

# 23<sup>rd</sup> International Workshop on Statistical Genetics and Methodology of Twin and Family Studies: Advanced course

- Pak Sham (director)  
- Lindon Eaves  
- Jeff Barrett  
- David Evans  
- Goncalo Abecasis   
- Mike Neale  
- Hermine Maes  
- Sarah Medland  
- Dorret Boomsma 
- Danielle Posthuma 
- Meike Bartels 
- John Hewitt (host)  
- Jeff Lessem 
- Marleen de Moor 
- Stacey Cherny   
- Ben Neale  
- Shaun Purcell  
- Manuel Ferreira   
- Nick Martin 
- Kate Morley 
- Allan McRae 

# Hunting QTLs – an introduction

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Nick Martin

Queensland Institute of Medical Research



Boulder workshop: March 2, 2009

# Mendel 1865 – genetics of discrete traits



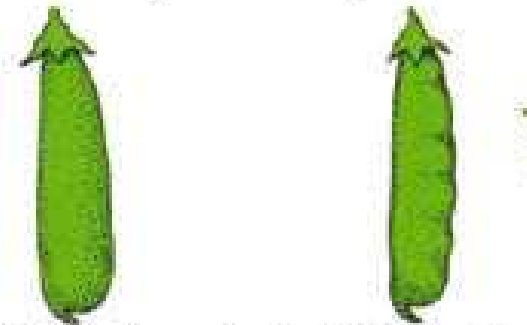
Round or wrinkled ripe seeds



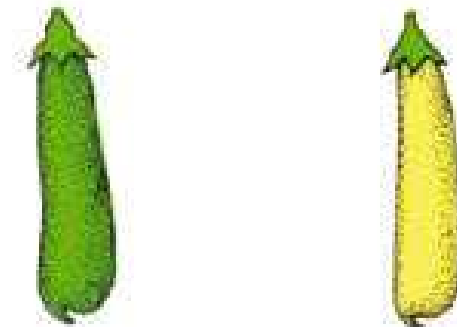
Yellow or green seed interiors



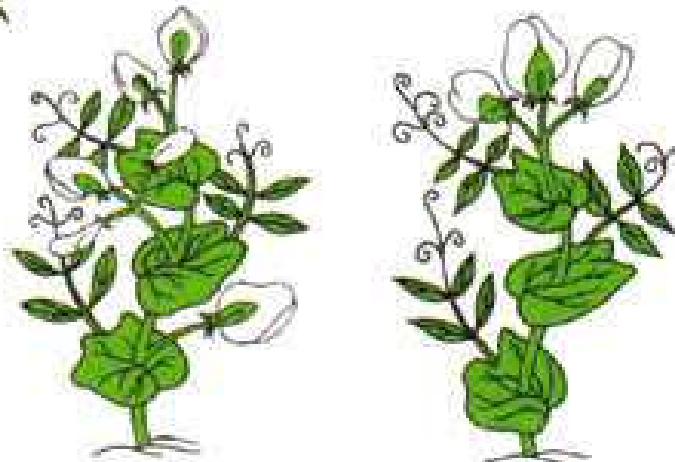
Purple or white petals



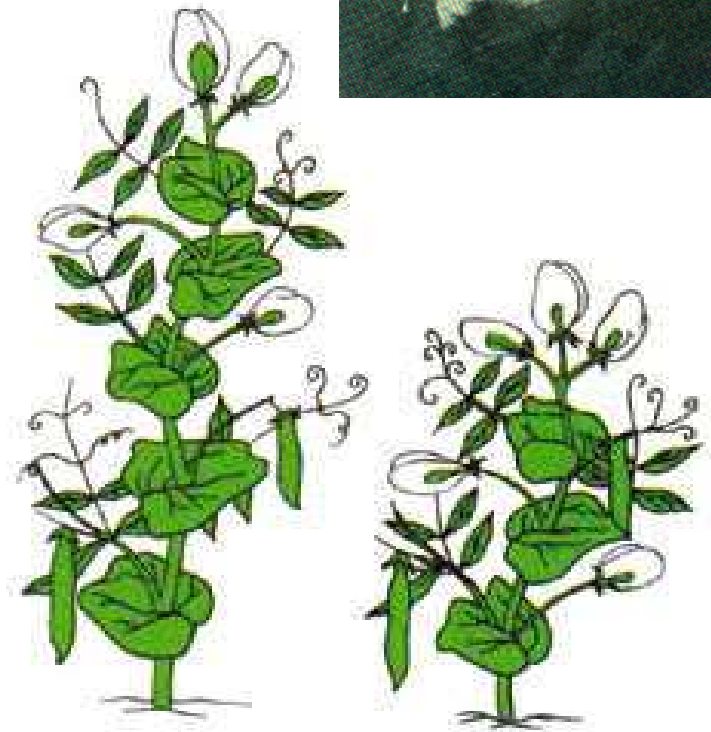
Inflated or pinched ripe pods



Green or yellow unripe pods



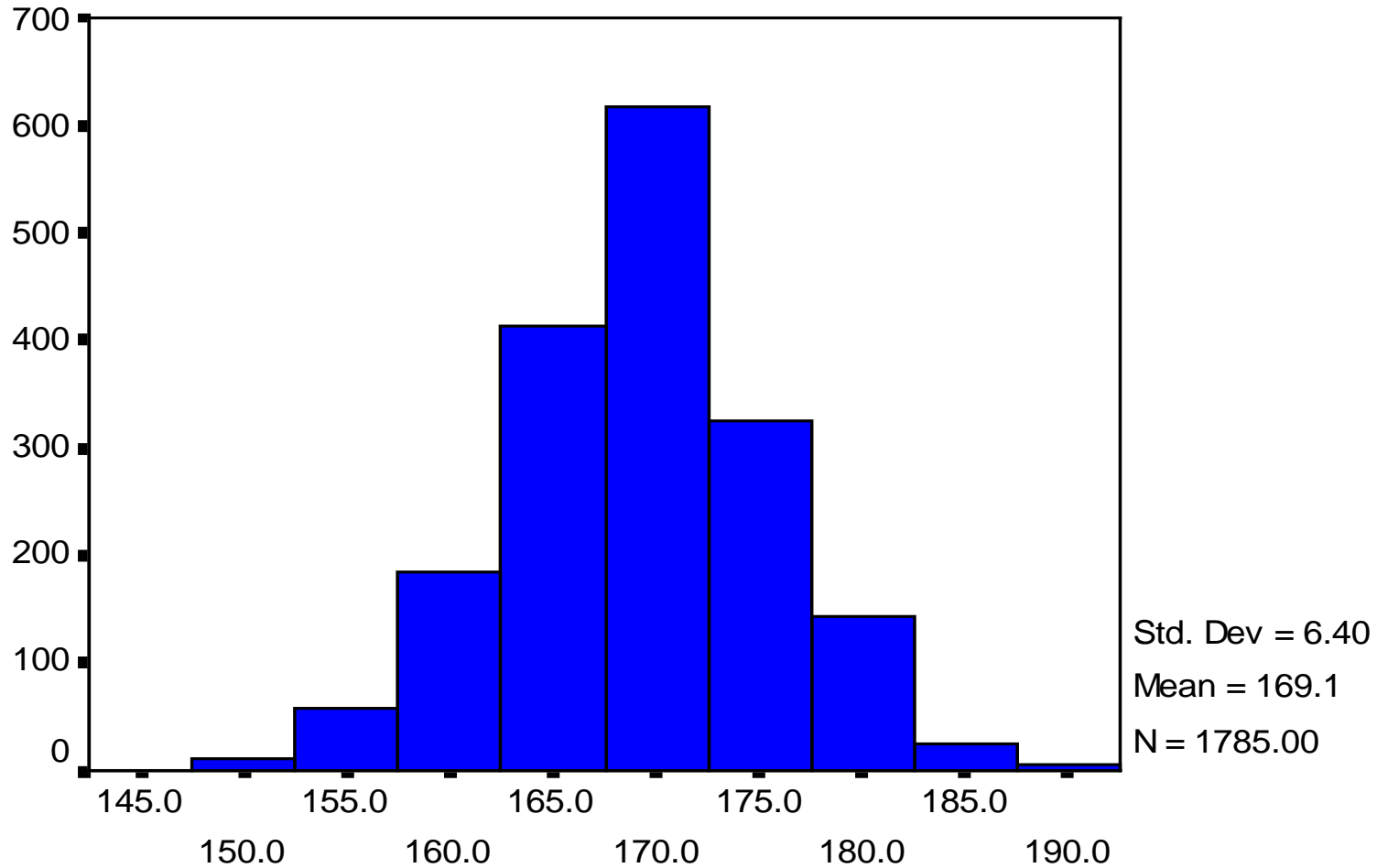
Axial or terminal flowers



Long or short stems

# Stature in adolescent twins

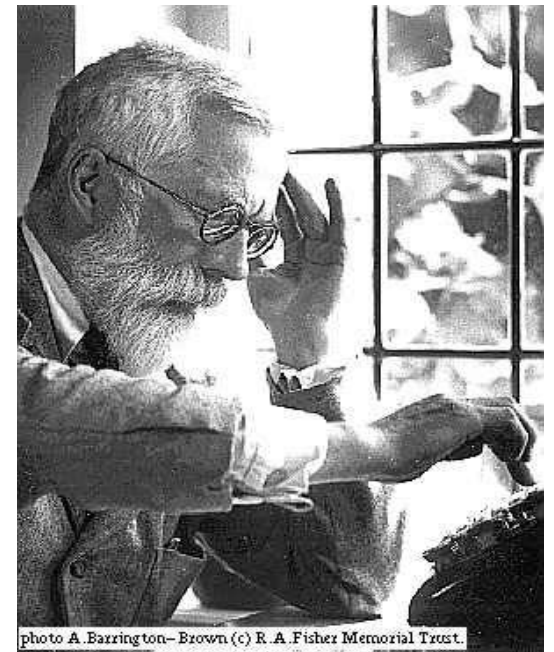
Women



Stature

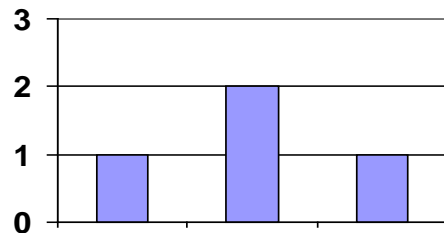
# R.A. Fisher, 1918

## The explanation of quantitative inheritance in Mendelian terms



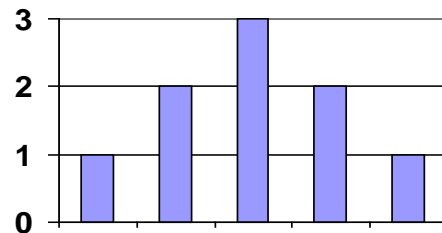
1 Gene

→ 3 Genotypes  
→ 3 Phenotypes



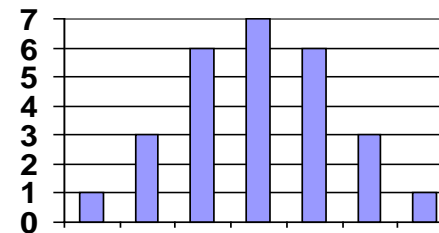
2 Genes

→ 9 Genotypes  
→ 5 Phenotypes



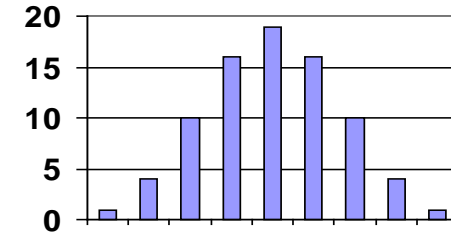
3 Genes

→ 27 Genotypes  
→ 7 Phenotypes



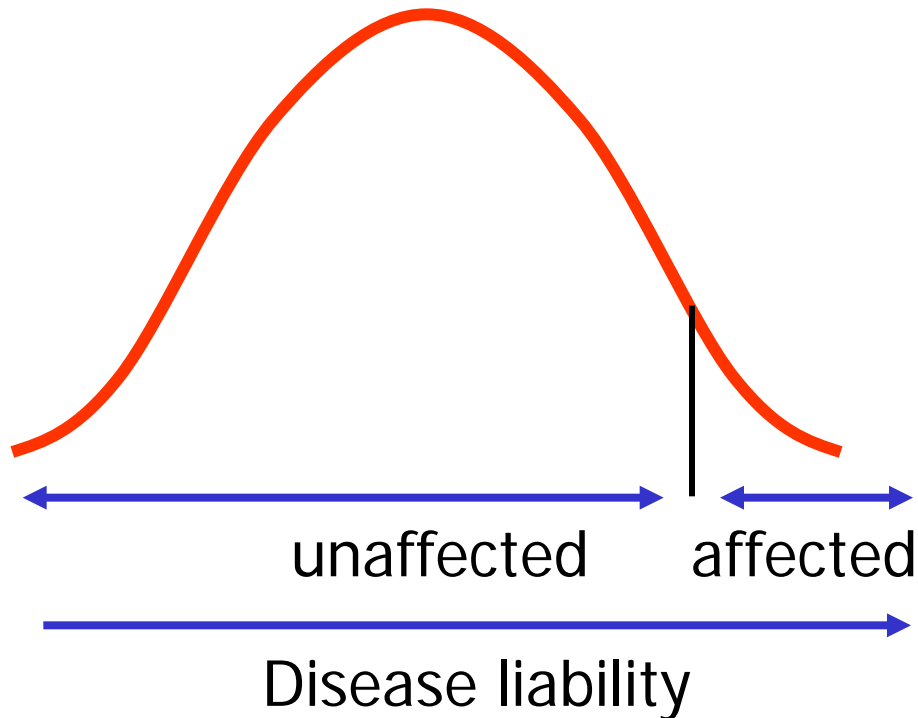
4 Genes

→ 81 Genotypes  
→ 9 Phenotypes

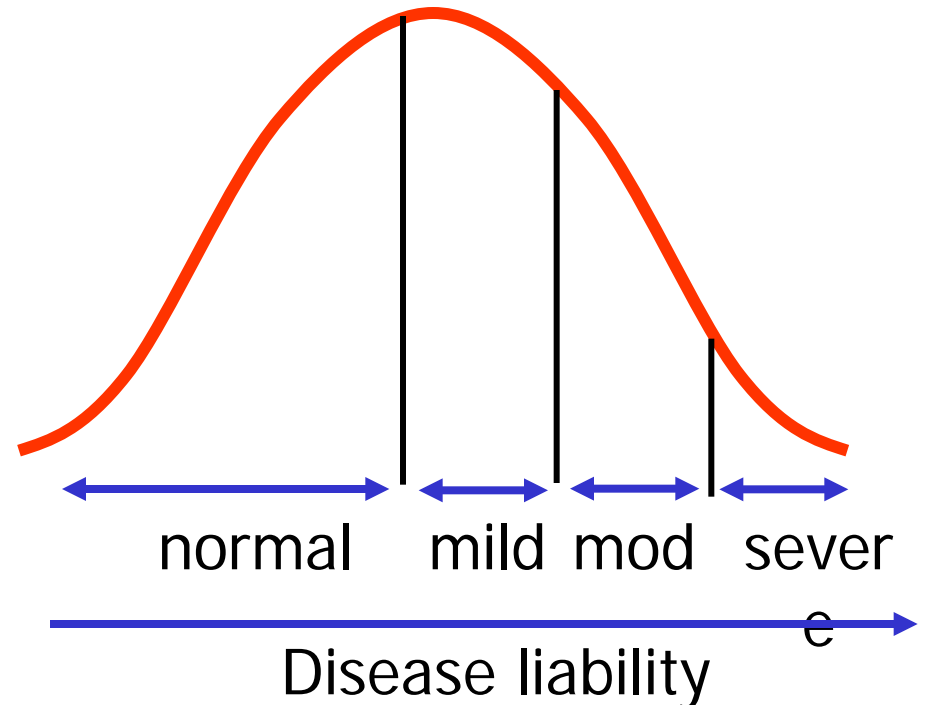


# Multifactorial Threshold Model of Disease

## Single threshold

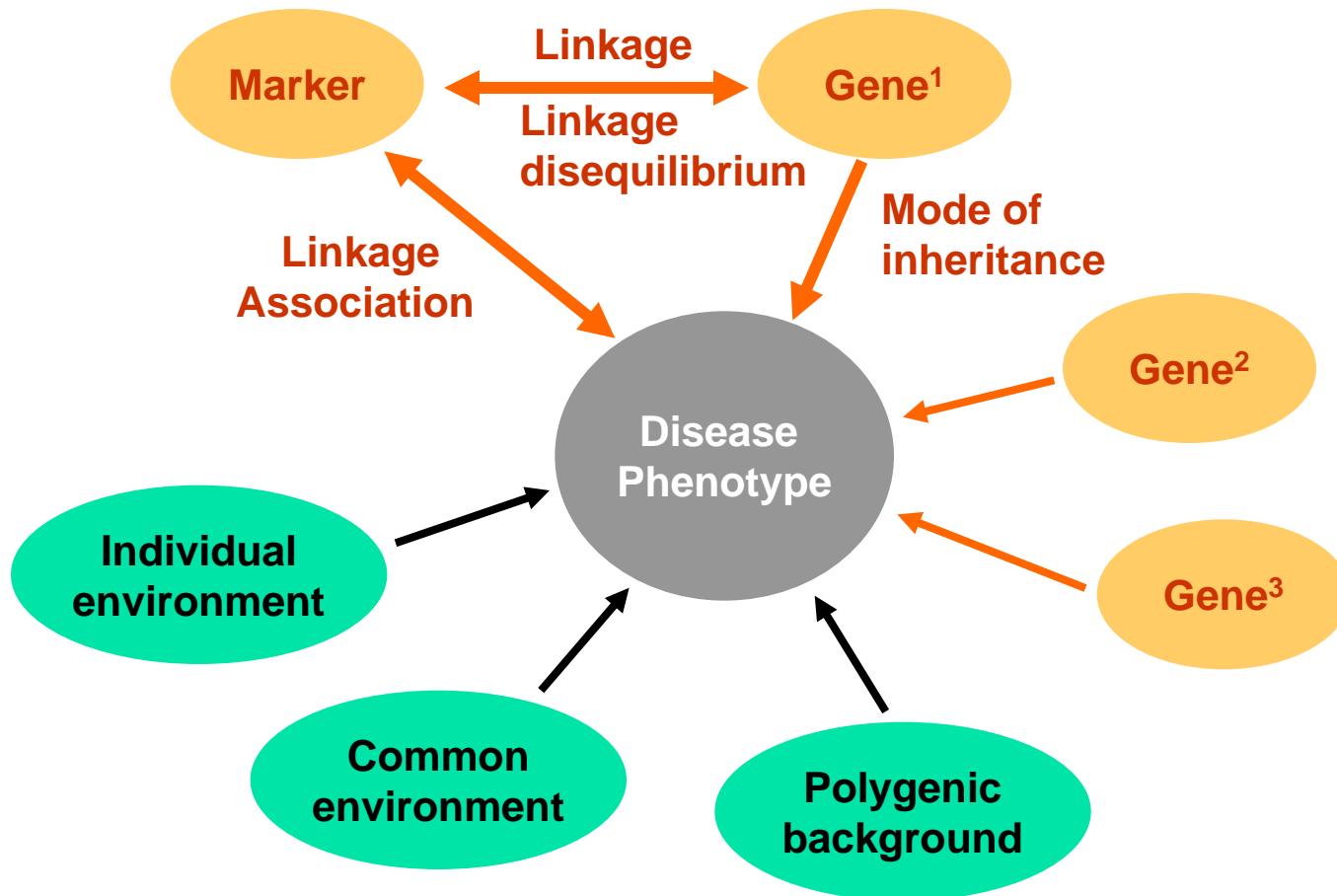


## Multiple thresholds



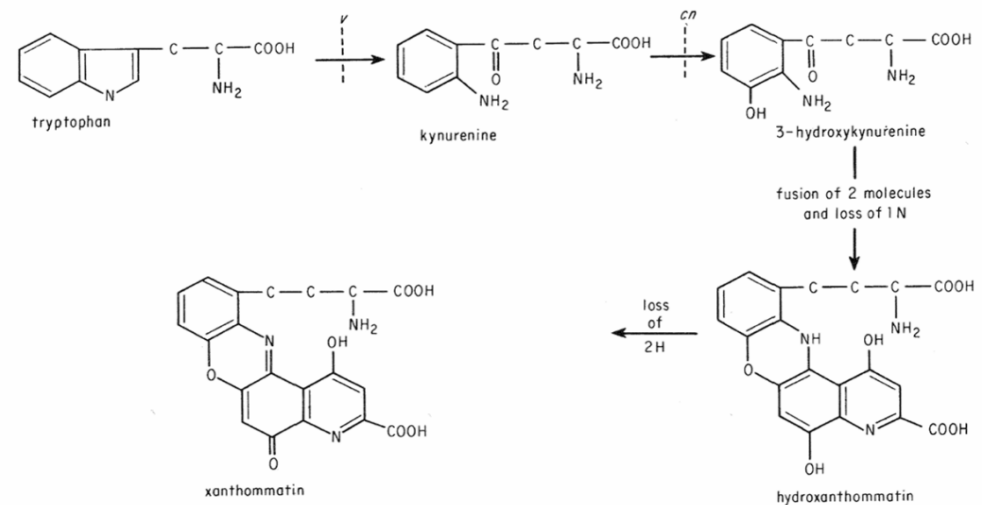
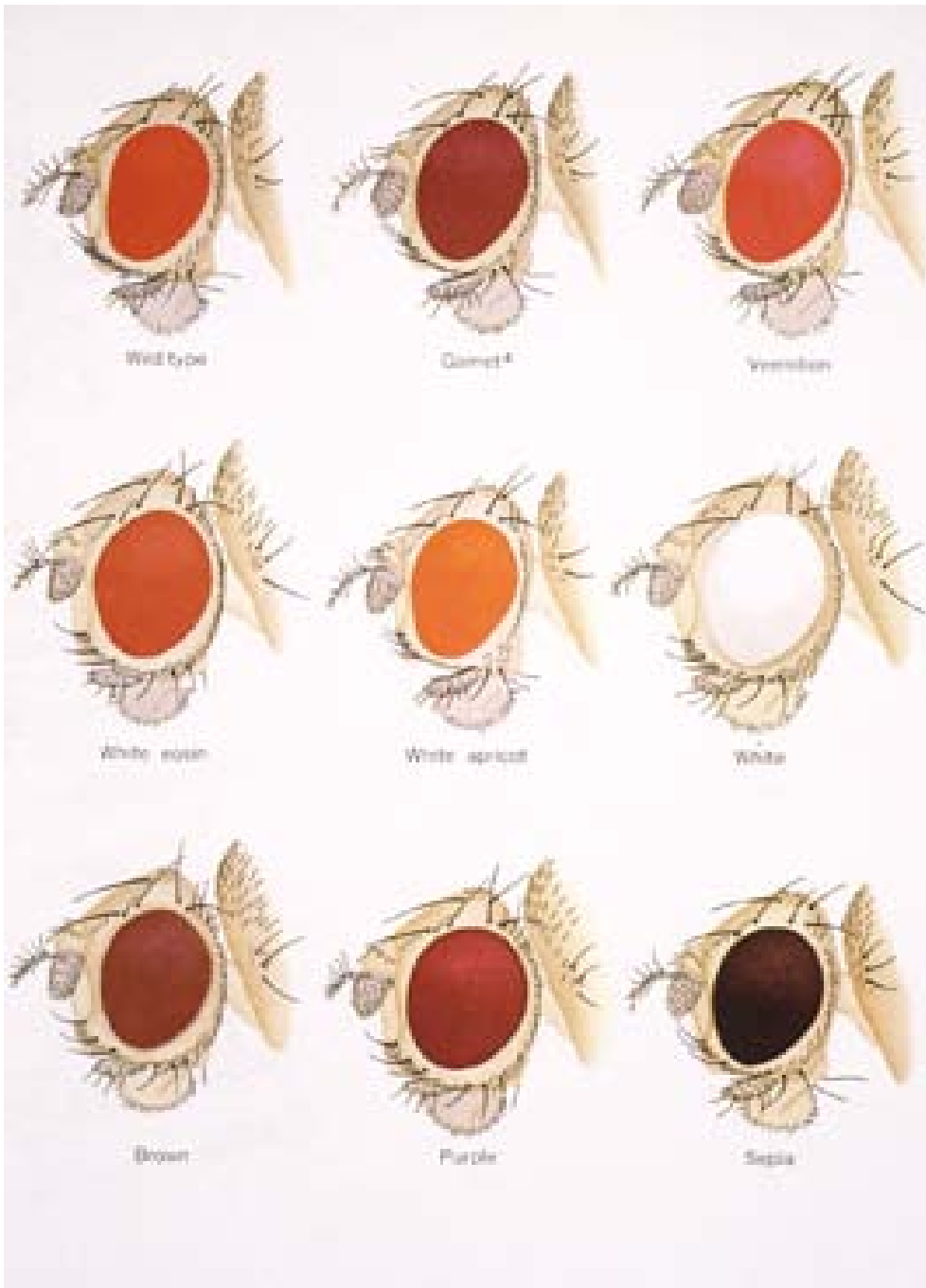
# Complex Trait Model

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# Using genetics to dissect metabolic pathways: *Drosophila* eye color

Beadle & Ephrussi, 1936







# Finding QTLs

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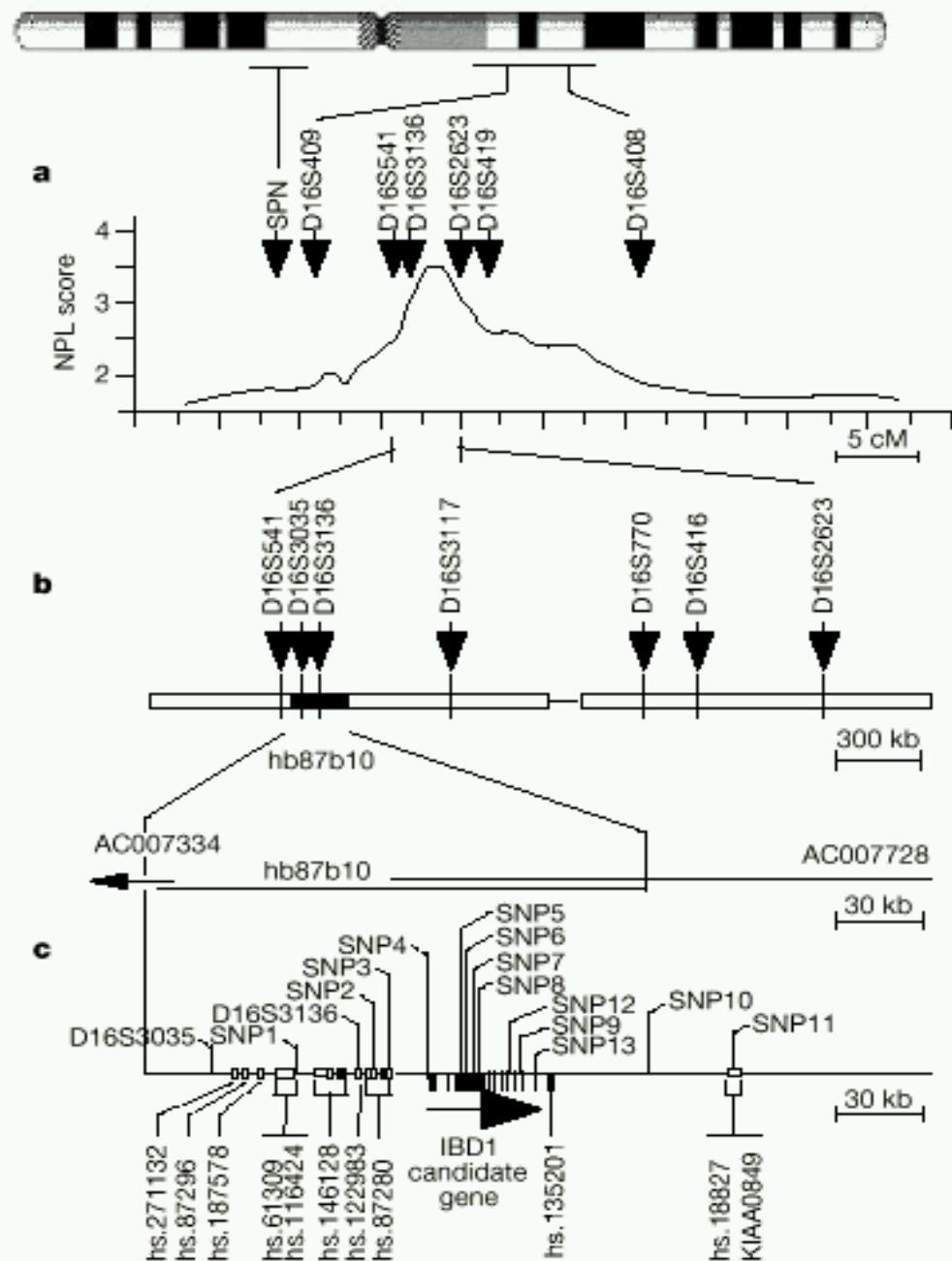
- Linkage
- Association

# Association of NOD2 leucine-rich repeat variants with susceptibility to Crohn's disease

Jean-Pierre Hugot<sup>†††</sup>, Mathias Chamailard<sup>\*†</sup>, Habib Zouali<sup>\*</sup>, Suzanne Lesage<sup>\*</sup>, Jean-Pierre Cézard<sup>‡</sup>, Jacques Belaiche<sup>§</sup>, Sven Almer<sup>||</sup>, Curt Tysk<sup>¶</sup>, Colm A. O'Morain<sup>#</sup>, Miquel Gassull<sup>☆</sup>, Vibeke Binder<sup>\*\*</sup>, Yigael Finkel<sup>††</sup>, Antoine Cortot<sup>‡‡</sup>, Robert Modigliani<sup>§§</sup>, Pierre Laurent-Puig<sup>†</sup>, Corine Gower-Rousseau<sup>‡‡</sup>, Jeanne Macry<sup>|||</sup>, Jean-Frédéric Colombel<sup>‡‡</sup>, Mourad Sahbatou<sup>\*</sup> & Gilles Thomas<sup>\*†§§</sup>

NATURE | VOL 411 | 31 MAY 2001

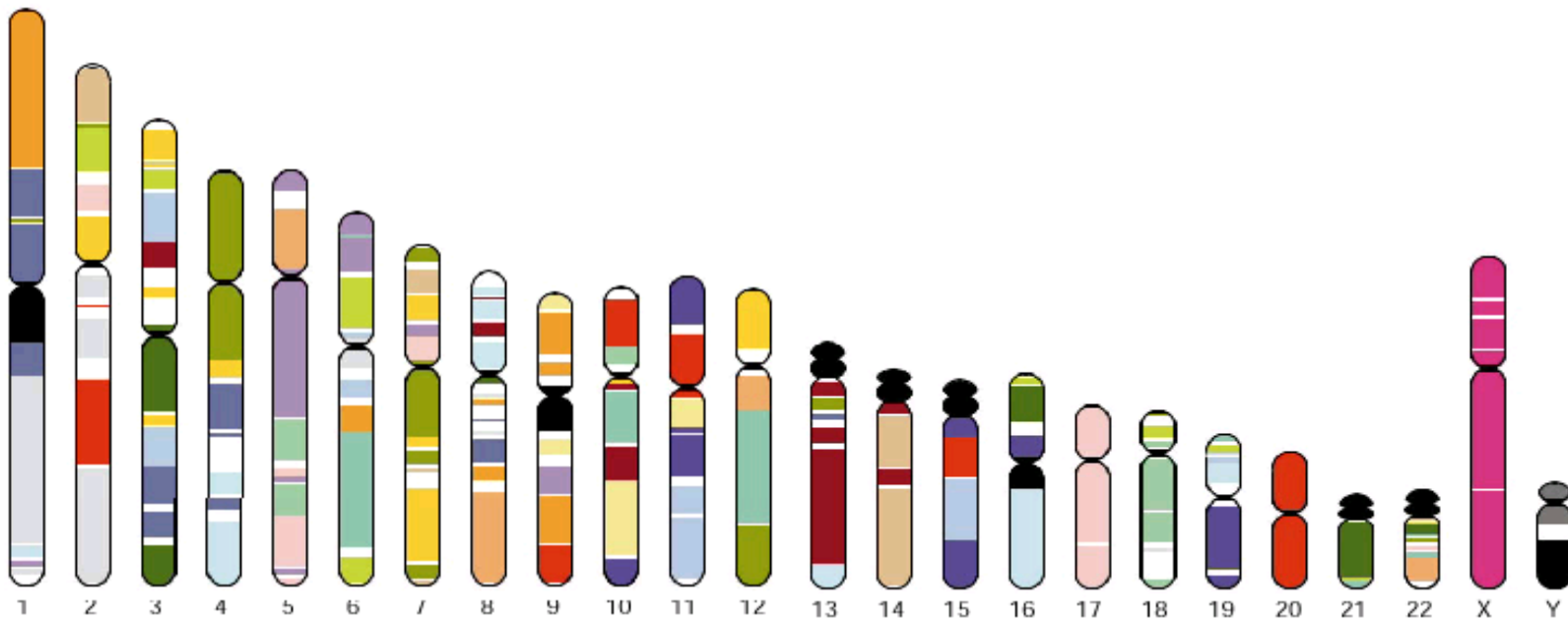
First (unequivocal) positional cloning of a complex disease QTL !





# Linkage analysis

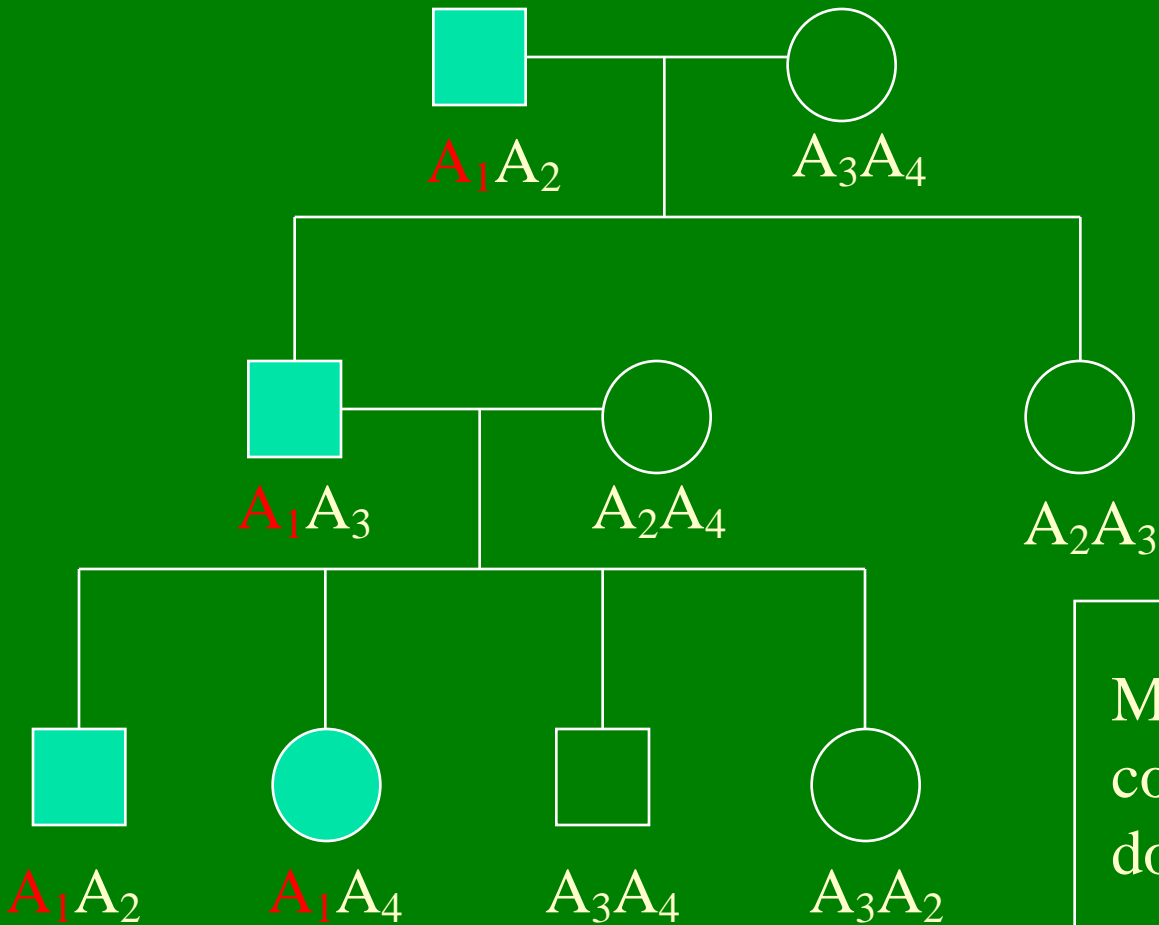
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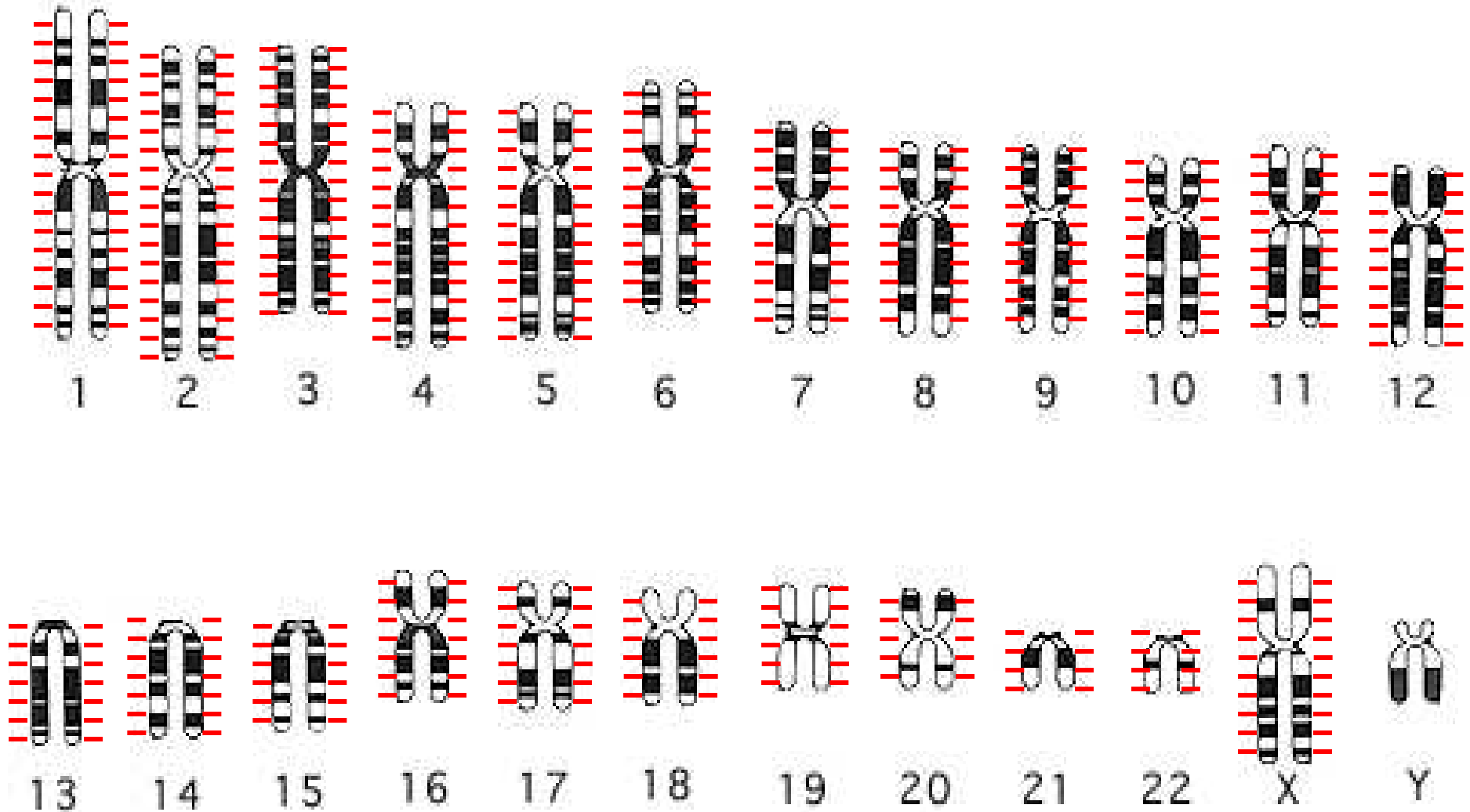
# Thomas Hunt Morgan – discoverer of linkage

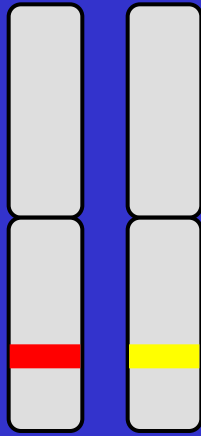


# Linkage = Co-segregation

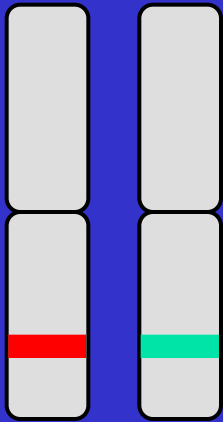
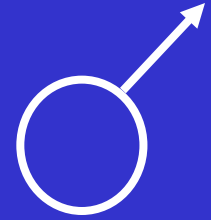
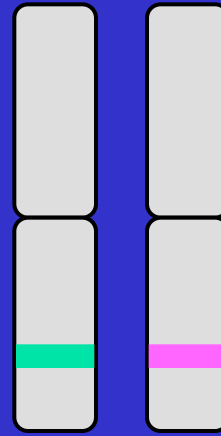


# Linkage Markers...

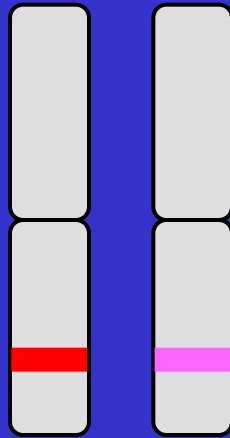




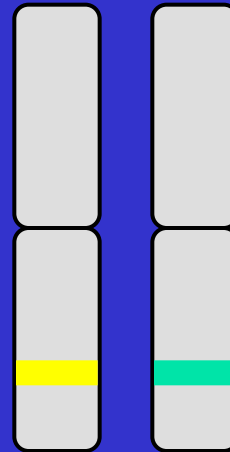
x



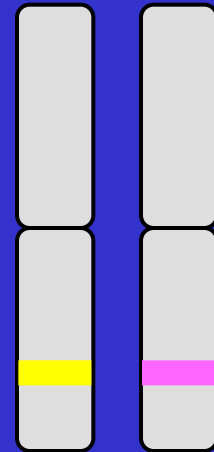
1/4



1/4



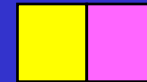
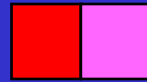
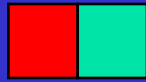
1/4



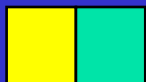
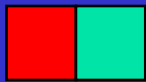
1/4

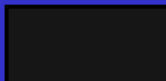
# IDENTITY BY DESCENT

Sib 1



Sib 2

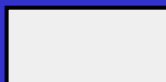


$4/16 = 1/4$  sibs share BOTH parental alleles IBD = 2



$8/16 = 1/2$  sibs share ONE parental allele IBD = 1

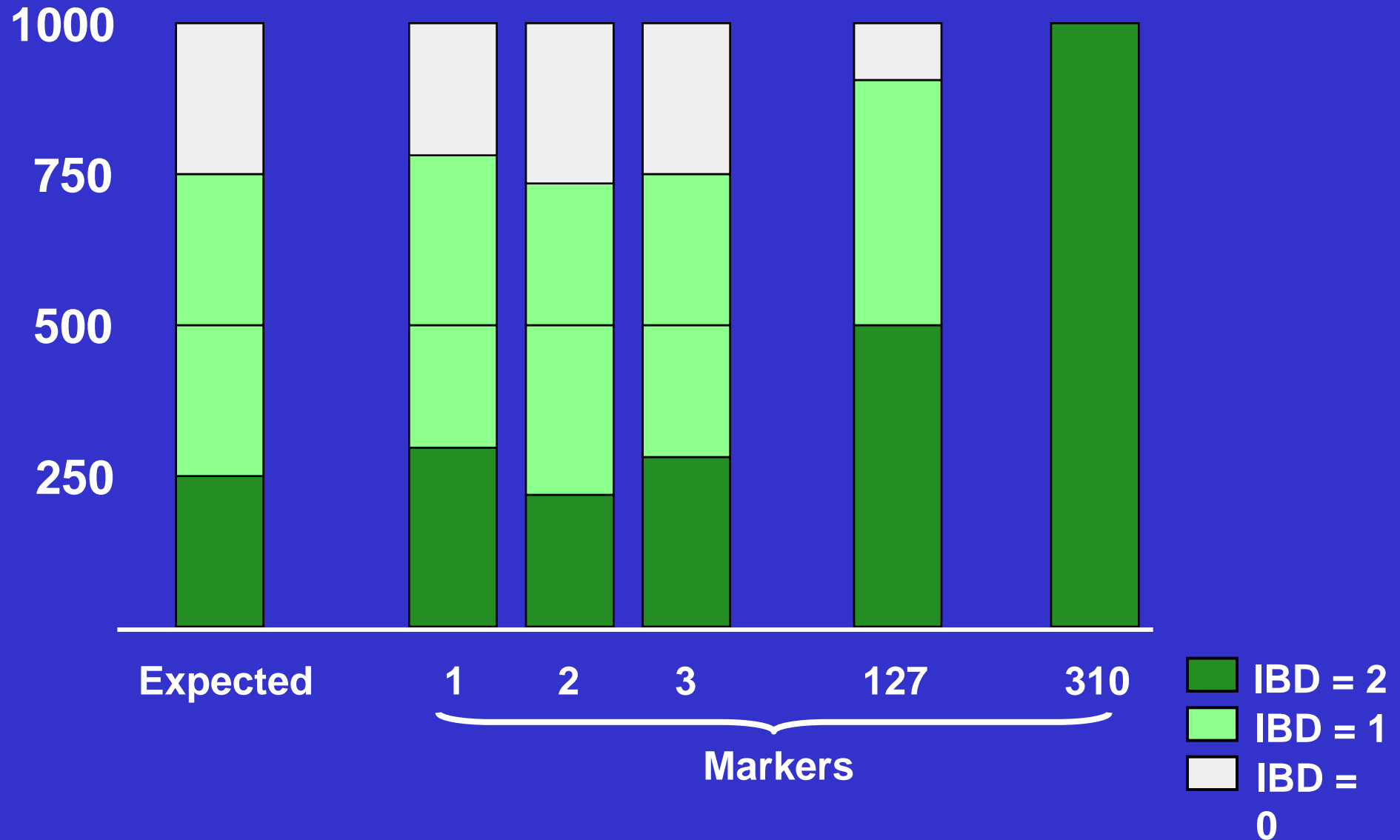


$4/16 = 1/4$  sibs share NO parental alleles IBD = 0



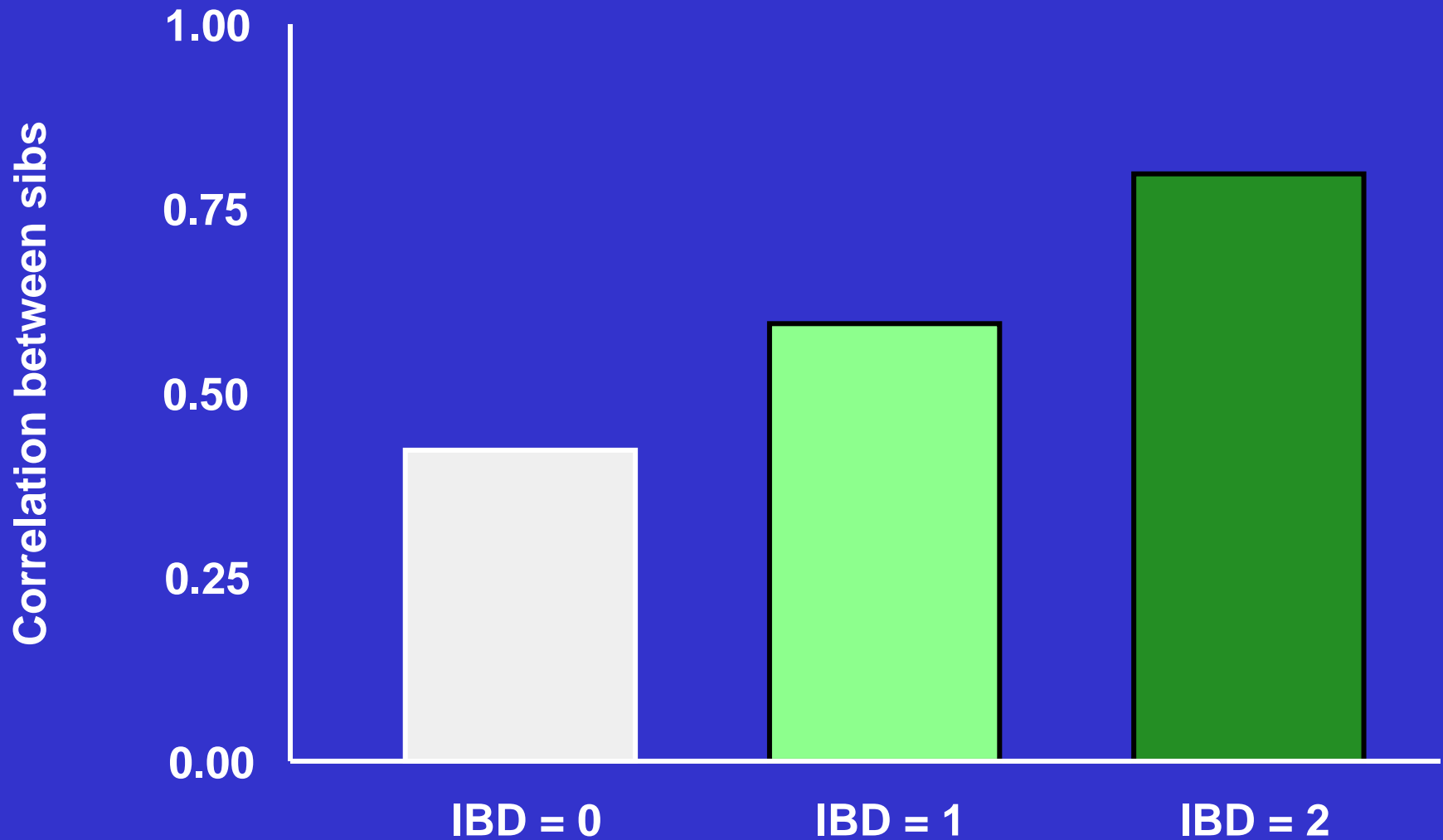
# For disease traits (affected/unaffected)

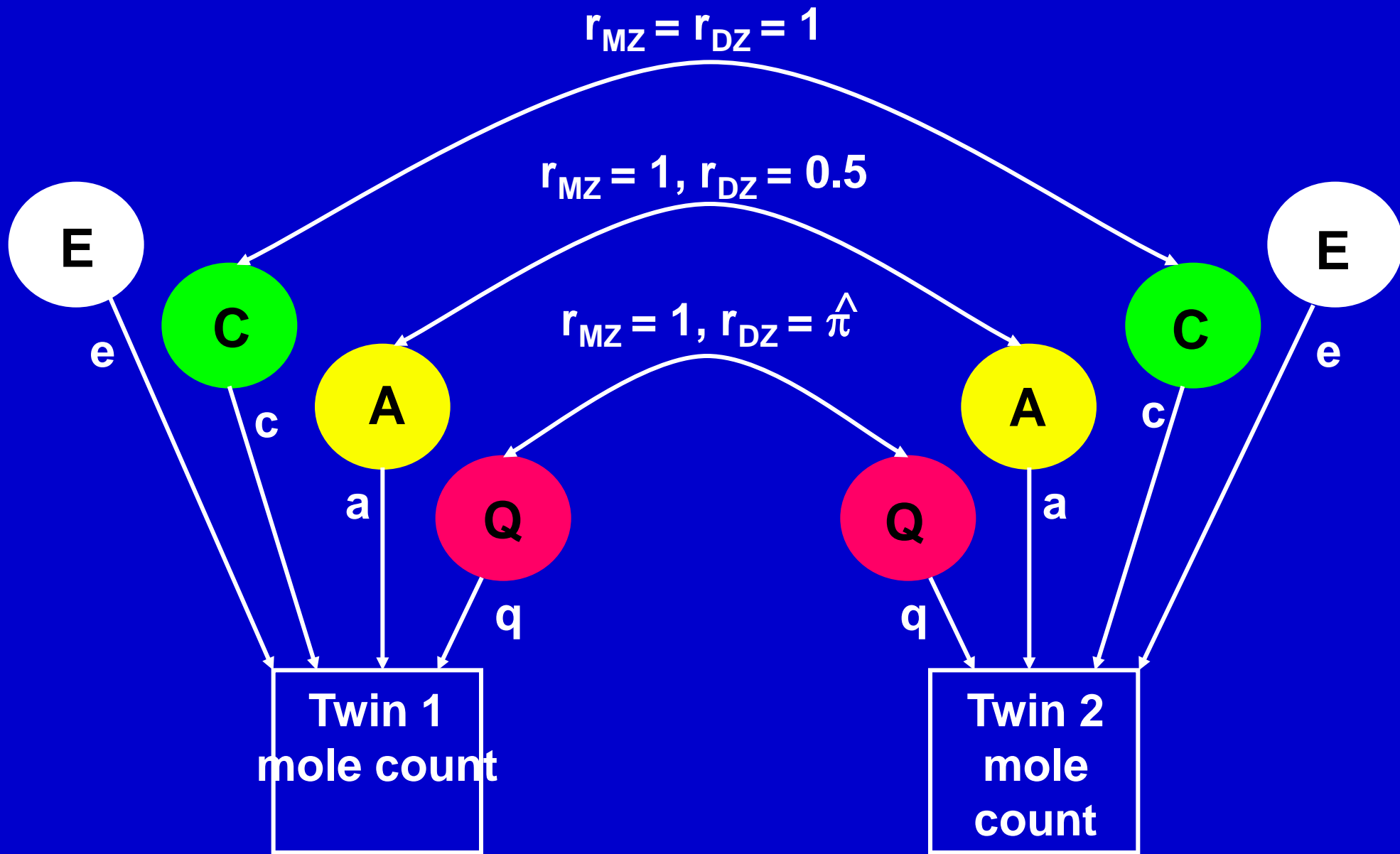
## Affected sib pairs selected



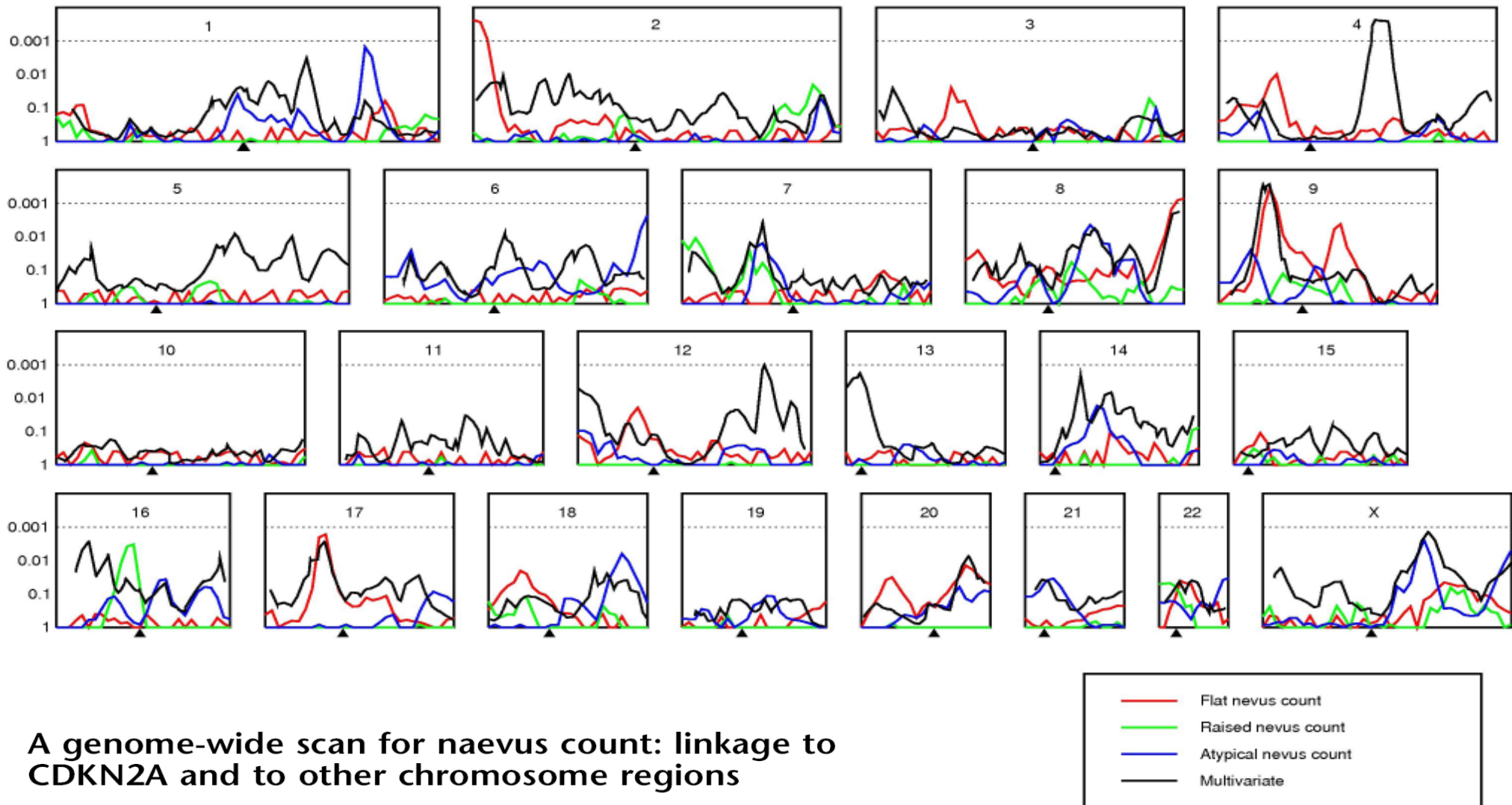
# For continuous measures

## Unselected sib pairs





# Linkage for mole counts in Australian twin families



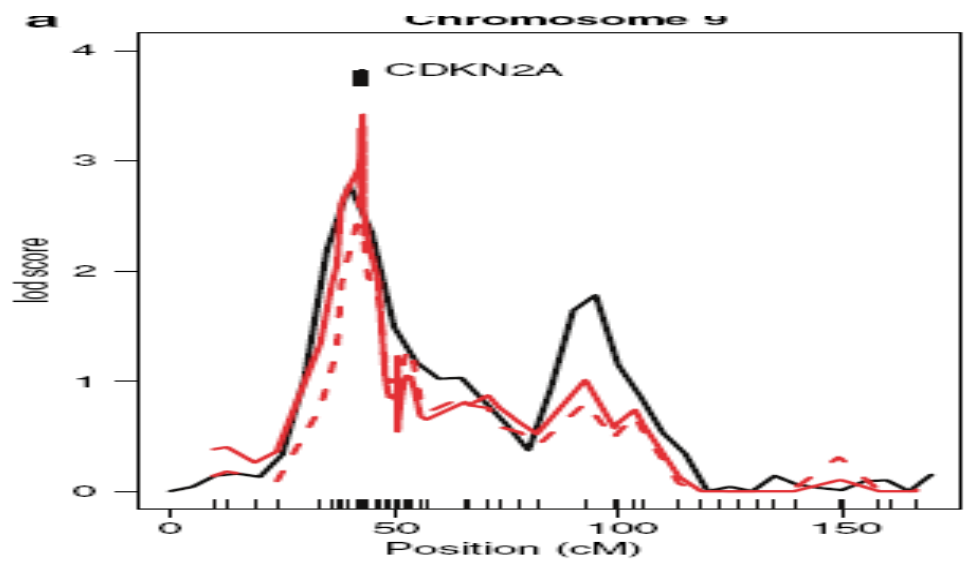
**A genome-wide scan for naevus count: linkage to CDKN2A and to other chromosome regions**

Gu Zhu<sup>1</sup>, Grant W Montgomery<sup>1</sup>, Michael R James<sup>1</sup>, Jeff M Trent<sup>2</sup>, Nicholas K Hayward<sup>1</sup>,  
Nicholas G Martin<sup>1</sup> and David L Duffy<sup>\*1</sup>

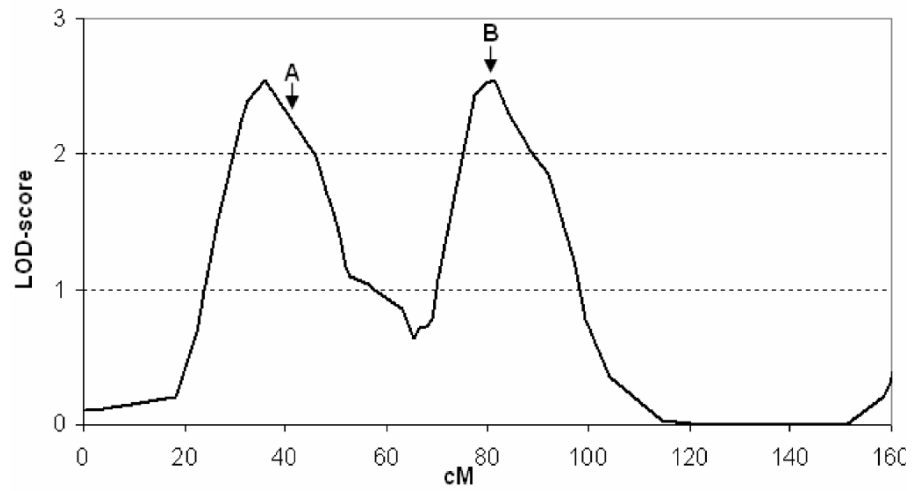
European Journal of Human Genetics (2007) 15, 94–102

# Flat mole count: chromosome 9 linkage in Australian and UK twins

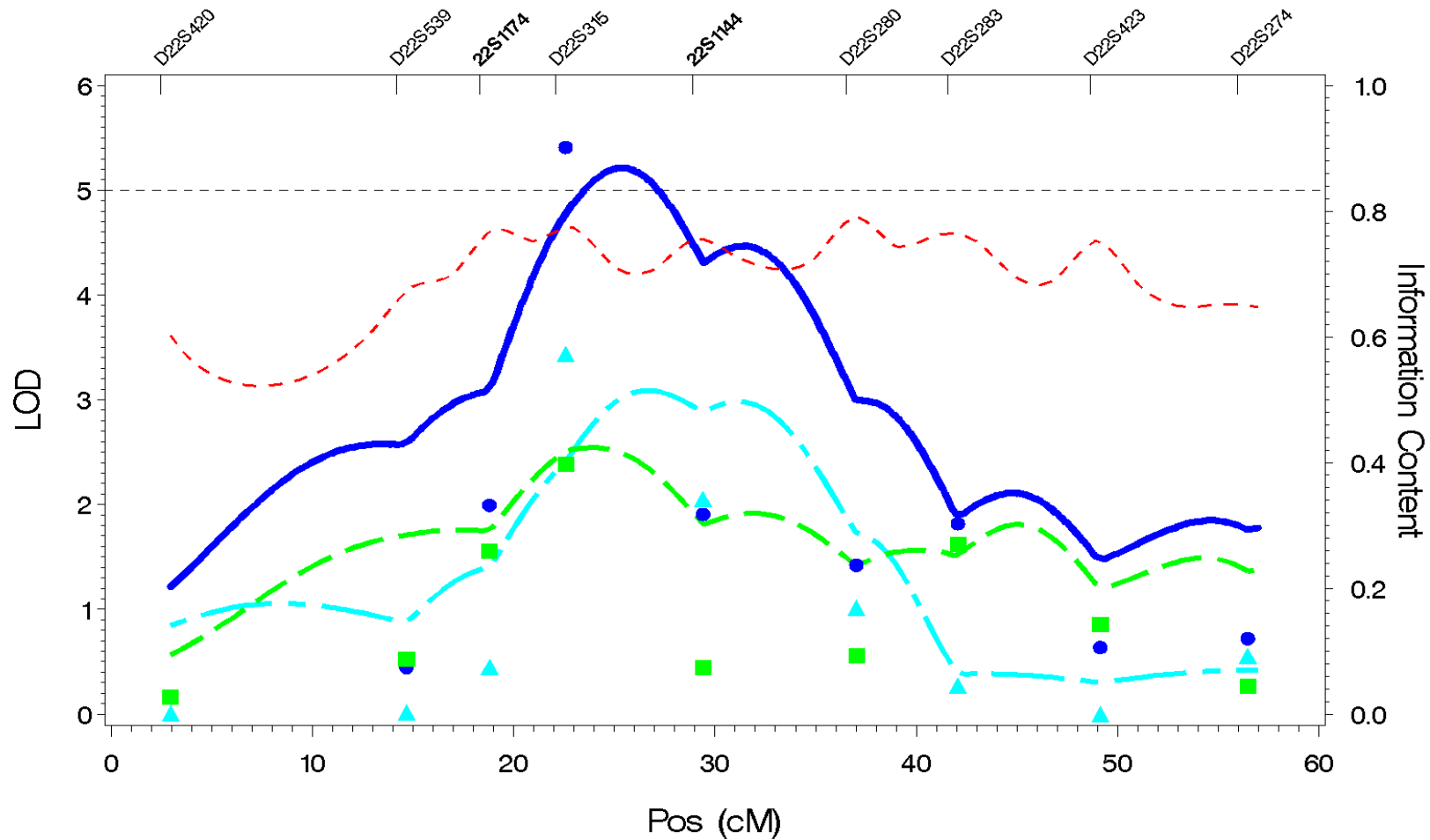
Australia



UK



# Linkage for MaxCigs24 in Australia and Finland





# Linkage

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- Doesn't depend on "guessing gene"
- Needs family data !
- Works over broad regions and whole genome
- Only detects large effects (>10%)
- Requires large samples (10,000's?)
- Can't guarantee close to gene
- **On the whole, has failed to deliver for complex traits**



# Association

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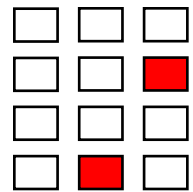
- Looks for correlation between specific alleles and phenotype (trait value, disease risk)



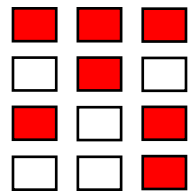
# Association studies

	1	2	3	4	5	
	1234567890	1234567890	1234567890	1234567890	1234567890	
191,044,935	GTGAGTATGG	CTTTCTTCCC	TCTCCCGCCA	CCCCCTGCC	CACACTGCAA	Intron 1
191,044,985	GCTGCAAACG	CGGTACTTTC	GGGCTCGCCT	TTGACGTTAG	GAAACTAGCC	
191,045,035	TGAGCCTATG	CAGGGAAAAA	AAATCGAAAA	GGTCAATTTG	TTAAGTAAGG	
191,045,085	TTAAATCTGG	GTGATGCTCG	GGTACAGTTT	AAGAACCGAG	GGAGACAGTT	
191,045,135	GATATGAGGG	CGGTGGTTGA	TGCGCTAAGA	AATTGCGGGT	TGGCTTTTTG	
191,045,185	TCCTCCTGCA	TTCAAAATGA	CATCAGAATC	CTGCGGCTGA	AGCGCGTCCC	<a href="#">rs34774923</a>
191,045,235	CAGCATTTCAT	ACGTTGCATG	ATGAGTTTCT	ATCAGCTTAC	ACAGCTACTG	
191,045,285	GAAGGTGATG	CTCTTGCTGG	TTCTGAATAT	ACTCGTTTAA	AATCCATTTT	
191,045,335	TGTTTTTTAA	TTATAGAGCA	GATCTCACCC	AGTCCGAATG	TGGAACAAAT	
191,045,385	AATTGTTATG	CAGCGTCTGC	TTAAAAGAAG	TGTCGTAGGT	GGGGAAGGAA	
191,045,435	GAGCGCAGGG	GAATCAGTCA	CCCACCTCTT	TGTACAGTCT	CTGGCGTGGT	<a href="#">rs10218513</a>
191,045,485	CCAGAACCTC	CTGCTCTAAA	GAGAGAAGCG	TGGGCCGGCT	CCAGACAGTT	
191,045,535	CCATGTCTGT	CCTTTTCATT	AAAGTGCAAA	ACGTCTCGGA	ATTGTAATTA	
191,045,585	ACCTTGCAAA	CAAACCTGATG	CCCTTTGTGA	GCCAGAAATA	GTGTCTGCCT	
191,045,635	TTTGAACATA	ATTCATTAAC	AATCTTTTAA	AATACCTTAG	TGATTATAGG	
191,045,685	TAGCCCTGCC	CTTAGTTGTA	AAACTAGTAG	ATACGGTCAG	ATTAATGGAC	
191,045,735	GAAACTGCTC	AGTACATGAG	GTTTAAATGT	TAGGTGGATA	AGACTTATTT	
191,045,785	GAAGAGTTCT	TGCTTTGCCT	TATGCGGTTT	GTCTCTAGTT	ACTGGGTGAC	
191,045,835	TTTATTTGGT	AAAAATGCGT	TCAGCTGCAG	TAGCATATTC	AAGTGTGCT	<a href="#">rs2746073</a>
191,045,885	AGTTAGTAAT	TATCTTTTTA	ATTTTTTGT	TTAG		

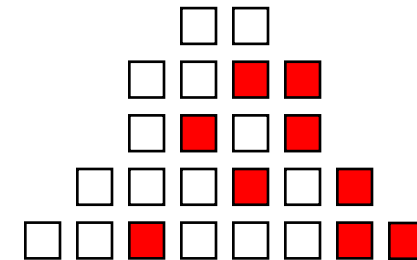
Single Nucleotide Polymorphisms



Controls



Cases



Quantitative Traits

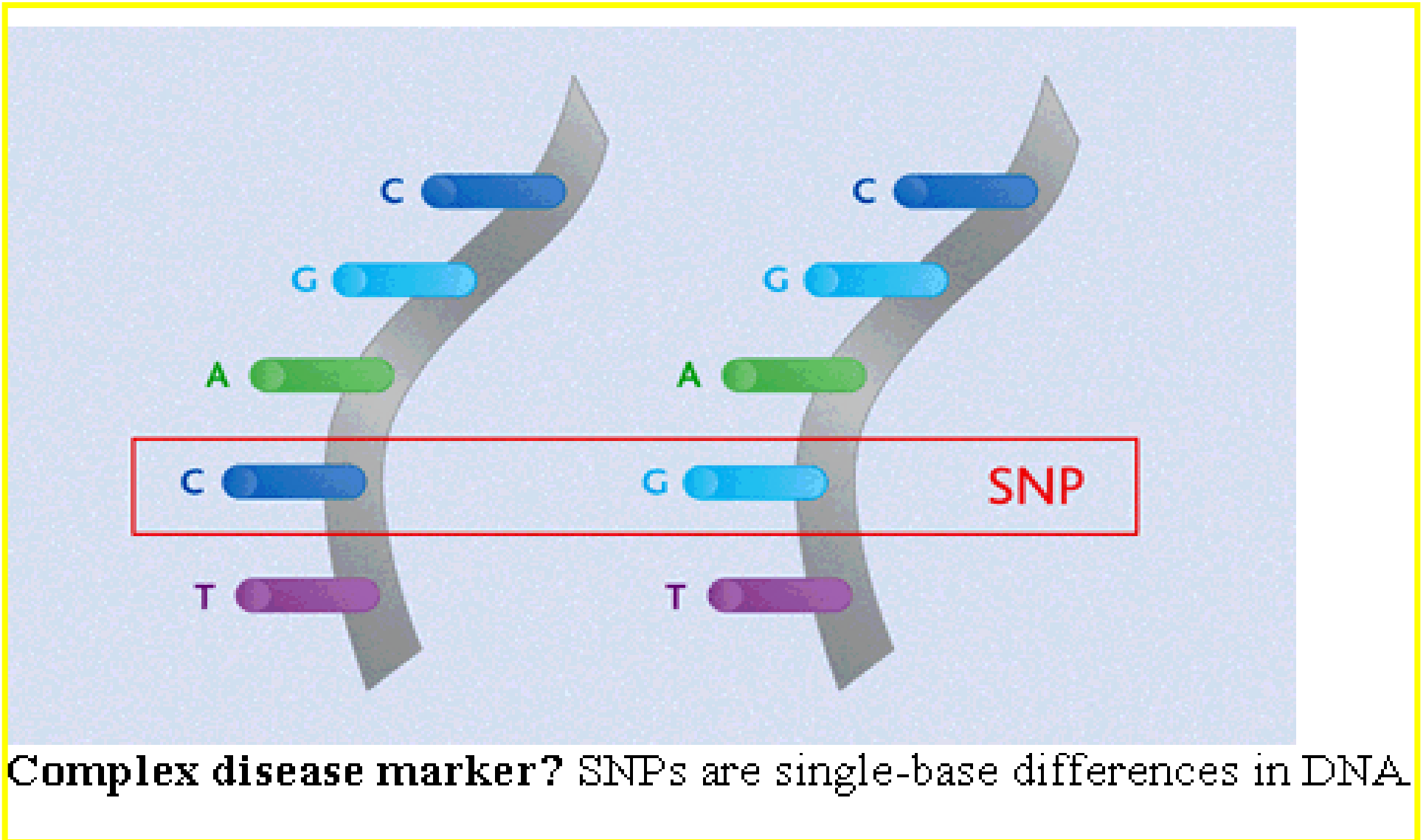


# Association

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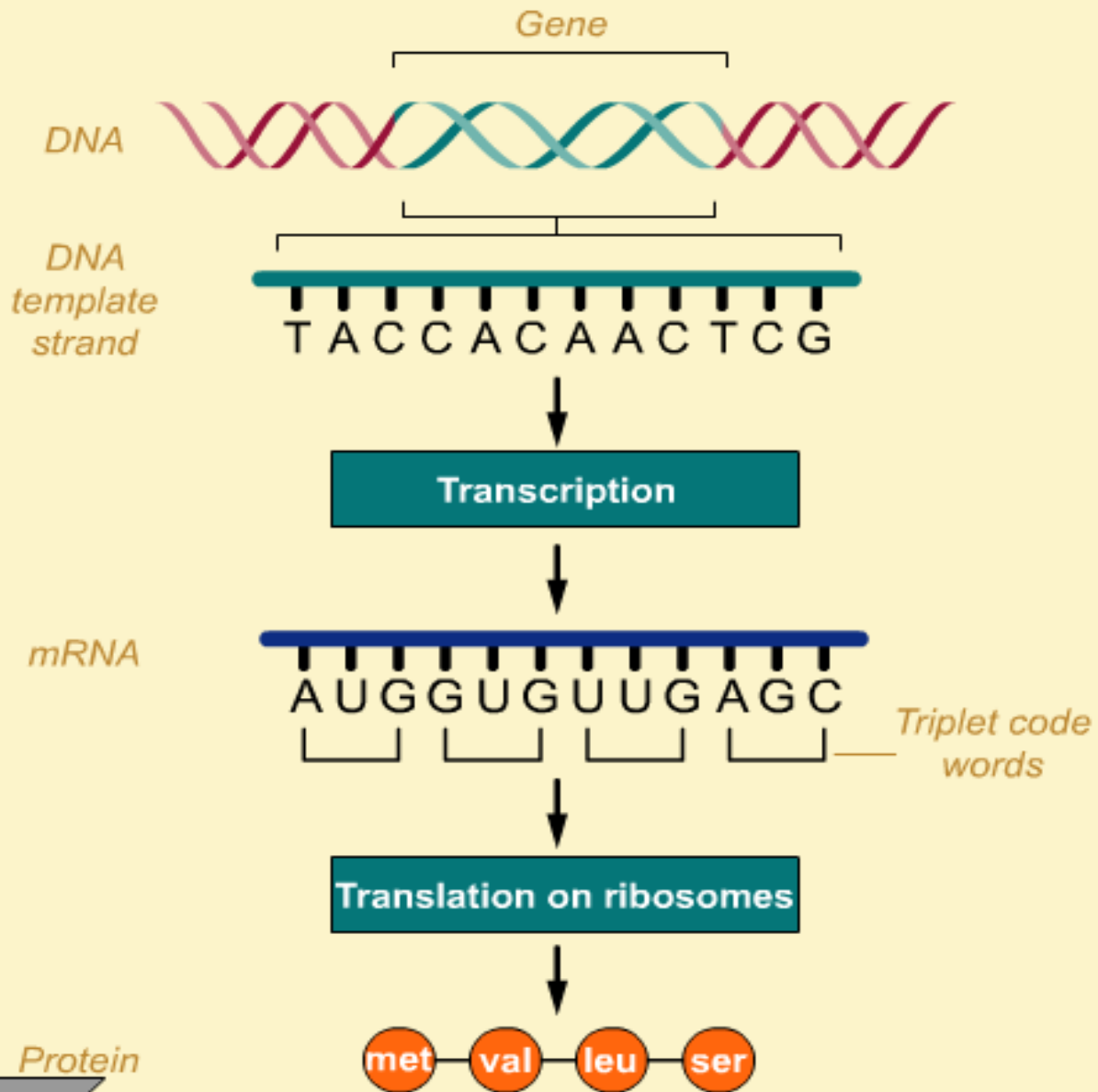
- More sensitive to small effects
- Need to “guess” gene/alleles (“candidate gene”) or be close enough for linkage disequilibrium with nearby loci
- May get spurious association (“stratification”) – need to have genetic controls to be convinced

# Variation: Single Nucleotide Polymorphisms



Complex disease marker? SNPs are single-base differences in DNA.

# The Flow of Genetic Information



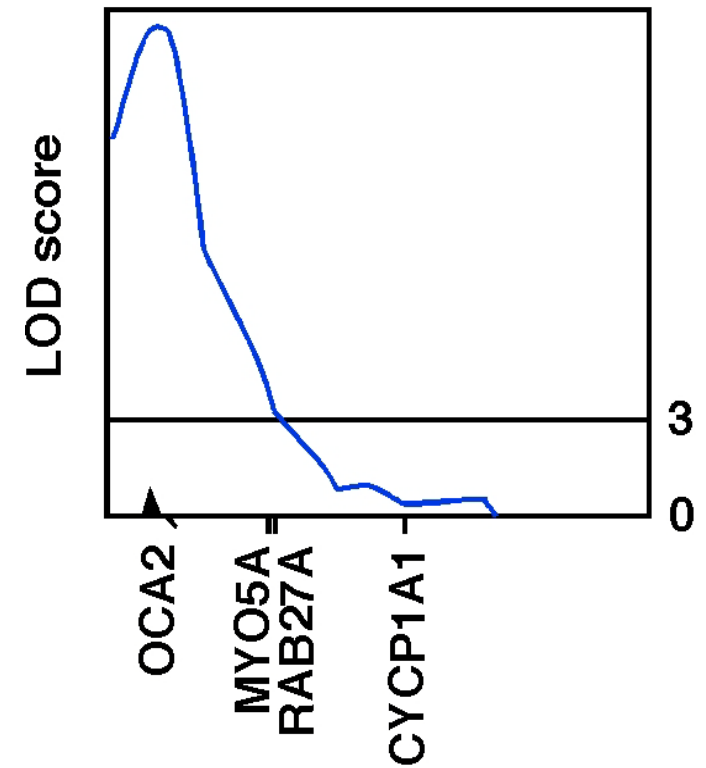
Differences (between subjects) in DNA sequence are responsible for (structural) differences in proteins.

# Human OCA2 and eye colour



QTL for Eye Colour

Chromosome 15

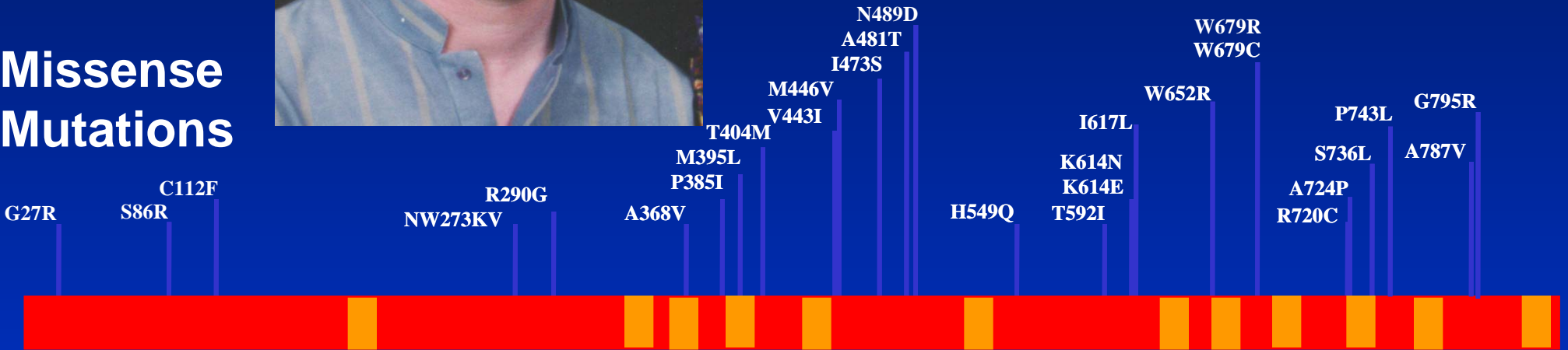


Zhu et al., *Twin Research* 7:197-210 (2004)



# Oculotaneous Albinism type 2

## Missense Mutations

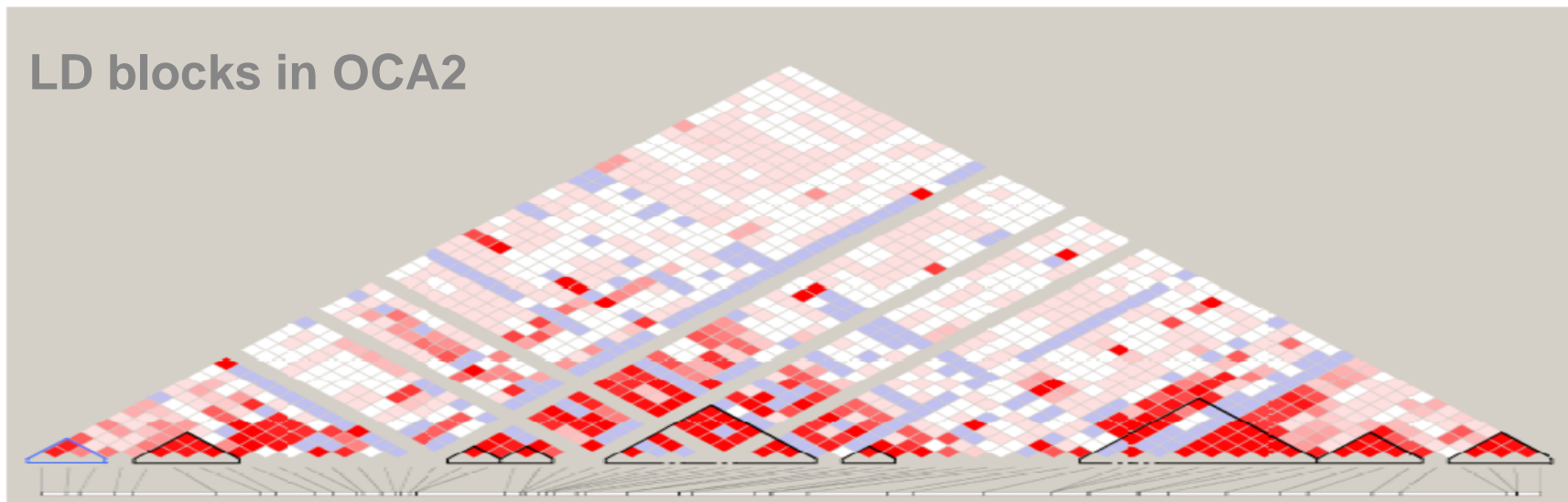


## Polymorphisms

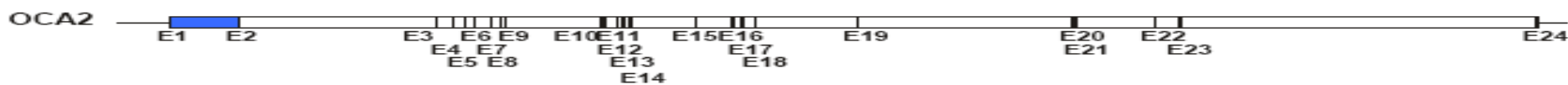
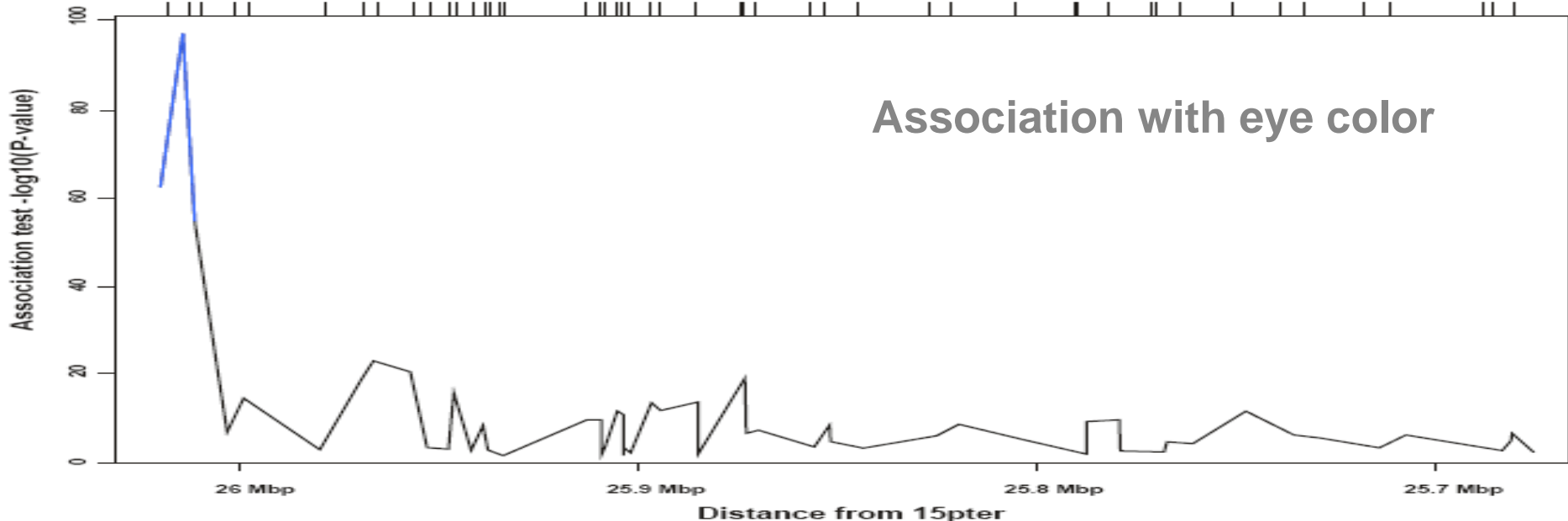


# OCA2 Mutations

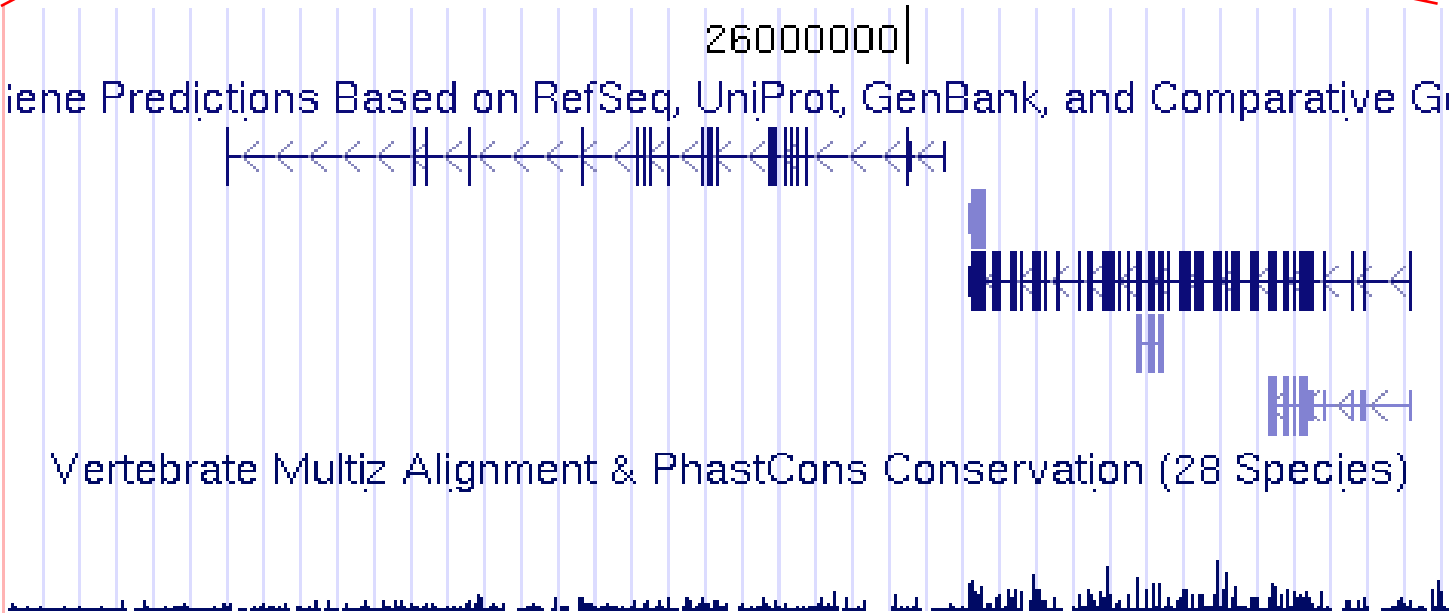
# LD blocks in OCA2



rs7495174 rs6497268 rs11855019 rs4778137 rs7179994 rs1448481 rs1597195 rs1375164 rs1448485 rs1716759 rs2290100 rs3794604 rs5553 C/G rs277875 A/G290Gly rs2015343 Val380Met rs2834272 Ala481Thr rs4778220 Val519Ala rs17170989 rs1800414 rs2311843 rs977589 rs1800416 rs2703922 rs2594602 rs2703969 rs1800417 rs1800418 rs1973448 rs1004611 rs12562307 rs4778190 rs4778185 rs1498521 rs1874835 rs17673969 rs3947367 rs1498519 rs989869



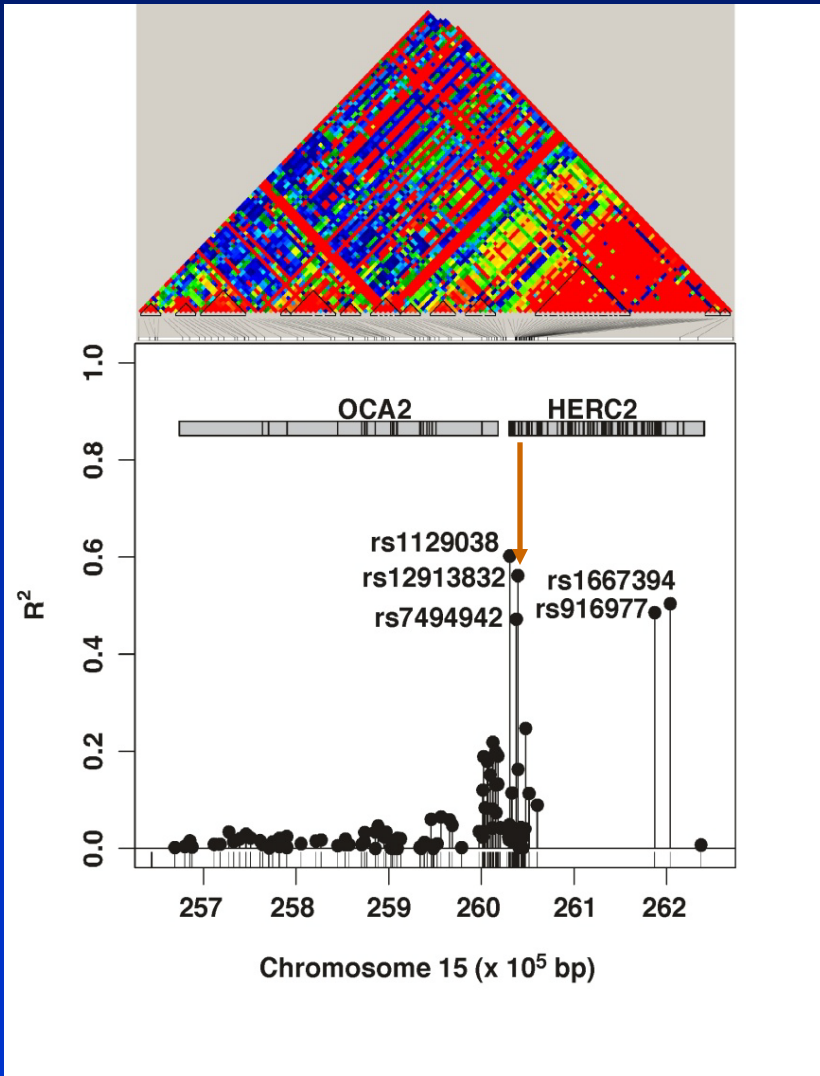
chr15 (q12-q13.1)





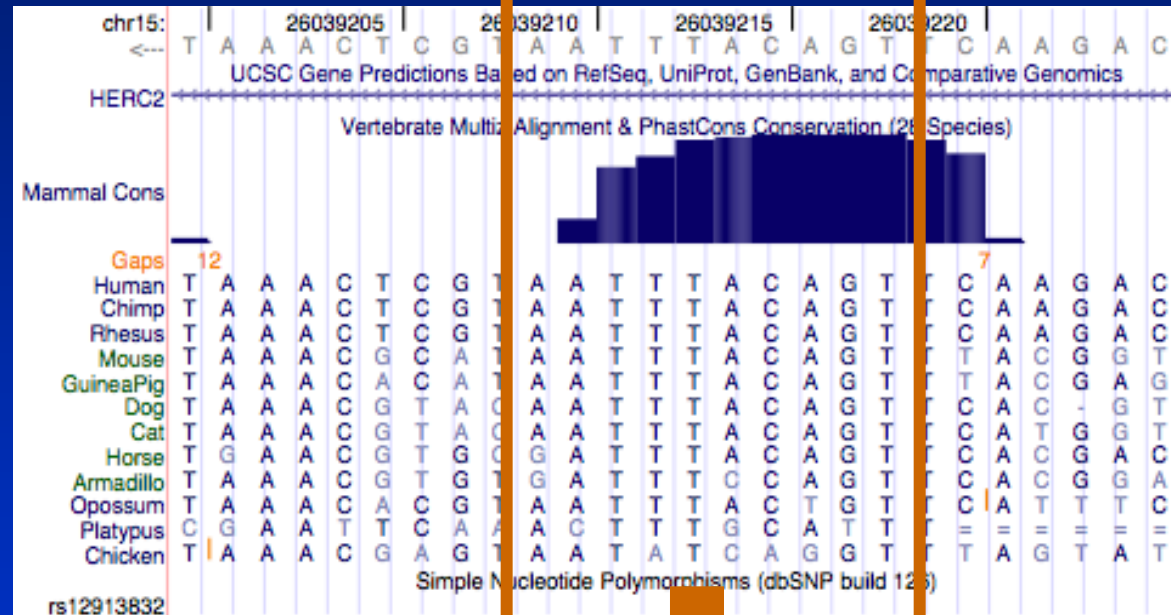
# A single SNP within intron 86 of HERC2 determines Blue-Brown eye colour

Sturm et al., AJHG 82: 424-431, 2008



rs12913832 C = Blue

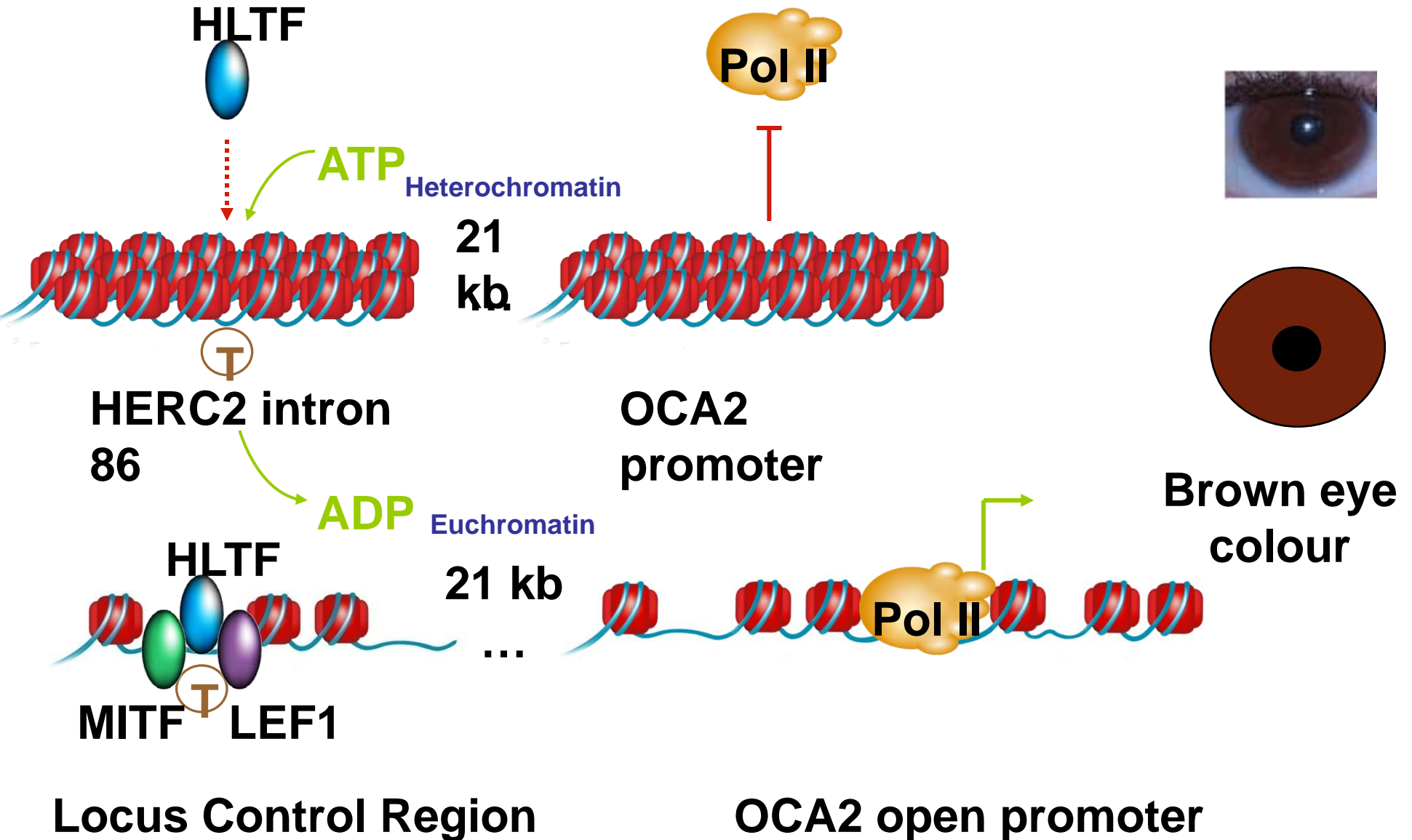
rs12913832 T = Brown



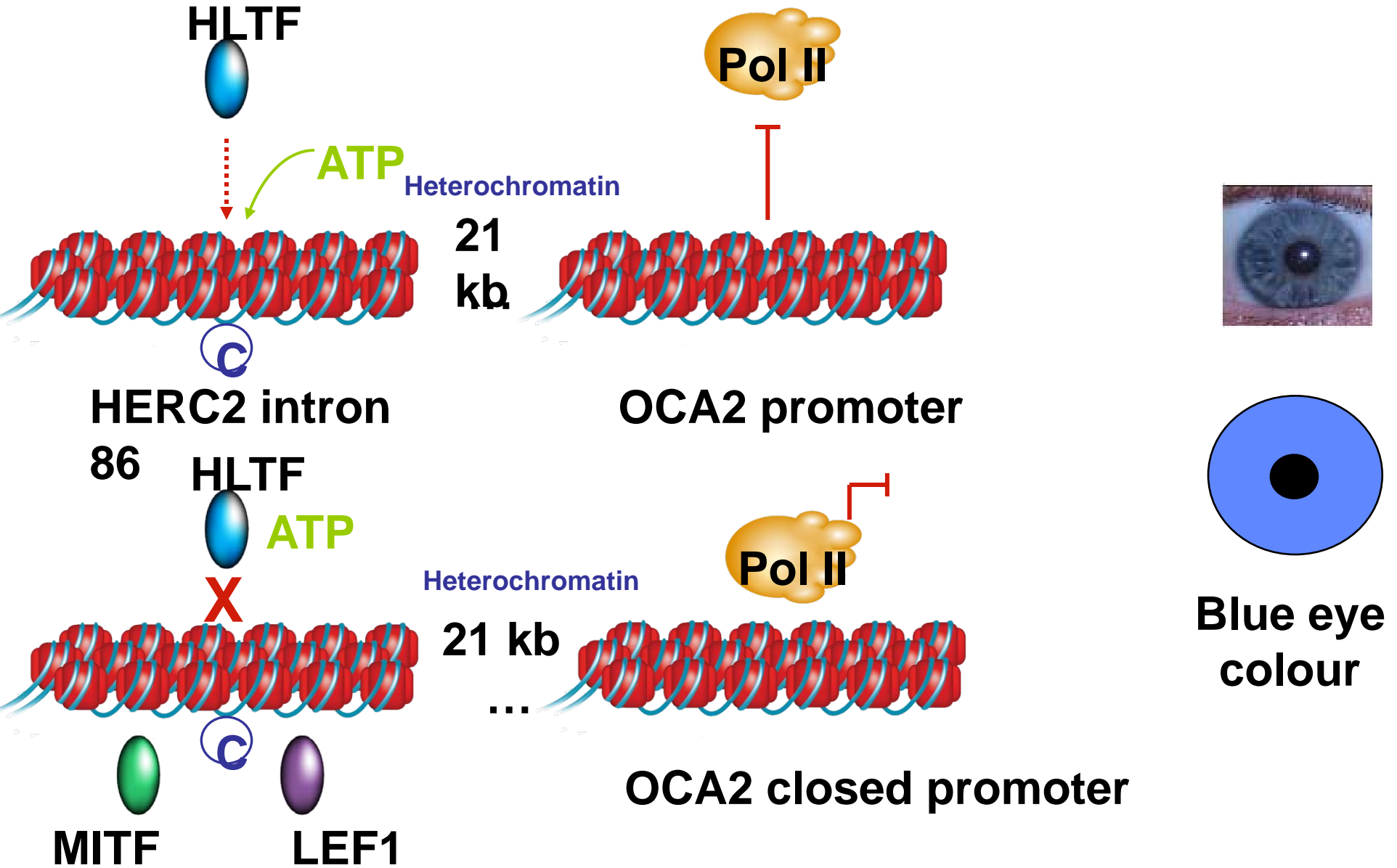
HLTF binding

HLTF = Helicase-like transcription factor  
 Uses ATP for energy to initiate unwinding of DNA, so allowing transcription

# SWI/SNF Chromatin remodelling model for OCA2 gene activation



# SWI/SNF Chromatin remodelling model for OCA2 gene activation

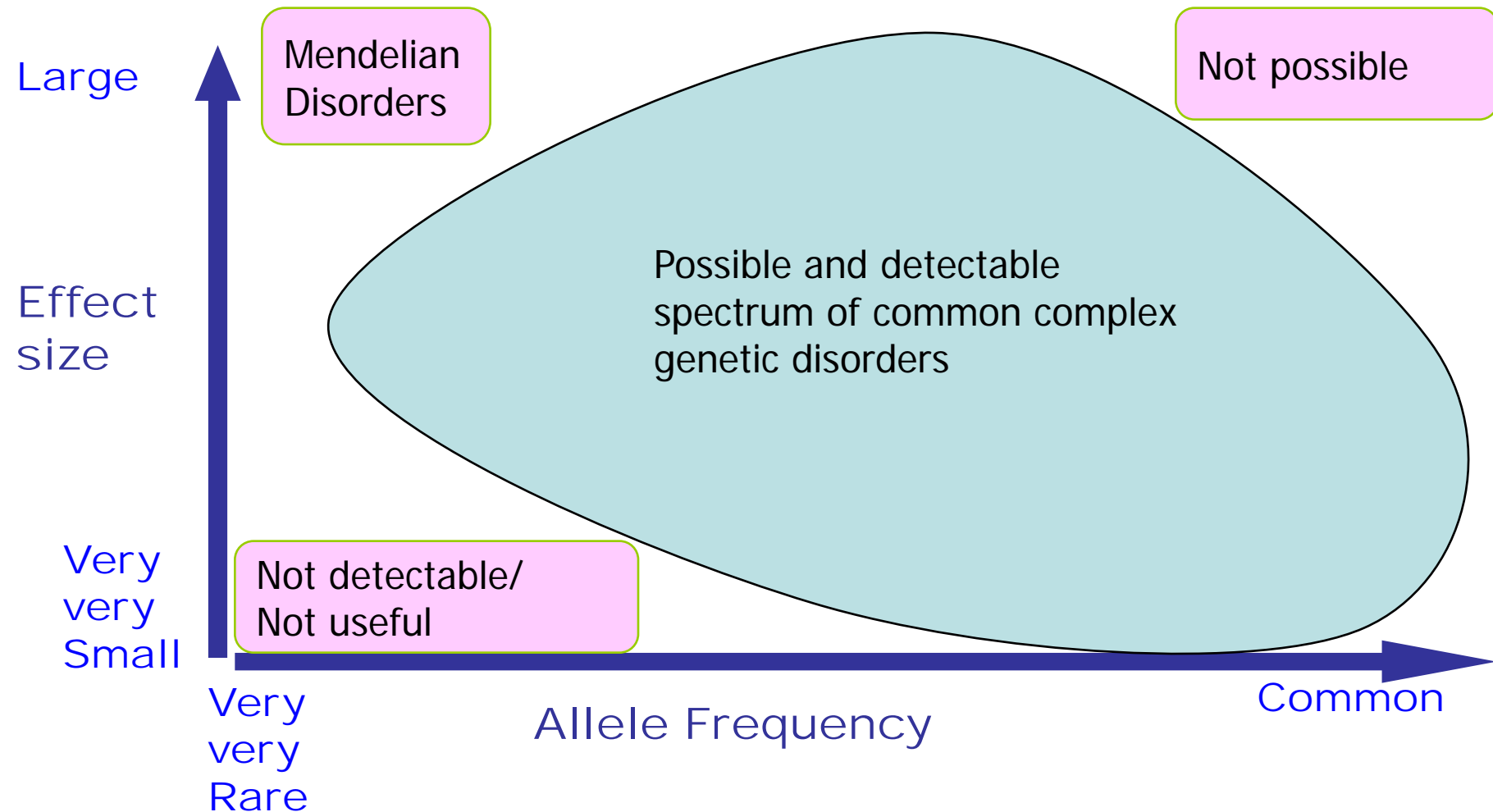


# Genetic architecture

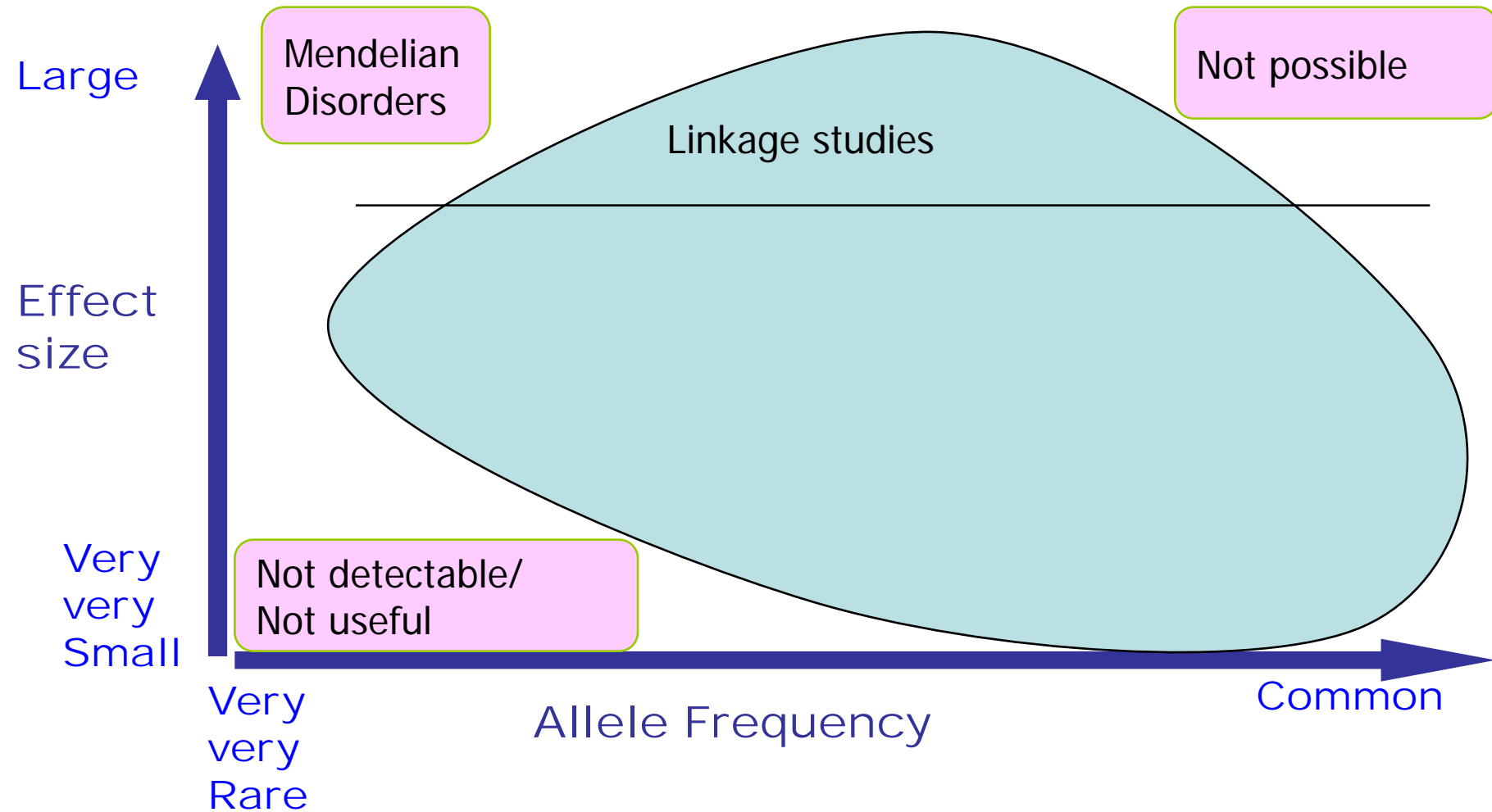
- Number of variants
- Effect size of variants
- Frequency of risk allele
- Genetic mode of action

} **Genetic variance**

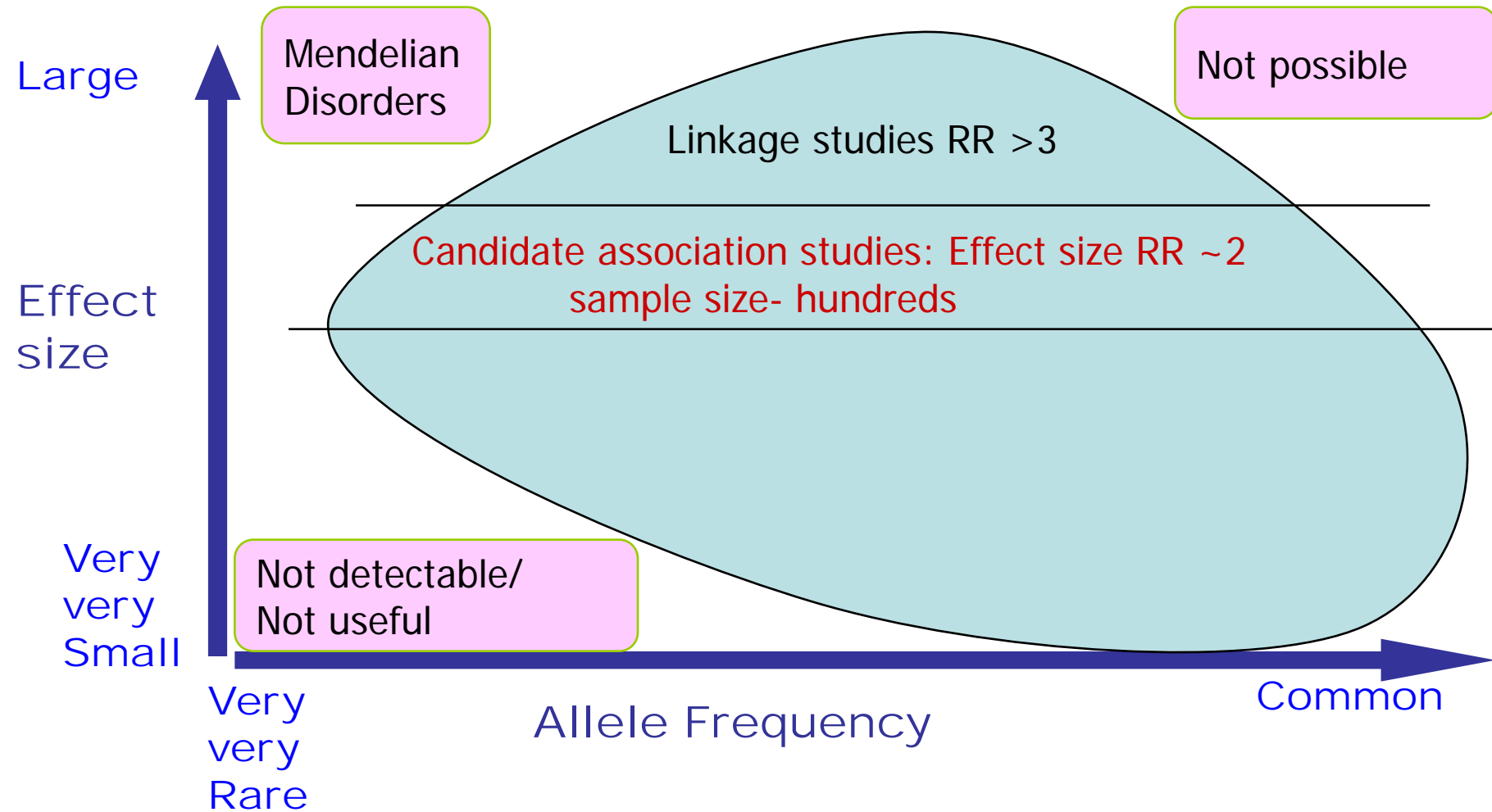
# What is the genetic architecture of complex genetic disorders?



# A potted history of genetic studies



# A potted history of genetic studies



# Association studies until 2007

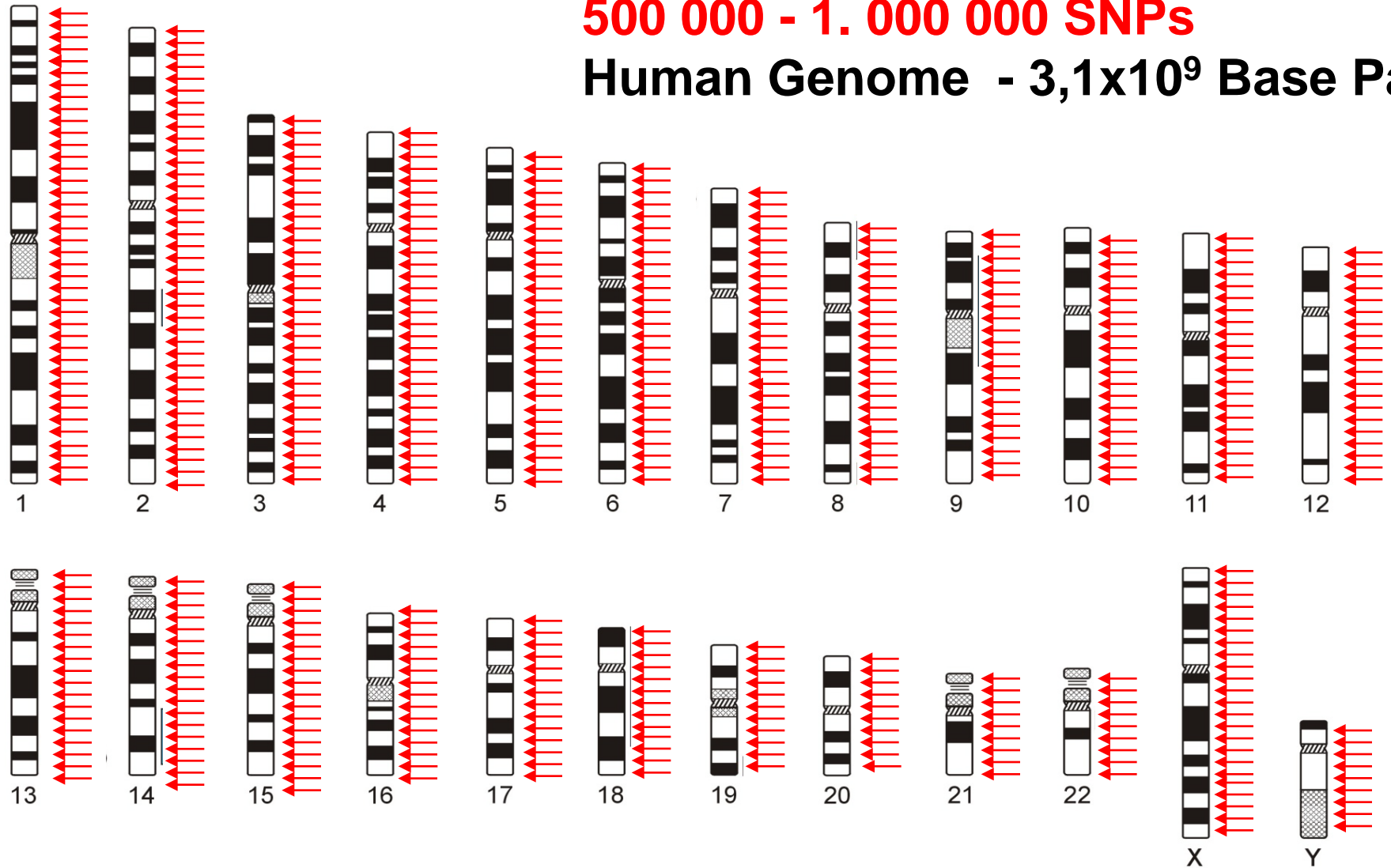
- Candidate regions
- Some successes but generally:
  - Poor replication
  - Small sample sizes
- Conclusion
  - effect sizes are smaller than expected
  - Poor selection of candidate regions



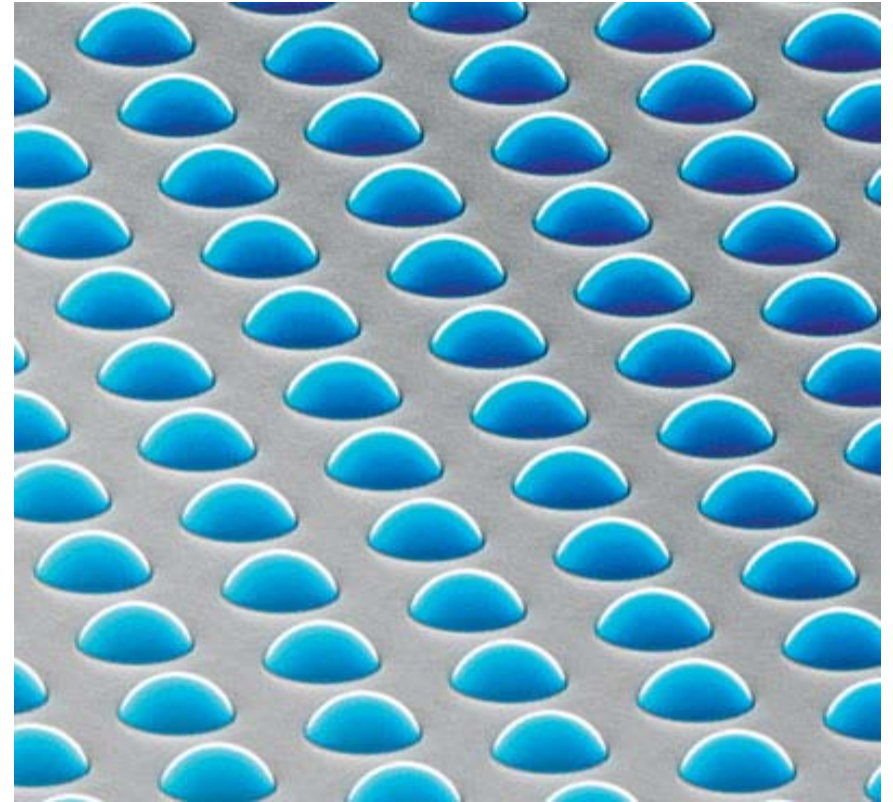
# Genome-Wide Association Studies

**500 000 - 1. 000 000 SNPs**

**Human Genome -  $3,1 \times 10^9$  Base Pairs**



# High density SNP arrays – up to 1 million SNPs

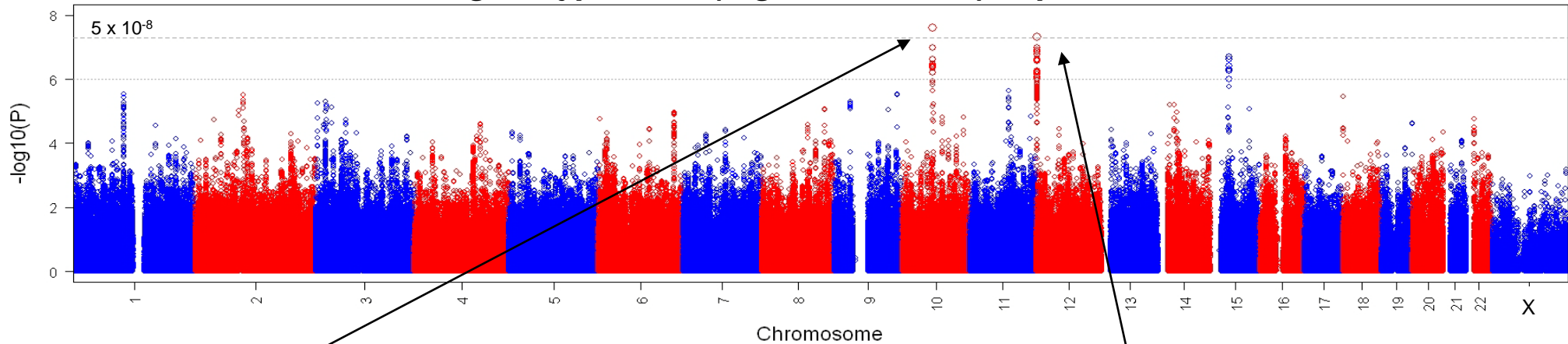


# Genome-wide Association Studies

- Multiple testing !
  - Declare significance at  $0.05/500,000 = 10^{-8}$
- Follow-up significant SNPs by genotyping in independent replication samples
- Validated associations:
  - significant across replication samples
  - same SNP
  - same allele

# Bipolar GWAS of 10,648 samples

>1.7 million genotyped and (high confidence) imputed SNPs



## *Ankryin-G (ANK3)*

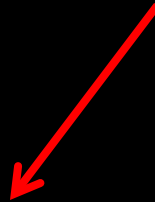
Sample	Cases	Controls	<i>P</i> -value
STEP	7.4%	5.8%	0.0013
WTCCC	7.6%	5.9%	0.0008
EXT	7.3%	4.7%	0.0002
<b>Total</b>	<b>7.5%</b>	<b>5.6%</b>	<b><math>9.1 \times 10^{-9}</math></b>

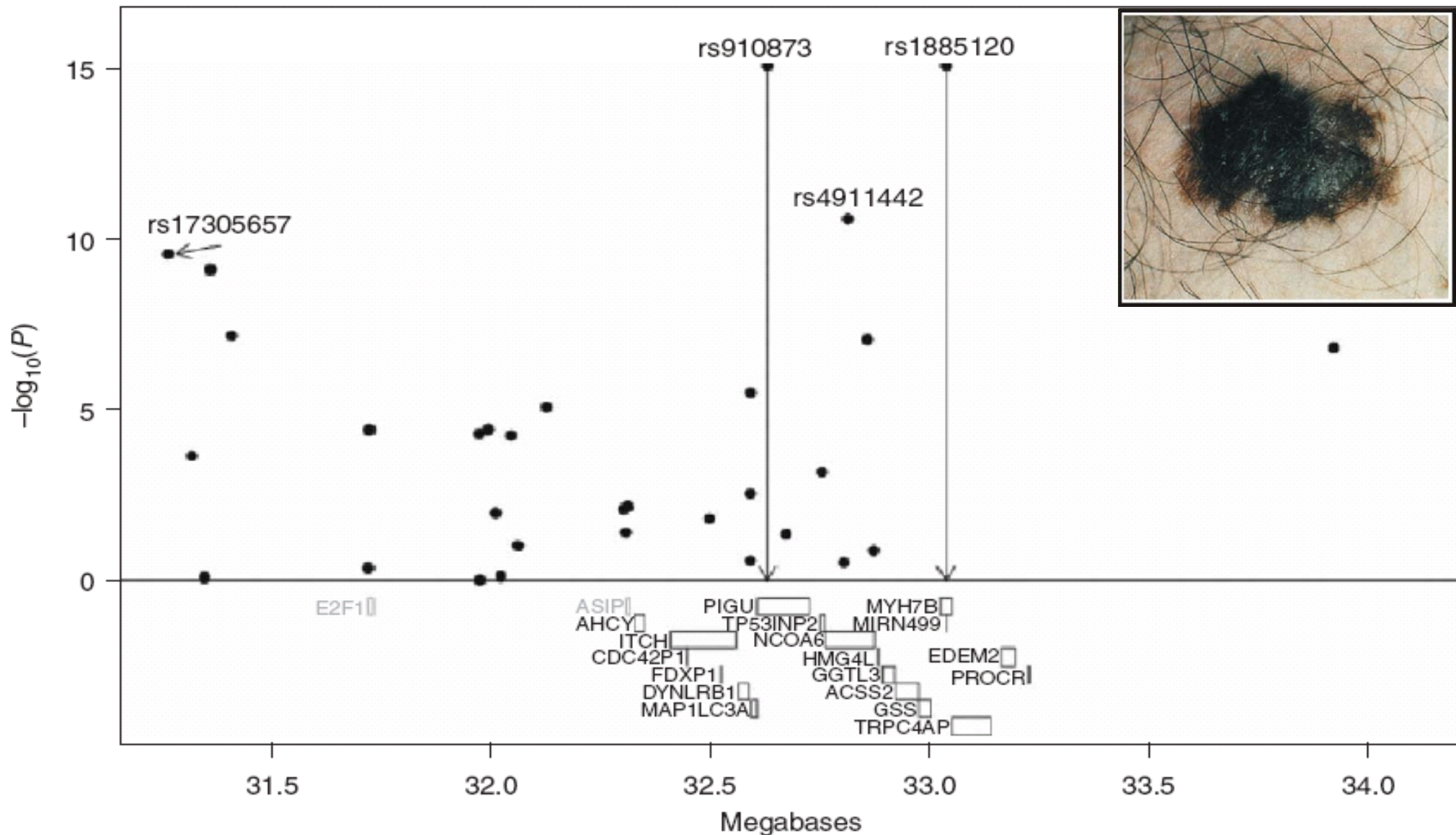
## *CACNA1C*

Sample	Case	Controls	<i>P</i> -value
STEP	35.7%	32.4%	0.0015
WTCCC	35.7%	31.5%	0.0003
EXT	35.3%	33.7%	0.0108
<b>Total</b>	<b>35.6%</b>	<b>32.4%</b>	<b><math>7 \times 10^{-8}</math></b>

Ferreira *et al* (*Nature Genetics*, 2008)

# GWAS for major depression

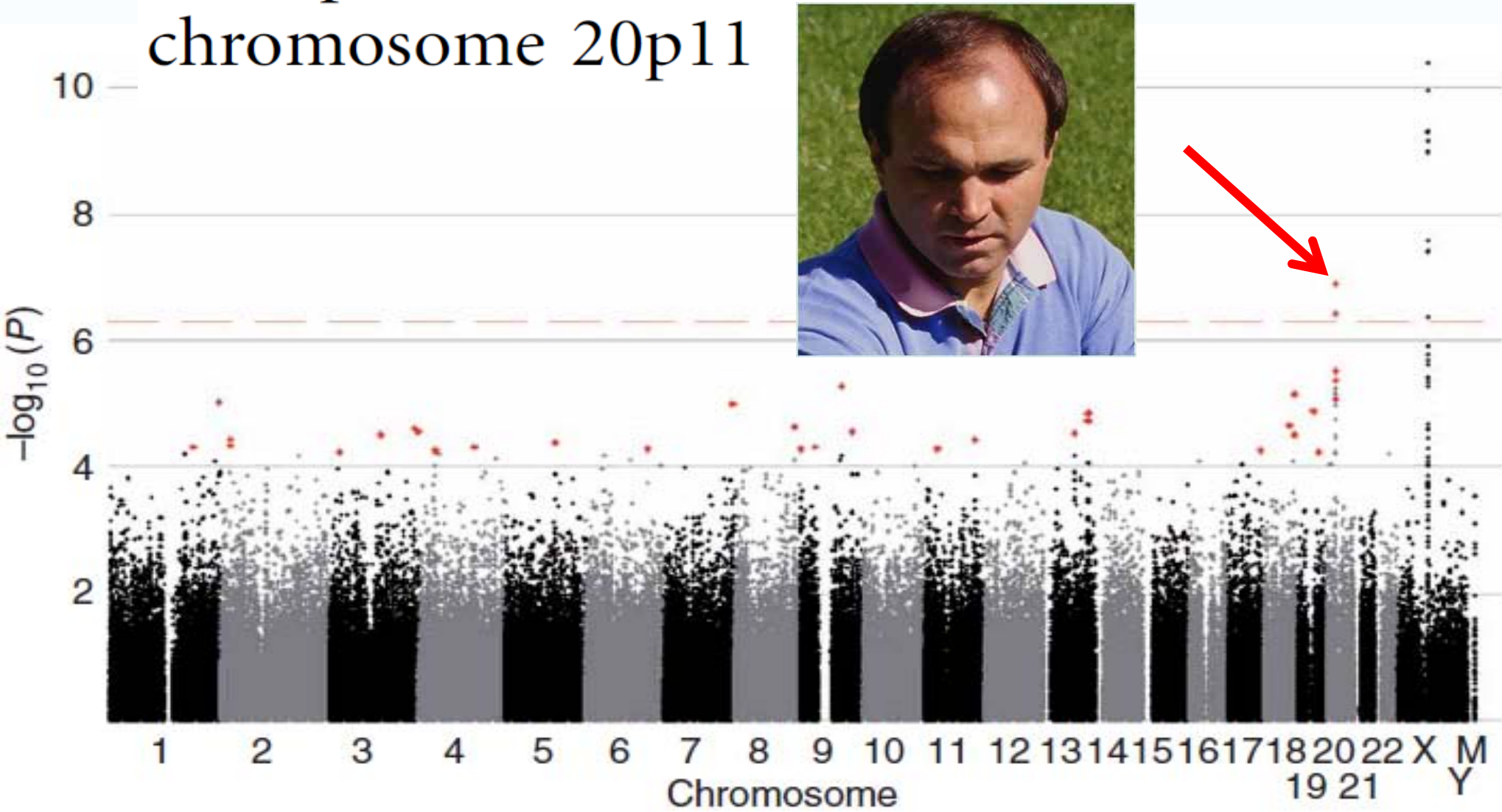




**GWAS for Melanoma** Association analysis of SNPs across a region of chromosome 20q11.22 for the combined sample. The x-axis is chromosomal position, the left y-axis  $-\log_{10}(p)$  for genotyped SNPs. *Nature Genetics* 2008 Jul;40(7):838-40.



# Susceptibility variants for male-pattern baldness on chromosome 20p11

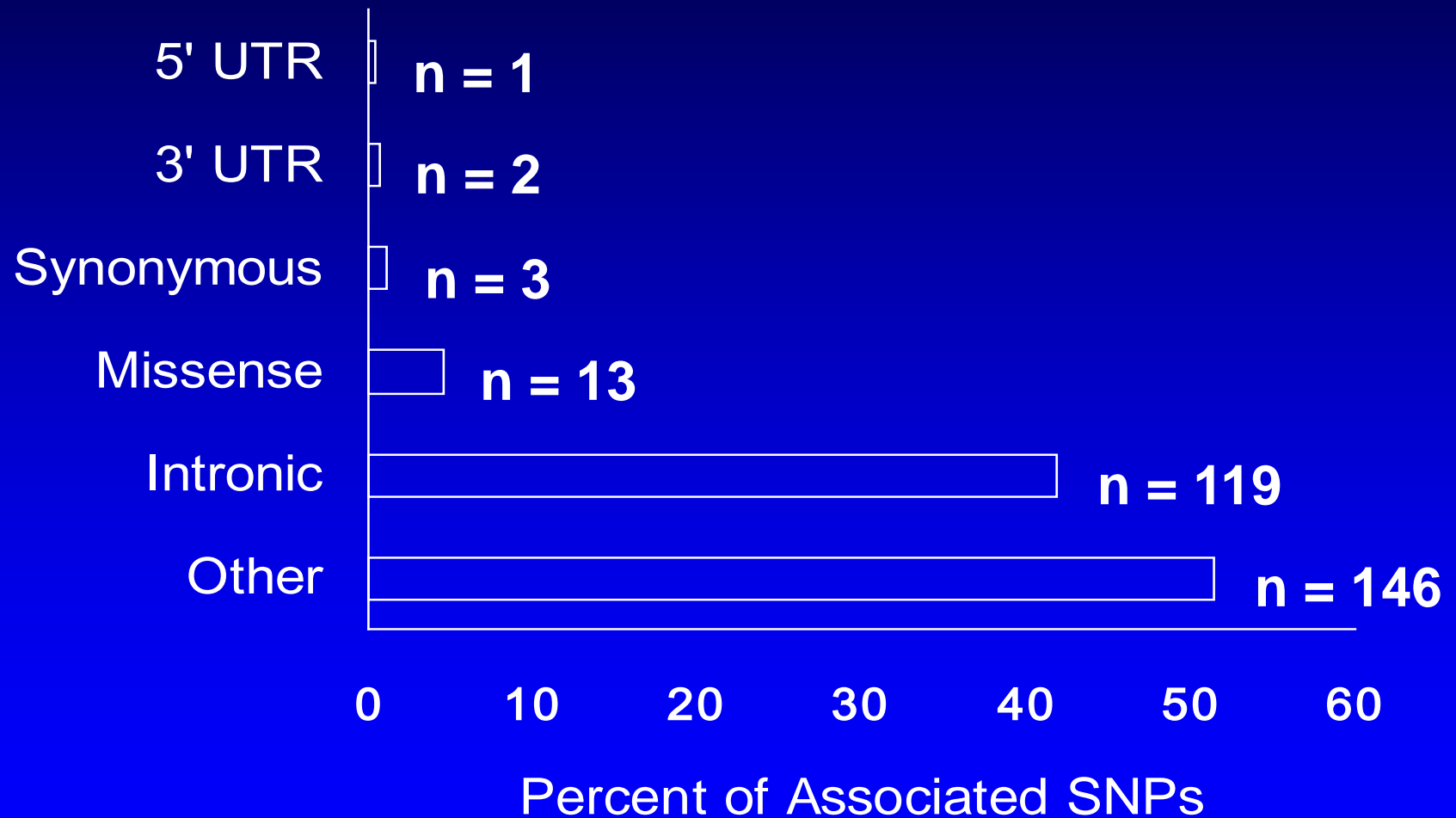


# First quarter 2008

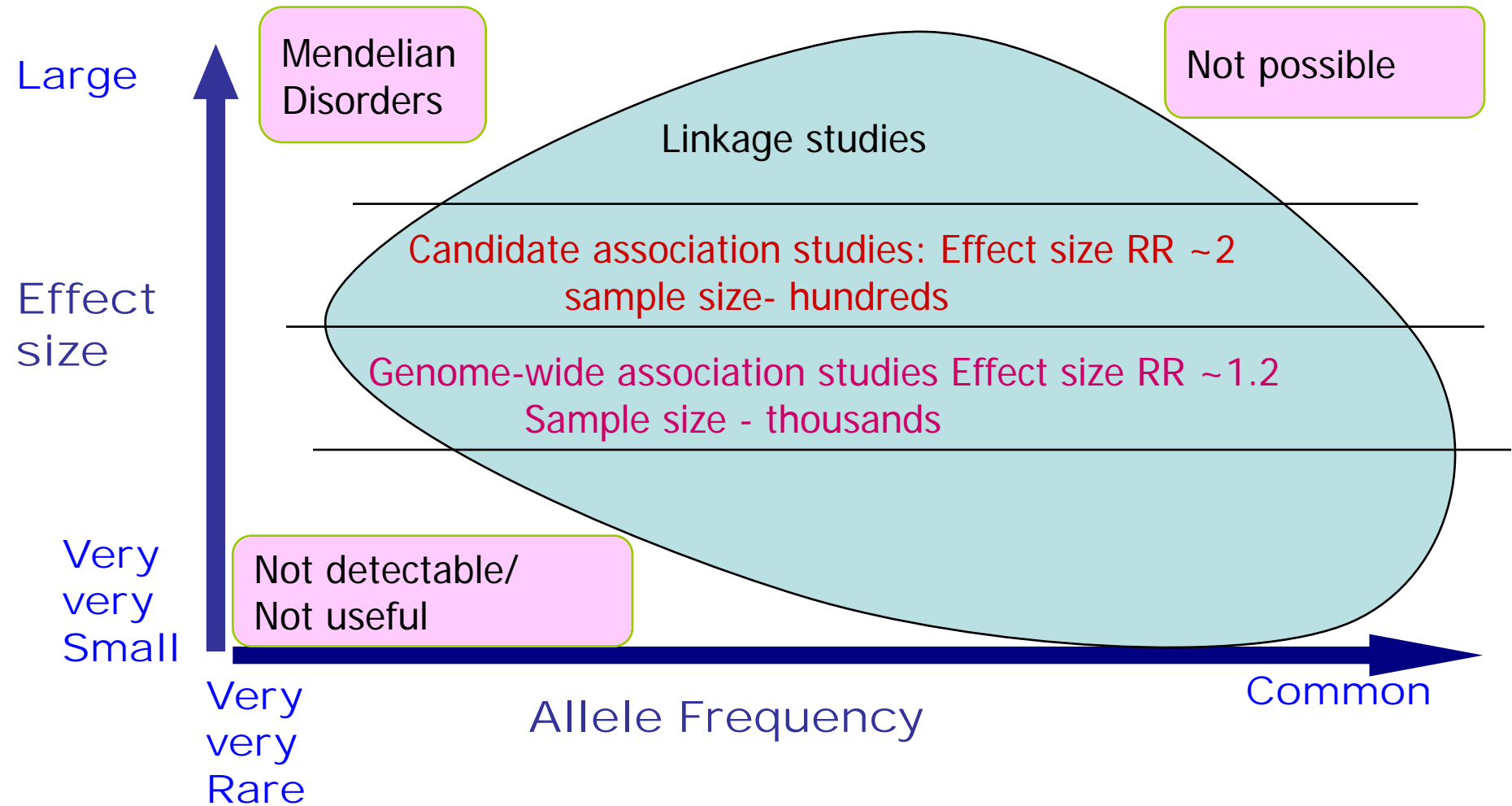




# Functional Classification of 284 SNPs Associated with Complex Traits

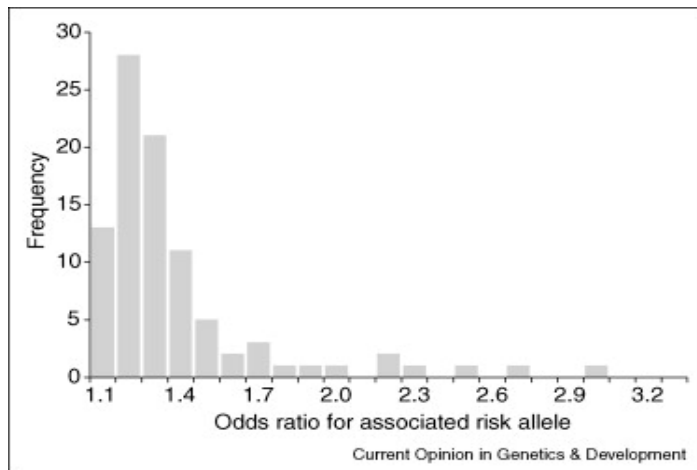


# A potted history of genetic studies



# Conclusions about genetic architecture of complex genetic traits based on GWAS results

Effects sizes of validated variants from 1st 16 GWAS studies



Genetic variation is not tagged  
by GWAS SNPs

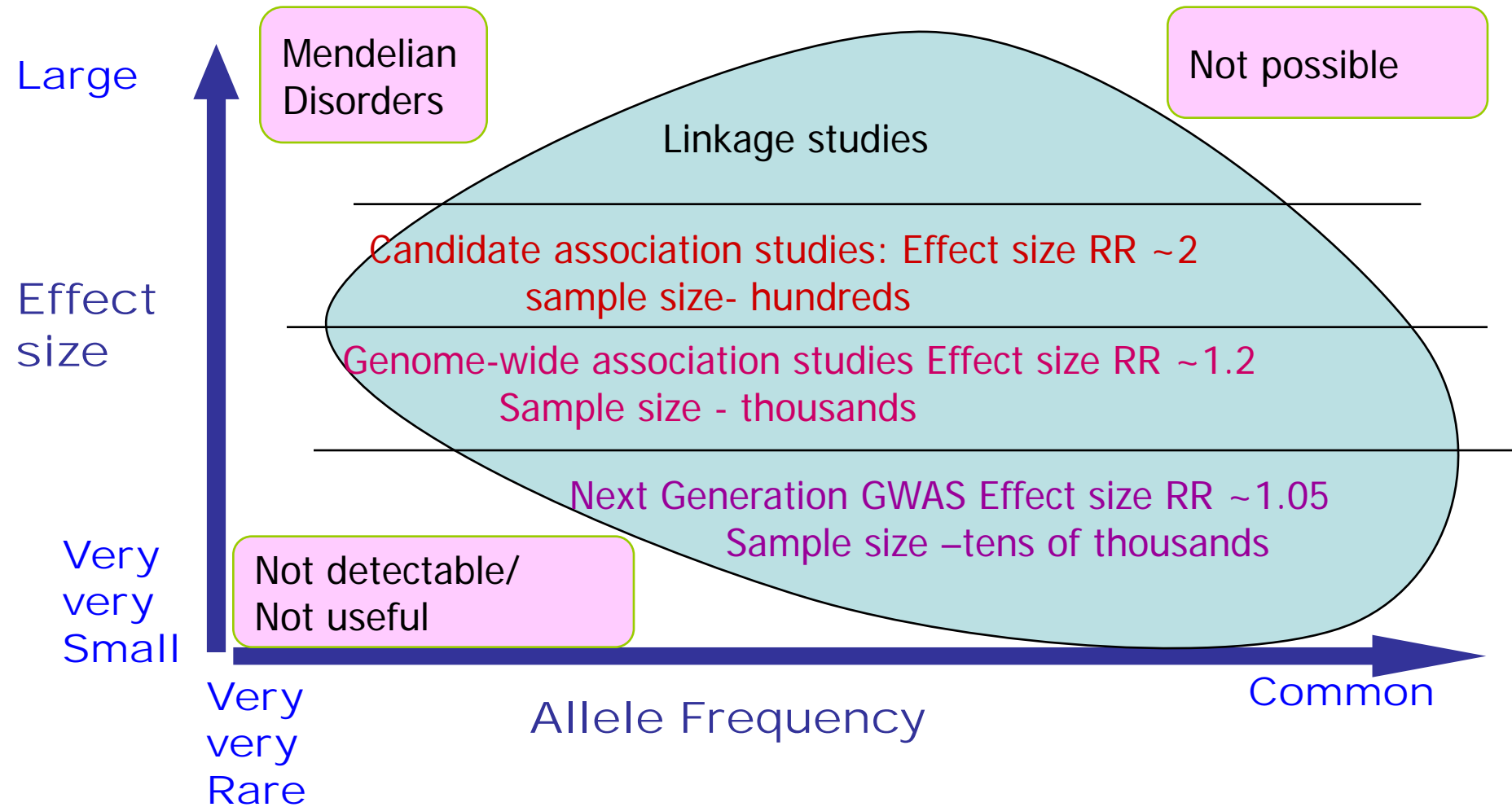
Effect sizes are smaller

**Prediction of individual genetic risk of complex disease**

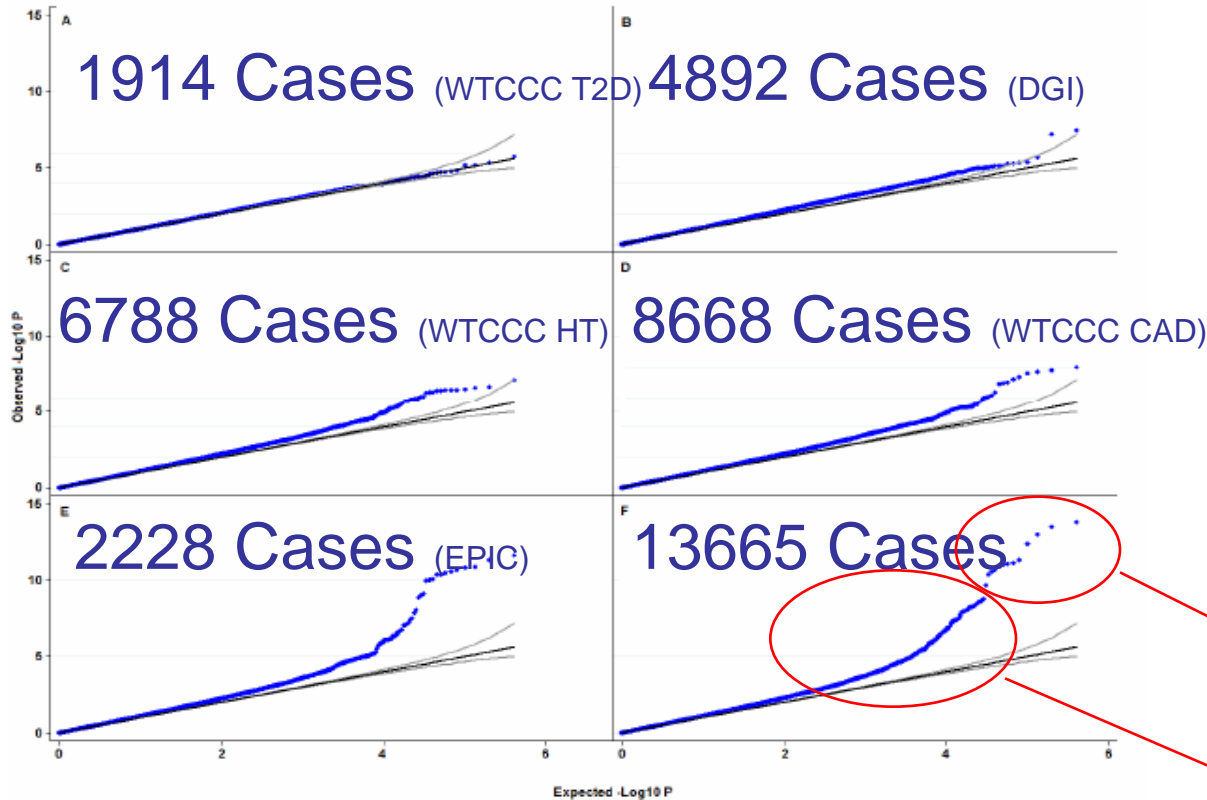
Naomi R Wray<sup>1</sup>, Michael E Goddard<sup>2</sup> and Peter M Visscher<sup>1</sup>

Current Opinion in Genetics & Development 2008, 18:257-263

# A potted history of genetic studies



# GWA of Height



Significant result

Other loci?

- ▶ Large numbers are needed to detect QTLs !!!
- ▶ Collaboration is the name of the game !!!

# Usual paradigm of association studies

- Identify significant SNPs (not many)
- Genotype in independent replication samples
- Find causal variants to inform on biology

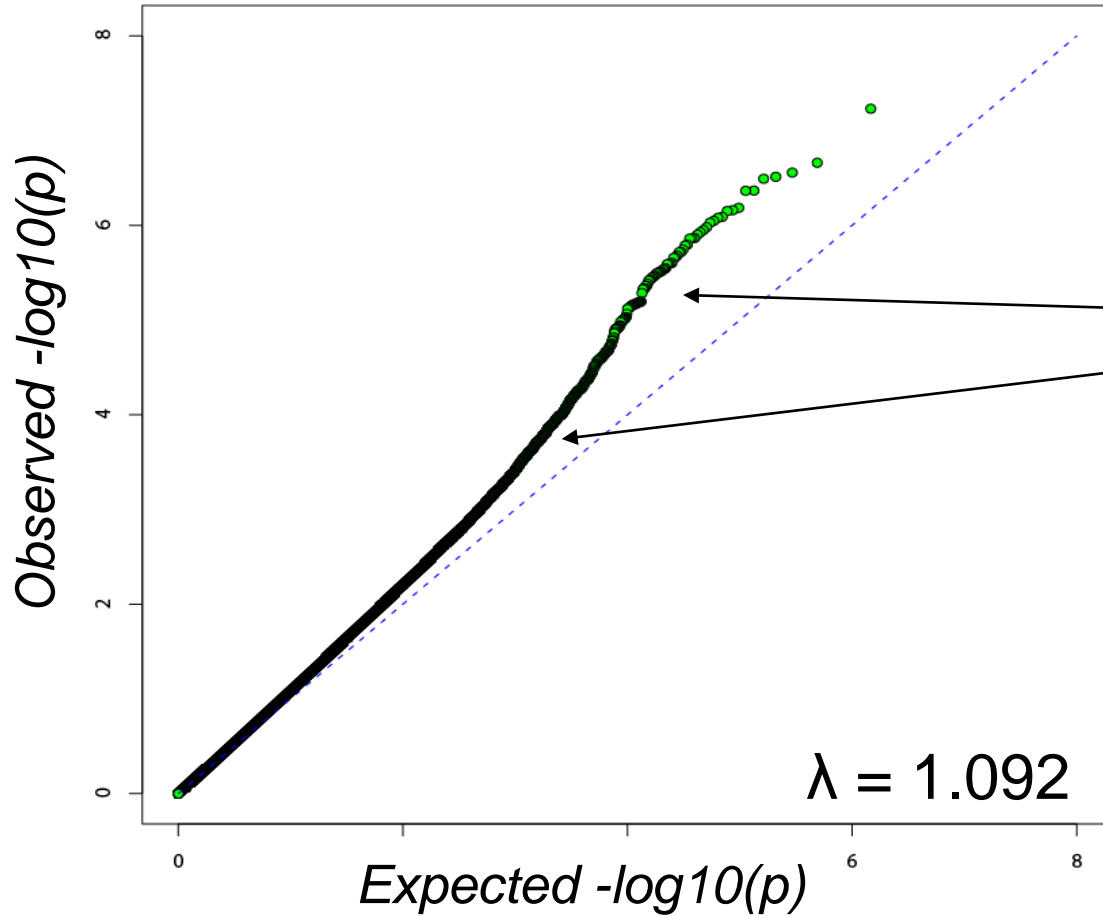
But only a small proportion of the known genetic variance is explained

What would we expect under a polygenic model of many, many variants each of small effect?

# A 5% risk increasing common variant

Relative risk	1.05
Population MAF	0.2
MAF in cases	0.2079
MAF in controls	0.1999
Power $\alpha=1\times 10^{-6}$ ( $N=10,000$ cases, $10,000$ controls)	0.2%
Proportion ranked in top half ( $N=ISC$ )	72%

# Schizophrenia (ISC) Q-Q plot



**Consistent with:**

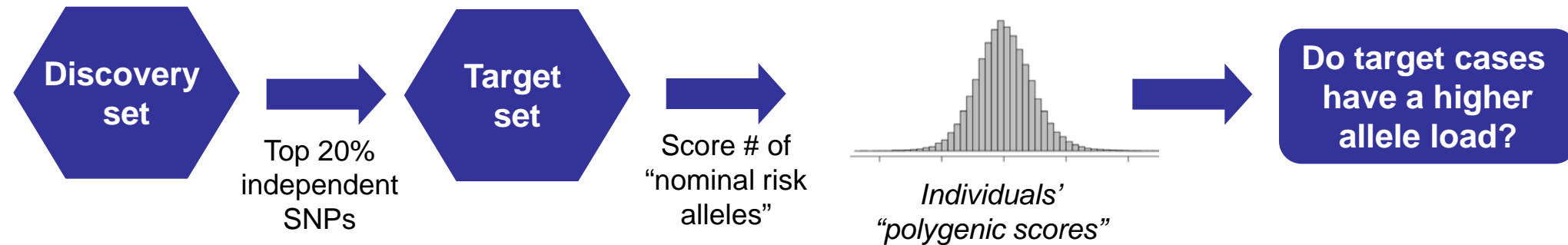
*Stratification?*

*Genotyping bias?*

*Distribution of true polygenic effects?*



# Indexing polygenic variance with large sets of weakly associated alleles



ISC

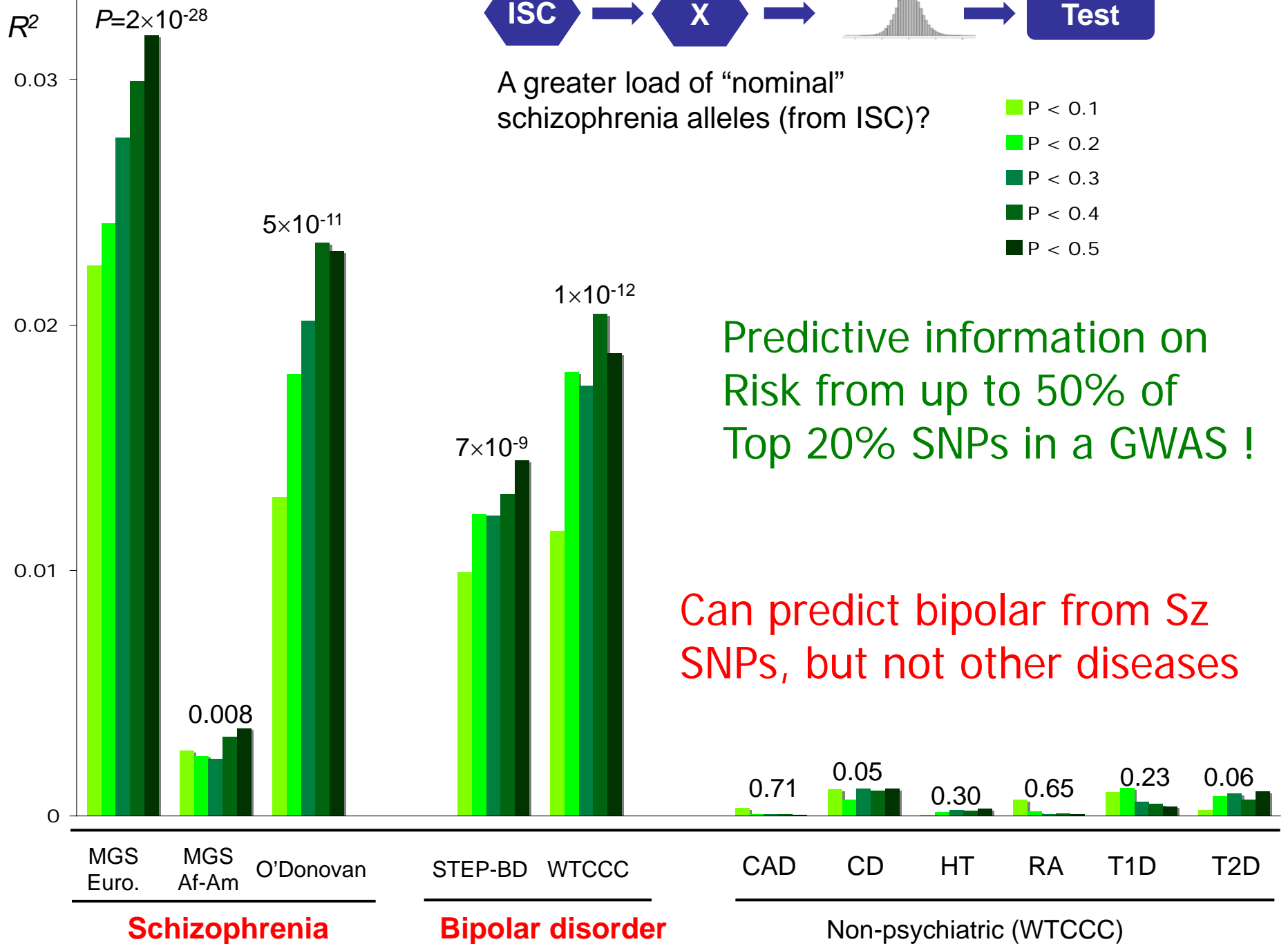
- ISC
- Independent SCZ studies (MGS, O'Donovan)
- Bipolar disorder (STEP-BD, WTCCC)
- Non-psychiatric disease (WTCCC)

**Douglas Levinson, Pablo Gejman,**  
Jianxin Shi and colleagues



A greater load of “nominal” schizophrenia alleles (from ISC)?

- P < 0.1
- P < 0.2
- P < 0.3
- P < 0.4
- P < 0.5



Predictive information on Risk from up to 50% of Top 20% SNPs in a GWAS !

Can predict bipolar from Sz SNPs, but not other diseases

$P=2 \times 10^{-28}$

$5 \times 10^{-11}$

$1 \times 10^{-12}$

$7 \times 10^{-9}$

0.008

0.71

0.05

0.30

0.65

0.23

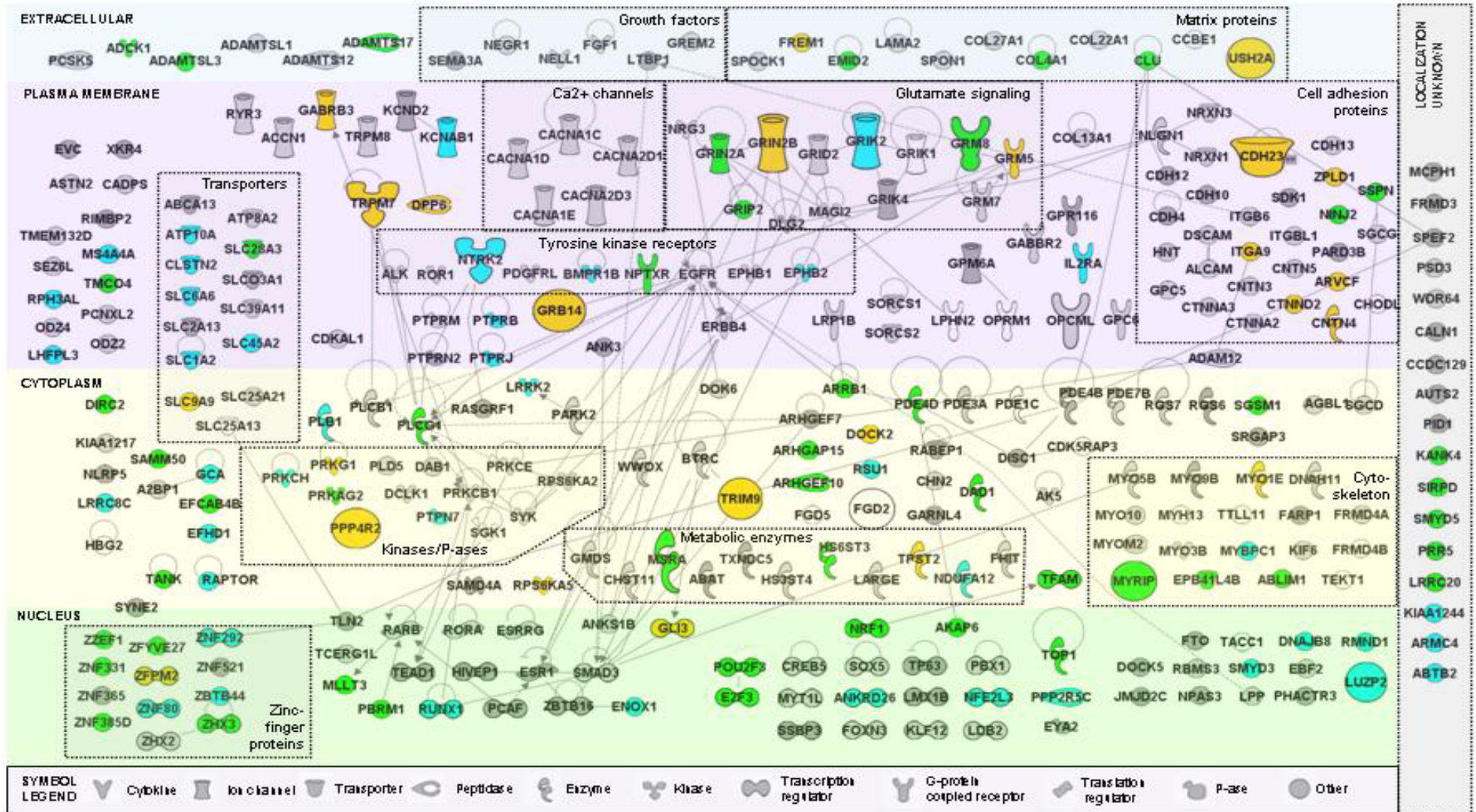
0.06

Schizophrenia

Bipolar disorder

Non-psychiatric (WTCCC)

# Pathway (Ingenuity) analysis of GWAS for smoking





# Even for “simple” diseases the number of alleles is large

---

- Ischaemic heart disease (LDR) >190
- Breast cancer (BRAC1) >300
- Colorectal cancer (MLN1) >140

# Multiple Rare Alleles Contribute to Low Plasma Levels of HDL Cholesterol

Jonathan C. Cohen,<sup>1,2,3,\*</sup> Robert S. Kiss,<sup>5,\*</sup>  
 Alexander Pertsemlidis,<sup>1</sup> Yves L. Marcel,<sup>5,†</sup> Ruth McPherson,<sup>5</sup>  
 Helen H. Hobbs<sup>1,3,4</sup>

Heritable variation in complex traits is generally considered to be conferred by common DNA sequence polymorphisms. We tested whether rare DNA sequence variants collectively contribute to variation in plasma levels of high-density lipoprotein cholesterol (HDL-C). We sequenced three candidate genes (*ABCA1*, *APOA1*, and *LCAT*) that cause Mendelian forms of low HDL-C levels in individuals from a population-based study. Nonsynonymous sequence variants were significantly more common (16% versus 2%) in individuals with low HDL-C (<fifth percentile) than in those with high HDL-C (>95th percentile). Similar findings were obtained in an independent population, and biochemical studies indicated that most sequence variants in the low HDL-C group were functionally important. Thus, rare alleles with major phenotypic effects contribute significantly to low plasma HDL-C levels in the general population.

Complex disease: common or rare alleles?

Increasing evidence for  
 Common Disease – Rare Variant  
 hypothesis (CDRV)

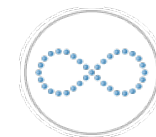
**Table 1.** Sequence variations in the coding regions of *ABCA1*, *APOA1*, and *LCAT*. Values represent the numbers of sequence variants identified in 256 individuals from the Dallas Heart Study (DHS) (128 with low HDL-C and 128 with high HDL-C) and 263 Canadians (155 with low HDL-C and 108 with high HDL-C) (17). NS, nonsynonymous (nucleotide substitutions resulting in an amino acid change); S, synonymous (coding sequence substitutions that do not result in an amino acid change). GenBank accession numbers for DHS *ABCA1*, *APOA1*, and *LCAT* sequences are NM\_005502, NM\_000039, and NM\_000229, respectively.

	Sequence variants unique to one group				Sequence variants common to both groups	
	Low HDL-C		High HDL-C		NS	S
	NS	S	NS	S		
	DHS					
<i>ABCA1</i>	14	6	2	5	10	19
<i>APOA1</i>	1	0	0	1	0	1
<i>LCAT</i>	0	1	1	0	1	1
	Canadians					
<i>ABCA1</i>	14	2	2	3	7	5
<i>APOA1</i>	0	1	0	0	2	0
<i>LCAT</i>	6	1	0	0	0	0

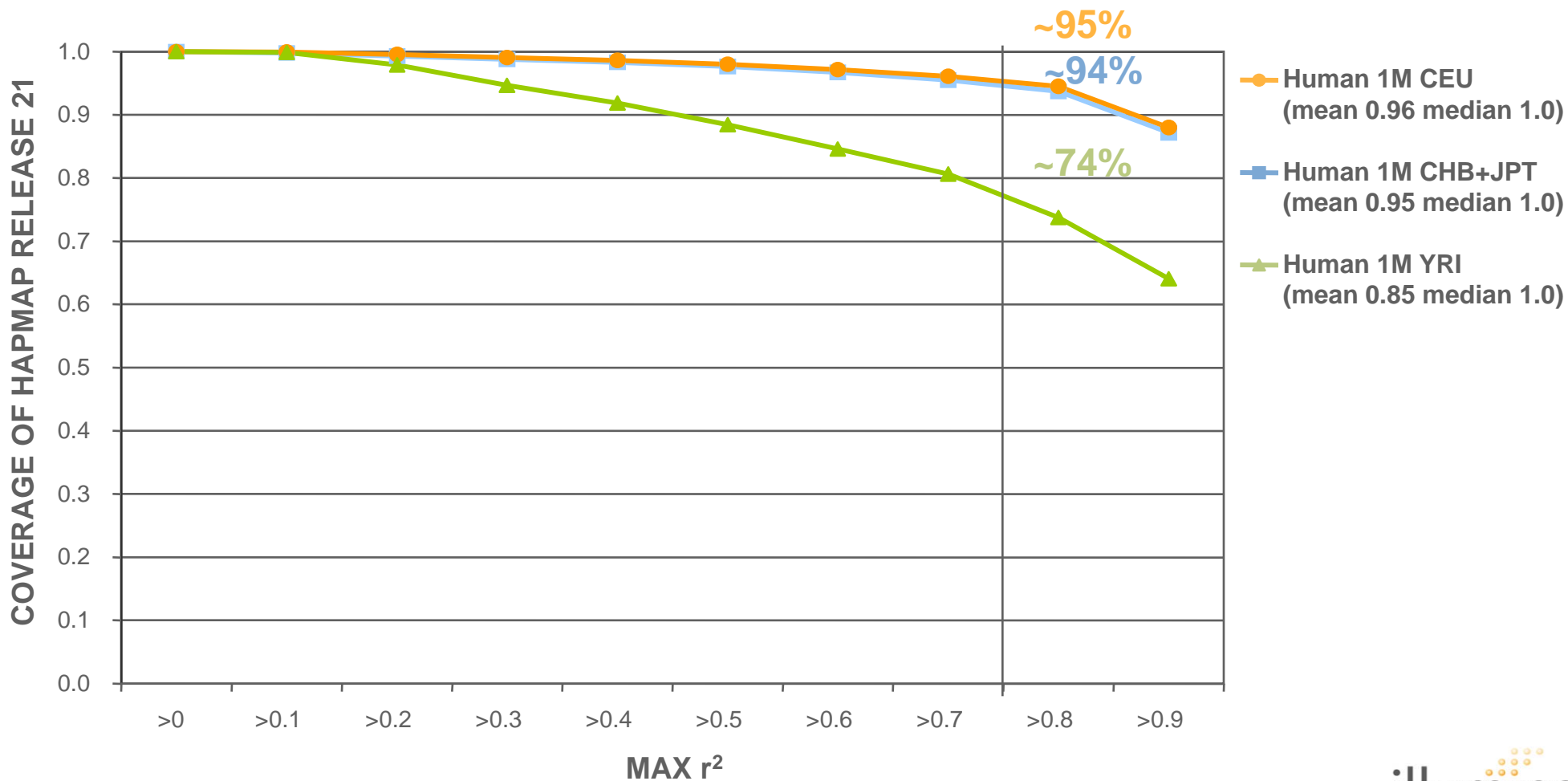
[Science 2004]



# Human 1M HapMap Coverage by Population



GENOME COVERAGE ESTIMATED FROM 990,000 HAPMAP SNPs IN HUMAN 1M





## products & services

- overview
- systems & software
- dna analysis solutions
- rna analysis solutions
- solexa applications
- services
- product literature

- print this page

## solexa sequencing applications

Illumina's Solexa Sequencing technology offers a powerful new approach to some of today's most important applications for genetic analysis and functional genomics, including:

### sequencing and resequencing

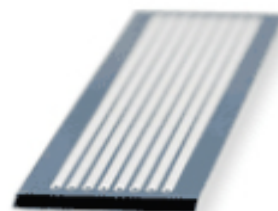
Whether you need to sequence an entire genome or a large candidate region, the Illumina Genome Analyzer System is today's most productive and economical sequencing tool. Solexa sequencing technology and reversible terminator chemistry deliver unprecedented volumes of high quality data, rapidly and economically.

### expression profiling

Sequencing millions of short cDNA tags per sample, the Genome Analyzer allows you to generate digital expression profiles at costs comparable to current analog methods. Because our protocol does not require any transcript-specific probes, you can apply the technology to discover and quantitate transcripts in any organisms, irrespective of the annotation available on the organism.

### small rna identification and quantification

Solexa sequencing technology also offers a unique and powerful solution for the comprehensive discovery and characterization of small RNAs in a wide range of species. The massively parallel sequencing protocol allows researchers to discover and analyze genome-wide profiles of small RNA in any species. With the potential to generate several million sequence tags economically, the Illumina Genome Analyzer offers investigators the opportunity to uncover global profiles of small RNA at an unprecedented scale.

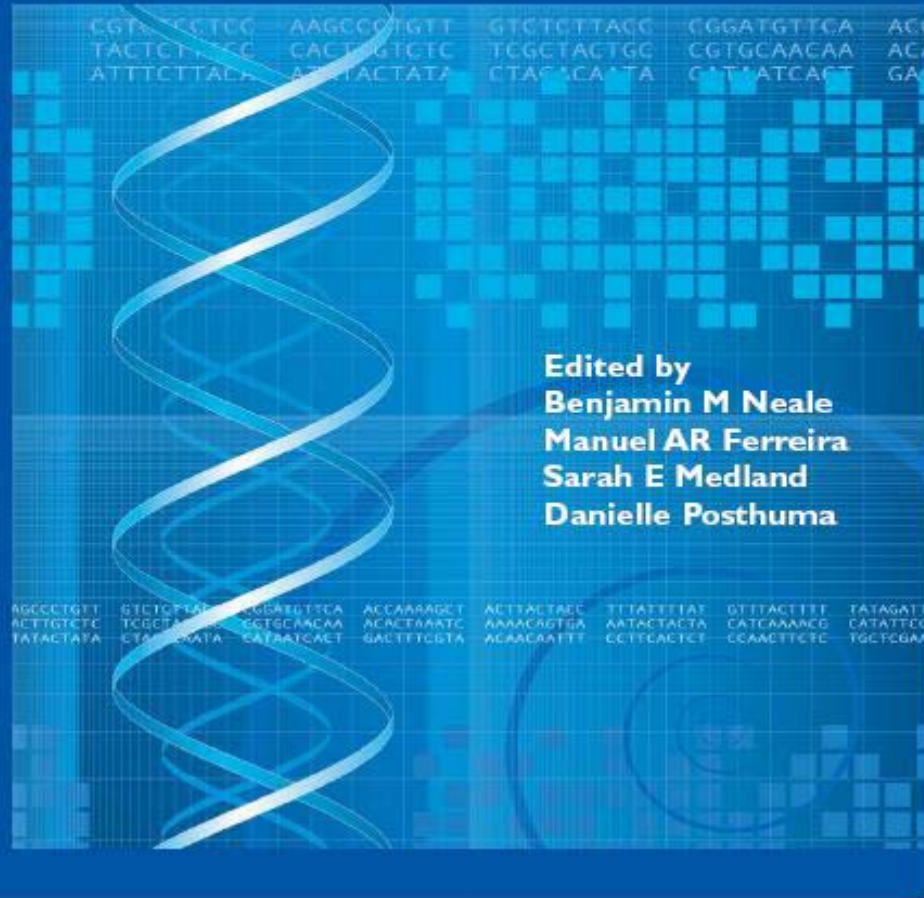


## important information

- product literature
- publications
- faqs
- have a rep contact me

# STATISTICAL GENETICS

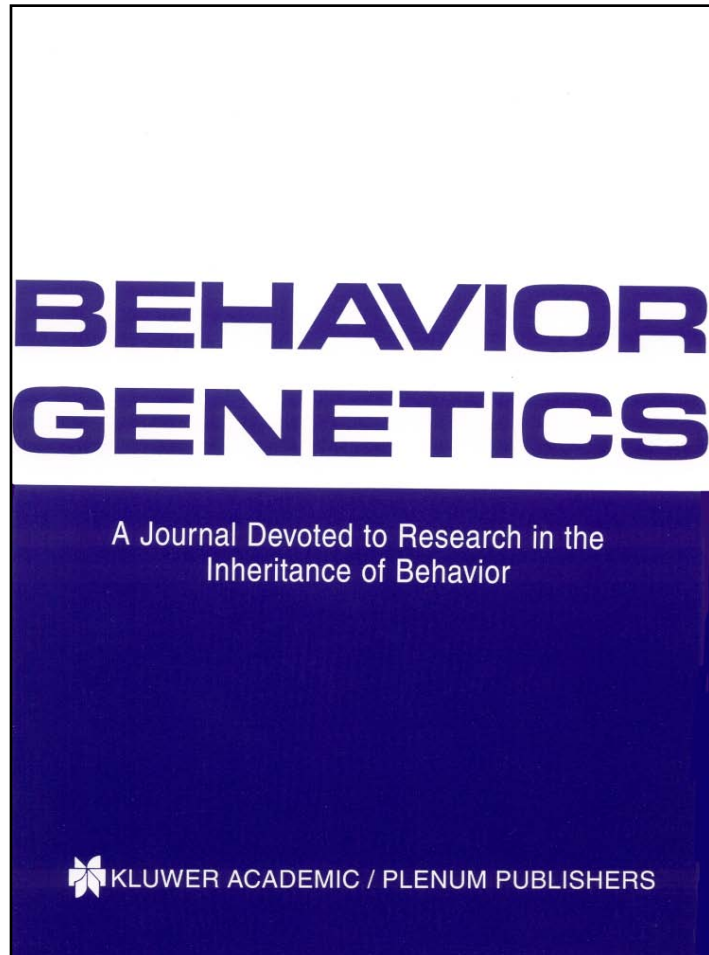
Gene Mapping Through Linkage  
and Association



<http://www.genemapping.org/>



# We also run two journals (1)



- Editor: John Hewitt
- Editorial assistant  
Christina Hewitt
- Publisher: Kluwer  
/Plenum
- Fully online
- <http://www.bga.org>

# We also run two journals (2)



- Editor: Nick Martin
- Editorial assistant + subscriptions: Lorin Grey
- Publisher: Australian Academic Press
- Fully online
- <http://www.ists.qimr.edu.au/journal.html>