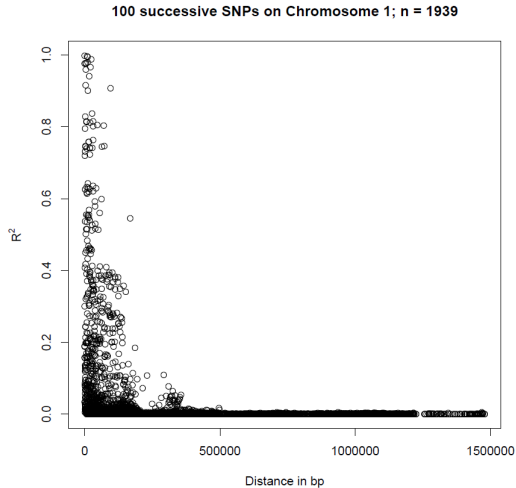


# Another Heatmap (HaploView software)

100 (successive) SNPs on chromosome 1 of a sample of 45 individuals from a Chinese population of the HapMap project ([www.hapmap.org](http://www.hapmap.org)), 27 remaining after removing monomorphics.



# LD and physical distance



# Computer exercises

- ❶ Install the R packages `genetics`, `HardyWeinberg` and `LDheatmap`.
- ❷ Load the database <http://www-eio.upc.es/jan/data/bsg/CHBChr2-2000.rda>
- ❸ Calculate the statistics  $D$ ;  $D'$ ;  $R^2$  and  $\chi^2$  for SNPs 12 and 13. Interpret the results.
- ❹ Repeat the exercise 3 for SNPs 12 and 1000.
- ❺ Select the first 100 SNPs from the database that have complete information (no missings).
- ❻ Compute 4 matrices of association statistics, for  $D$ ;  $D'$ ;  $R^2$  and  $\chi^2$  respectively.
- ❼ Extract the subdiagonal part of each matrix into a vector.
- ❽ Make a scatterplot matrix of the 4 association statistics. Are they related?
- ❾ Make an LDheatmap for each of the four association statistics. Are the results similar?