



THE UNIVERSITY OF  
ALABAMA AT BIRMINGHAM



# Genetic Linkage Analysis

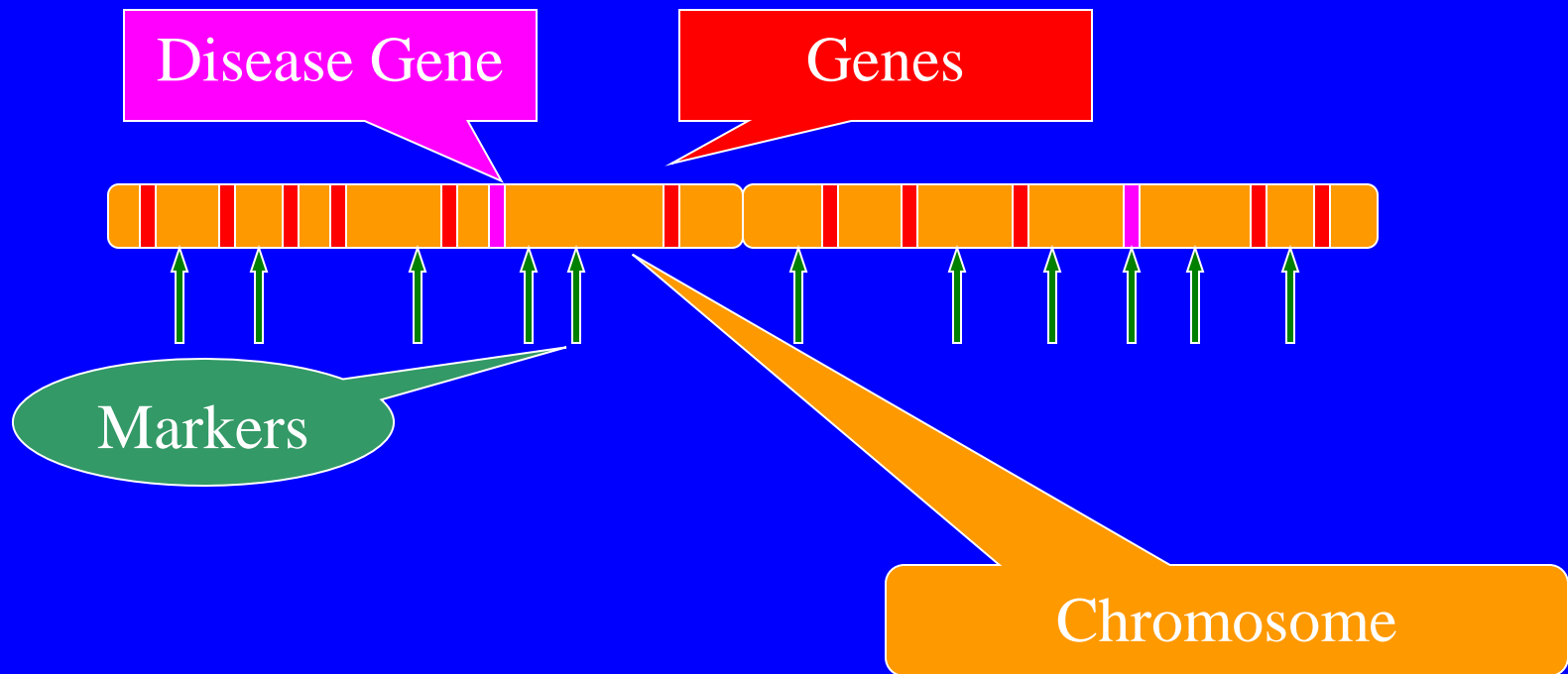
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# How do we find genes implicated in disease or trait manifestation?



# Markers: Microsatellites

- A microsatellite consists of a specific sequence of DNA bases or nucleotides which contains mono, di, tri, or tetra tandem repeats.

For example,

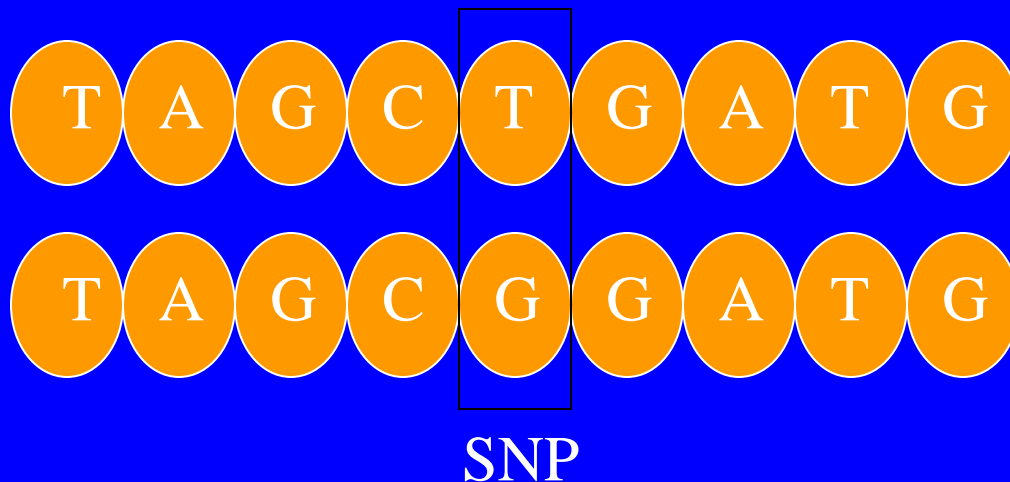
- AAAAAAAAAAAA would be referred to as (A) 11
  - GTGTGTGTGTGT would be referred to as (GT) 6
  - CTGCTGCTGCTG would be referred to as (CTG) 4
  - ACTCACTCACTCACTC would be referred to as (ACTC) 4
- In the literature they can also be called simple sequence repeats (SSR), short tandem repeats (STR), or variable number tandem repeats (VNTR). Alleles at a specific location (locus) can differ in the number of repeats. Microsatellites are inherited in a Mendelian fashion.

# Single Nucleotide Polymorphisms (SNPs)

SNP: DNA sequence variations that occur when a single nucleotide (A, T, C, or G) in the genome sequence is altered.

On average, a SNP exists about every 100-300 base pairs

About 12 millions SNPs on a genome



# HERITABILITY

- Traits are **familial** if members of the same family share them, for whatever reason.
- Traits are genetically **heritable** only if the similarity arises from shared alleles and genotypes.
- To quantify the degree of heritability, one must distinguish between two sources of phenotypic variation:

Hereditary (i.e., Genetic) and environmental

Phenotype = heritable effect + environmental effect

Variance (Phenotype) = Variance (heritable effect)  
+ Variance (environmental effect)

Heritability =  $h^2$  = Proportion of the total variance that is due to hereditary

$$h^2 = \frac{\text{Var}(\text{heritable effect})}{\text{Variance}(\text{Phenotype})} = \text{broad sense heritability}$$

# HERITABILITY IN THE BROAD SENSE

- Limitations of  $h^2$  :
  - 1) not a fixed characteristic of a trait, but depends on the population in which it was measured and the set of environments in which that population grew;
  - 2) if genotype and environment **interact** to produce phenotype, no partition of variation can actually separate causes of variation;
  - 3) high  $h^2$  does not mean that a trait cannot be affected by its environment.

Therefore  $h^2$  has limited meaning and use in humans other than as a parameter to allow for familial correlations.

# Estimation of narrow sense heritability

- see also papers by:
  - Visscher PM, Hill WG, Wray NR. Heritability in the genomics era--concepts and misconceptions. *Nat Rev Genet.* 2008 Apr;9(4):255-66. doi: 10.1038/nrg2322. Epub 2008 Mar 4.
  - Visscher PM, Medland SE, Ferreira MAR, Morley KI, Zhu G, et al. (2006) Assumption-Free Estimation of Heritability from Genome-Wide Identity-by-Descent Sharing between Full Siblings. *PLoS Genet* 2(3): e41. doi:10.1371/journal.pgen.0020041

# Penetrances

- Probability of expressing a phenotype given genotype. Penetrance is either “complete” or “incomplete”. If  $Y$  is a phenotype and  $(A, a)$  is a disease locus, then penetrances are
- $f_2 = P(Y|AA)$ ,  $f_1 = P(Y|Aa)$ ,  $f_0 = P(Y|aa)$
- Complete penetrance:  $(f_2, f_1, f_0) = (1, 1, 0)$  for autosomal dominant disorder
- $(f_2, f_1, f_0) = (1, 0, 0)$  for autosomal recessive disorder

# Hardy-Weinberg Equilibrium

- Random mating and random transmission from each parent => random pairing of alleles

E.g. 2 alleles at one autosomal locus

$$P(A) = p, \quad P(a) = q, \quad p + q = 1$$

$$P(AA) = p^2 \quad P(Aa) = 2pq \quad P(aa) = q^2$$

- **Linkage - Location of genetic loci sufficiently close together on a chromosome that they do not segregate independently**

The proportion of recombinants between the two genes is called the recombination fraction and usually denoted by theta ( $\theta$ ) or r.

Which is same as the probability that an odd number of crossover events will take place between two loci.

Odd number of crossovers: Recombination

Even Number of crossovers: No recombination

# Recombination:



Maternal Chromosome



Paternal Chromosome



NR  $(1-\theta)/2$



NR  $(1-\theta)/2$



R  $\theta/2$



R  $\theta/2$

# Linkage Analysis

- **Linkage Analysis:** Method to map (find the location of) genes that cause or predispose to a disease (or trait) on the human chromosomes and based on the following observation.
- Chromosome segments are transmitted
- Co-segregation is caused by linked loci
- Linkage is a function of distance between loci and recombination event

# Recombination and Genetic Distance

- Probability of a recombination event between two loci depends on distance between them
- If loci are in different chromosome, then recombination fraction ( $\theta$ )=1/2
- Distance can be measured in various ways
- Physical Distance (Base pairs)
- Genetic Distance (Expected number of crossovers) Unit of measurement is Morgan (M)=100cM
- Recombination fraction =P ( odd number of crossovers)
- Since only recombination events are observed, map function are used to convert from Morgans (or cM) to recombination fractions
- 1cM = 1% chance of recombination between two loci.

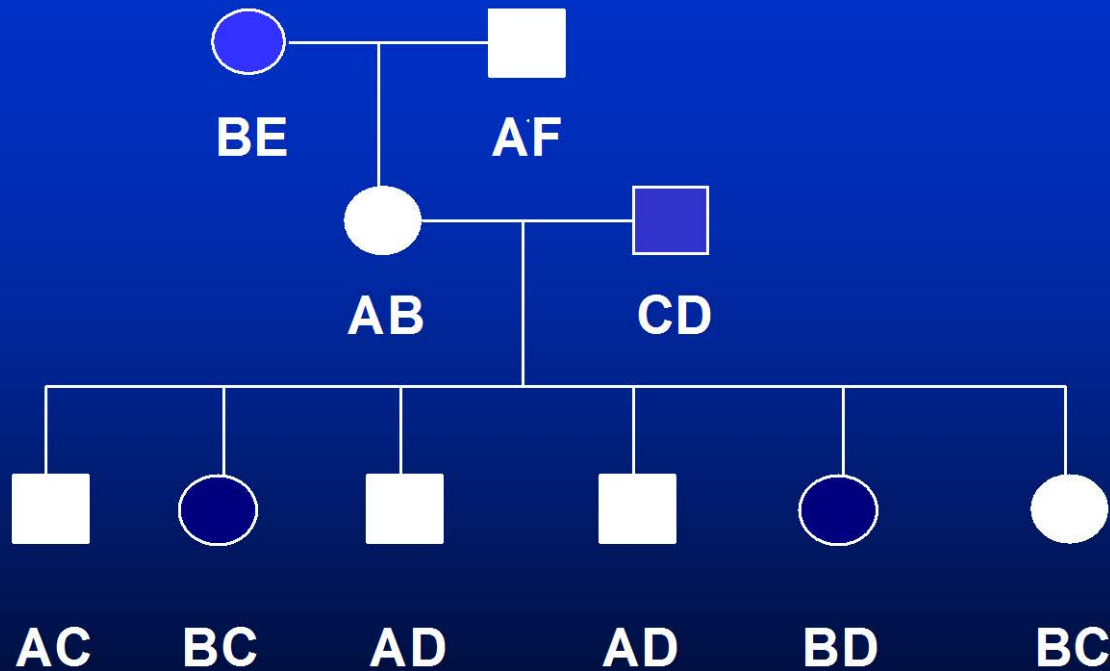
# Methods of Linkage Analysis

- A. Model-based (or parametric): Specify the disease genetic model and estimate  $\theta$  only (Assumes every parameter is known except recombination fraction) .
  
- B. Model-free (non parametric): Do not specify the disease genetic model and estimate functions of  $\theta$  and the genetic model parameters.  
  
Rely on identity-by-descent between sets of relatives.

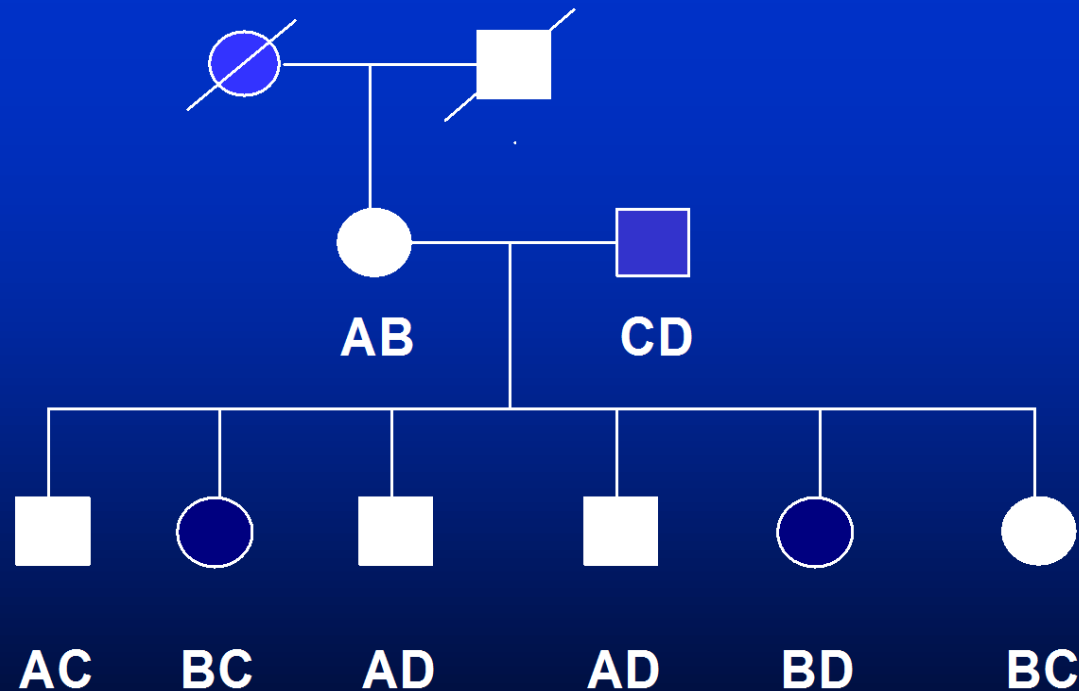
For model-based analysis, we specify

1. Disease allele frequency
2. Penetrances  $\equiv P(\text{Disease} \mid \text{genotype})$
3. Transmission probabilities:  $P(G_j \mid G_{fj}, G_{mj})$

# Phase Known Family



# Phase Unknown Family



# Standard Likelihood or LOD Score Method of Model-Based Linkage Analysis

$H_0 : \theta = 1/2$  (No linkage) vs.  $H_A : \theta < 1/2$

The LOD (“log of the odds”) score is

$$Z = \sum_i Z_i = \sum_i \log \frac{\max L_i(\text{data} \mid \theta = \hat{\theta})}{L_i(\text{data} \mid \theta = 1/2)}$$

where  $Z_i$  is the lod score for the  $i^{\text{th}}$  pedigree

and  $L_i(\bullet \mid \theta)$  is the likelihood of the recombination fraction value  $\theta$  for the  $i^{\text{th}}$  pedigree.

# Interpretation of Lodscore

- Lodscores are additive across independent pedigrees.
- Lodscore greater than or equal to 3 implies Evidence of linkage
- Lodscore less than or equal to -2 implies evidence of no linkage
- Lodscore of zero implies no information
- Positive lodscore for a family suggests support for linkage.

# Calculating Lodscore (Phase unknown family)

Likelihood=  $L(\theta) = P(\text{data}|\theta)$

$$= 1/2 [(\theta)^k (1-\theta)^{n-k} + (\theta)^{n-k} (1-\theta)^k]$$

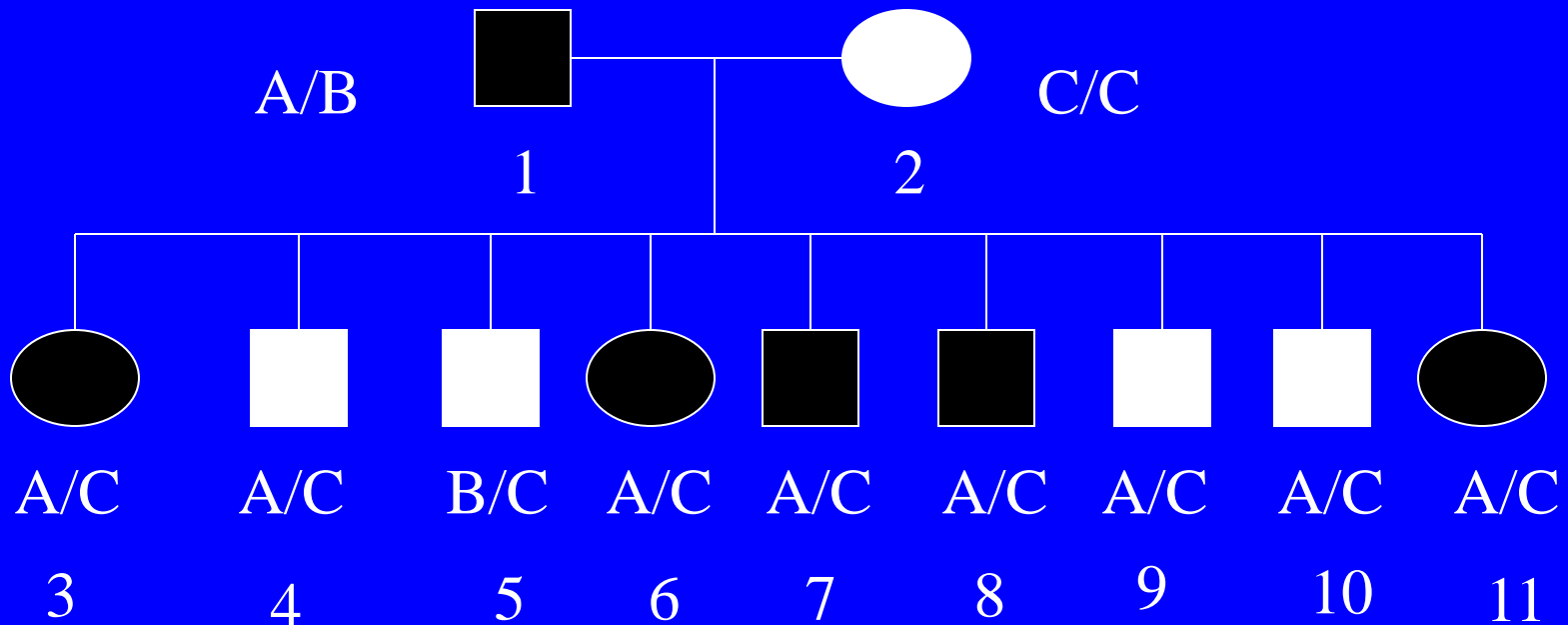
- $k$ : No. of recombinants
- $n$ : All meiosis

# Lodscore (Phase unknown)

$$\begin{aligned} Z = \text{Lodscore} &= \log \frac{\max(L(\text{data} | \theta))}{L(\text{data} | \theta = 1/2)} \\ &= \log \frac{\max 1/2 [(\theta)^k (1-\theta)^{n-k}] + 1/2 [(\theta)^{n-k} (1-\theta)^k]}{(1/2)^n} \\ &= \log \left\{ \max 2^{n-1} \left[ \theta^k (1-\theta)^{n-k} + \theta^{n-k} (1-\theta)^k \right] \right\} \end{aligned}$$

# Example of Phase Unknown family (i.e. grandparents unavailable)

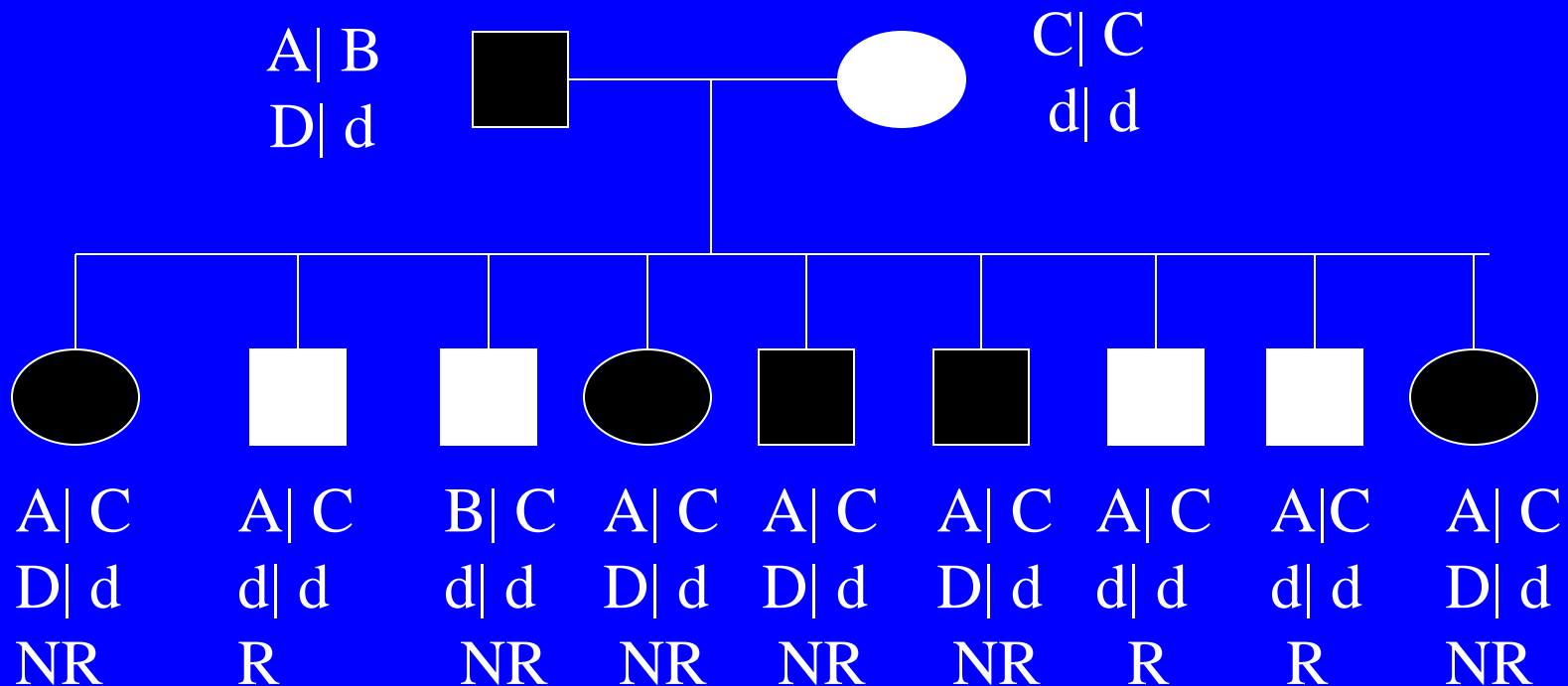
Assume disease is rare autosomal dominant disorder with complete penetrance, i.e. if disease locus has alleles (D, d), the individuals carrying D have disease and people with genotype D/D are rare. Only genotype individuals with D/d are alive. We only have marker alleles.



To calculate number of recombinants we need to know linkage phase whether disease allele D segregates with alleles A or B. There are two possibilities: A allele is paternal or maternal.

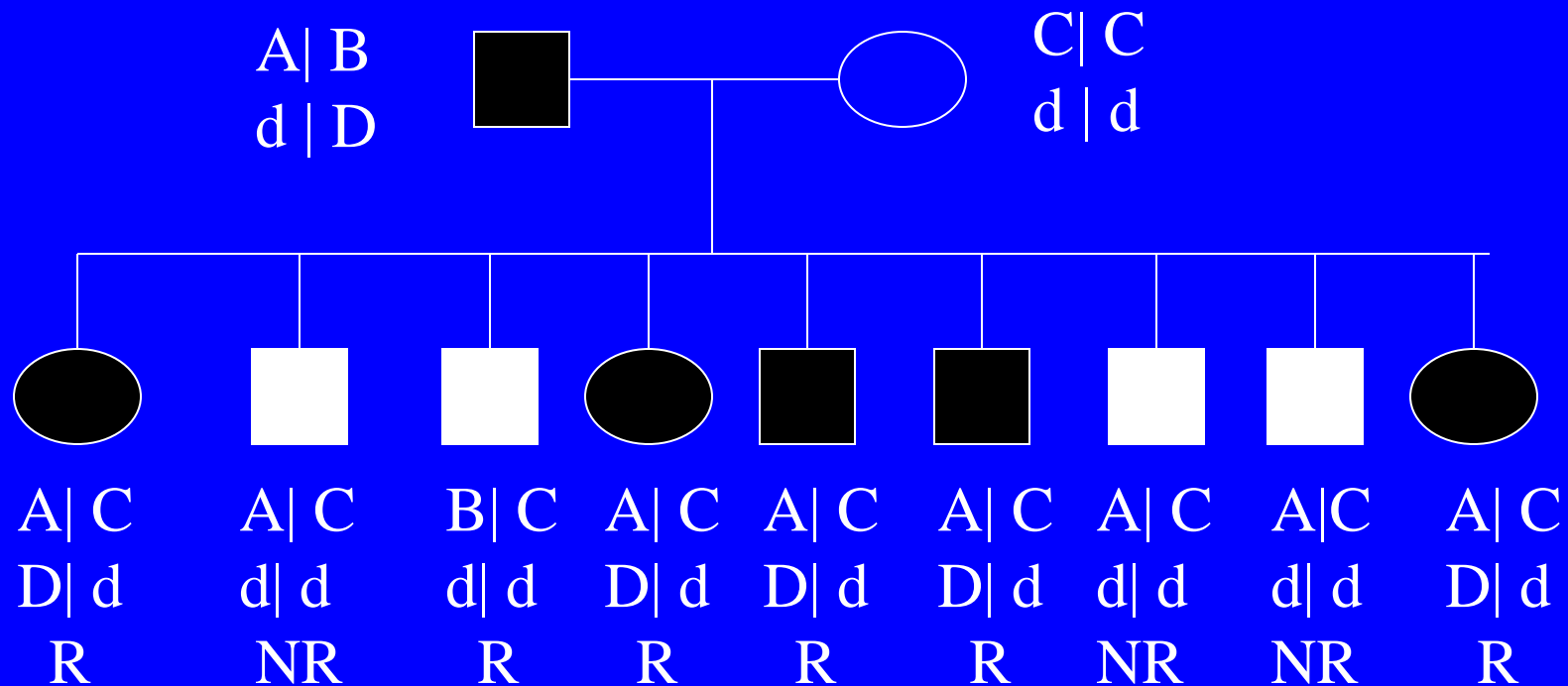
# Phase I (Assume A is linked to disease allele D or disease allele segregates with allele A.)

- Assume Alleles on the left of the vertical line are paternal & on the right of the vertical line are maternal.



- Likelihood =  $L = (\theta)^3(1 - \theta)^6$

# Phase II (Disease Allele D segregates with Allele B)



- $$L = (\theta)^6 (1 - \theta)^3$$

# Lodscore

- Likelihood= P(data| $\theta$ )

$$=L(\text{Phase I})+ L(\text{Phase II})$$

$$=1/2[(\theta)^3(1-\theta)^6]+1/2 [(\theta)^6 (1-\theta)^3]$$

- $$\text{Lod Score} = \log \left[ \frac{\max\left\{\frac{1}{2}[\theta^R(1-\theta)^{NR} + \theta^{NR}(1-\theta)^R]\right\}}{\left(\frac{1}{2}\right)^N}\right]$$

$$= \log(2^{N-1} \max[\theta^R(1-\theta)^{NR} + \theta^{NR}(1-\theta)^R])$$

$$\text{Lod Score} = \log (2^8 \max (\theta^3(1-\theta)^6+ \theta^6 (1-\theta)^3))$$

$\theta$	0.01	0.05	0.1	0.2	0.3	0.4
Lodscore	-3.62	-1.63	-0.87	-0.26	-0.06	0.00

**Conclusion:** The marker locus is unlinked to the disease locus.

# Calculating Lodscore (Phase known family)

- Likelihood=  $L(\theta) = P(\text{data}|\theta)$   
 $= (\theta)^k (1-\theta)^{n-k}$
- $k$ : No. of recombinants
- $n$ : All meioses

# Lodscore (Phase known family)

$$\begin{aligned} Z = \text{Lodscore} &= \log \frac{\max(L(\text{data} \mid \theta))}{L(\text{data} \mid \theta = 1/2)} \\ &= \log \frac{\max((\theta)^k (1 - \theta)^{n-k})}{(1/2)^n} \\ &= \log 2^n \max \theta^k (1 - \theta)^{n-k} \\ &= n \log(2) + \log \max \theta^k (1 - \theta)^{n-k} \end{aligned}$$

# The general Pedigree Likelihood

*Likelihood of the data*

$$\propto \sum_{G_1} \dots \sum_{G_n} \prod_{\text{founder } i} P(G_i) \prod_{\text{nonfounders } j} P(G_j | G_{f_j}, G_{m_j}) \\ \times \prod_{\text{observed } l} P(Y_l | G_l)$$

*$P(G_j | G_{f_j}, G_{m_j})$  is expressed as a function of 2 – locus transmission probabilities.*

# The general Pedigree Likelihood

- Calculating likelihood of large pedigrees is very difficult to calculate by hand.
- Solution: Elston-Stewart Algorithm or Lander-Green Algorithm

# Multipoint Linkage Analysis

- Uses joint information from two or more markers in a chromosomal region
- Uses linkage map rather than physical map
- Each analysis assumes a particular locus order
- Increases power to detect linkage to a disease by increasing the proportion of families with at least one informative marker in a region
- Assumes linkage equilibrium between markers

# Model Based Linkage Analysis

- Statistically, it is more powerful approach than any nonparametric method if model specification is correct.
- Utilizes every family member's phenotypic and genotypic information.
- Provides an estimate of the recombination fraction and lod score.

# Many factors those can influence the lodscore

- Misspecification of disease inheritance model
- Misspecification of marker allele frequency
- Misspecification of penetrance values
- Misspecification of disease allele frequency

# Model-Free Linkage Methods

- Model-free linkage methods do **not** require specification of a genetic model for the trait of interest: that is, they do not require a precise knowledge of the mode of inheritance controlling the disease trait.
- Model-free linkage methods are typically computationally simple and rapid.
- Model-free linkage methods can be used as a first screen of multiple markers to identify promising linkage relationships. Such promising linkage relationships can subsequently be confirmed by consideration of other markers, by standard model-based analysis, by other methods, or a combination of approaches. Alternative approaches rely exclusively on model-free methods, particularly for the analysis of complex disorders, at this level of analysis.

# IDENTITY BY DESCENT

- What are the probabilities  $f_2$ ,  $f_1$ , or  $f_0$  of sharing 2, 1, or 0 alleles, respectively, i.b.d. for different types of relatives?

Assume a large random mating population (no consanguinity):

- For identical twins:  $f_2 = 1, f_1 = 0, f_0 = 0$
- For siblings:  $f_2 = 1/4, f_1 = 1/2, f_0 = 1/4.$
- UNILINEAL RELATIVES - related by only “one line” of genetic descent, i.e., they can have at most one allele i.b.d., implying that  $f_2 = 0$ .

# Method: Haseman Elston (1972)

## Regression of $Y_j$ on $\pi_j$

Let  $\pi_{jt}$  = proportion of alleles shared i.b.d. at the trait locus by the  $j$ -th pair of sibs.

Regression of  $Y_j$  on  $\pi_{jt}$  is  $-2\sigma_a^2$

Regression of  $Y_j$  on  $\pi_{jm}$  is  $-2 \text{Corr}(\pi_{jt}, \pi_{jm}) \sigma_a^2$

$$= -2 [4\theta^2 - 4\theta + 1] \sigma_a^2$$

$$= -2 (1 - 2\theta)^2 \sigma_a^2$$

# Linkage Analysis Using Dense Set of SNPs

- In multipoint linkage analysis we assume that all markers are in linkage equilibrium
- So, markers in LD need to be eliminated to avoid inaccurate calculations that leads to inflation of LOD scores
- Note that linkage analysis requires use of genetic distance and not the physical distance, so need to get genetic map from Rutgers.

## High Resolution Linkage map

- <http://compgen.rutgers.edu/old/maps/index.shtml> (Matisse *et al.* A second-generation combined linkage physical map of the human genome. *Genome Res.* 2007 Dec;17(12):1783-6. Epub 2007 Nov 7.)

# Steps in Linkage Analysis Using Dense Set of SNPs

- Calculate Allele frequency of each marker
- Perform Hardy-Wineberg Equilibrium Test
- Mendelian Test
- Remove markers with LD
- Use appropriate linkage program to find gene locus for your trait

# Selecting SNPs for Linkage Strategy

- Choose a set of ten exclusive subsets of loci from the available GWAS genotype data for each chromosome
- Only consider SNPs with a minor allele frequency (MAF) of at least 0.25
- From the list of eligible SNP based on MAF, we select the first available SNP by position on the chromosome.
- We then include the next available eligible SNP that was at least 1 cM from the previously chosen locus that guarantees linkage equilibrium
- Continue this process until the end of the chromosome
- Subsequently, we remove all of the loci chosen for the first set from the list of available loci, and repeated the process for each subsequent set to obtain ten different complete sets of genome-wide SNPs in linkage equilibrium
- Analyze with appropriate analysis method

# Software

- LINKAGE (Lathrop et al., 1984)
- FASTLINK (Schaffer et al., 1994)
- VITESSE (O'Connell & Weeks, 1995)
- GENEHUNTER (Kruglyak et al., 1996)
- S.A.G.E. (Elston et al., 2004)
- MERLIN (Abecasis, 2000)
- ALEGRO (Gudbjartsson et al., 2000, 2005)
- SOLAR (Almasy et al., 1998)
- SNP HiTLink (Fukuda et al., 2009)