

Genome-wide association studies

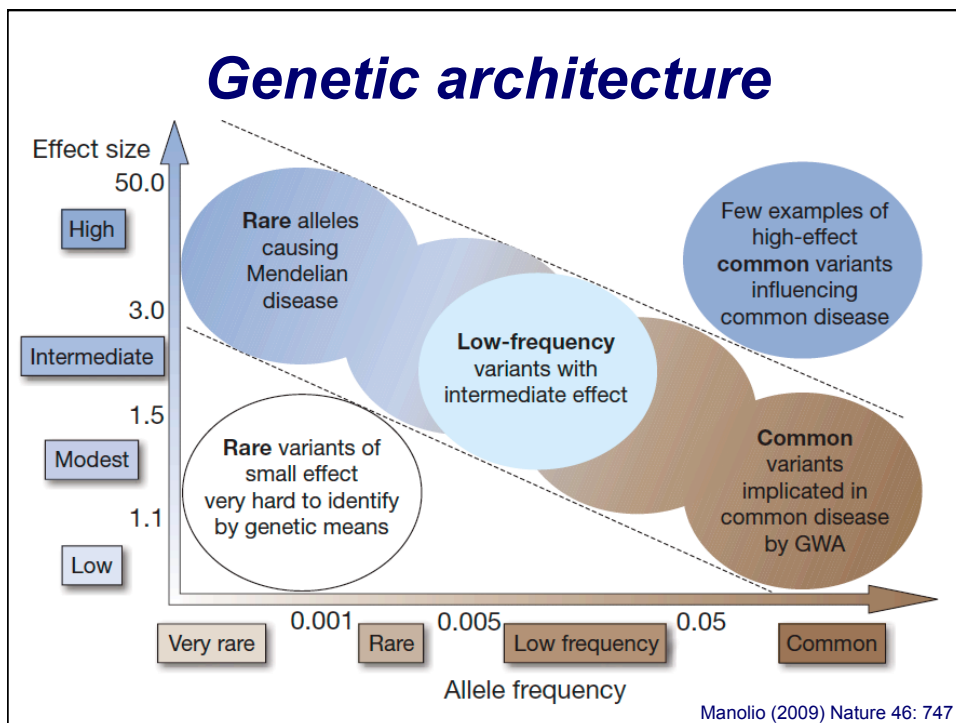
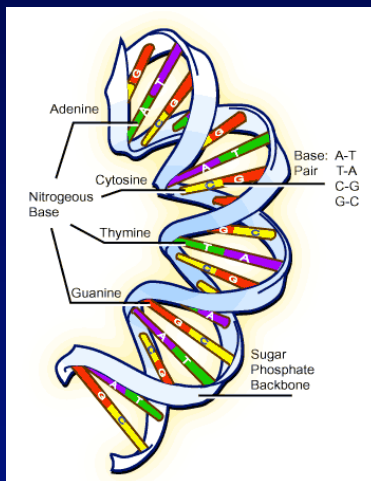
Karen Mohlke, PhD
Department of Genetics
University of North Carolina

Complex traits



Common and rare variants

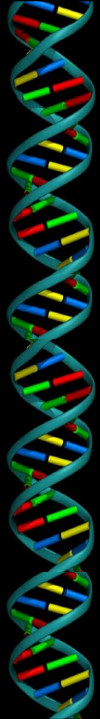
GGATTCAC**T**GCAAATCG
 GGATTCAC**T**GCAAATCG
 GGATTCACAGCAAATCG
 GGATTCAC**T**GCAAATCG
 GGATTCAC**T**GCAAATCG
 GGATTCAC**T**GCAAATCG
 GGATTCAC**T**GCAAAT**G**
 GGATTCACAGCAAATCG
 GGATTCACAGCAAATCG
 GGATTCAC**T**GCAAATCG



Genome-wide association (GWA)

- What is the goal?
- How are studies performed?
- What can we learn from the associated regions?
- What do the findings tell us about disease?

GWA Studies

- 
- Benefits of GWA vs classical mapping
 - More powerful vs linkage for common, low penetrance variants
 - Better resolution than linkage
 - No need to select candidate genes
 - Requirements of GWA
 - Catalog of human genetic variants
 - Low cost, accurate method for genotyping
 - Large number of informative samples
 - Efficient statistical design and analysis

Goals of a GWA study

- **Test a large portion of the common single nucleotide genetic variation in the genome for association with a disease or variation in a quantitative trait**
- **Find disease/quantitative trait-related variants without a prior hypothesis of gene function**

Steps in a GWA study

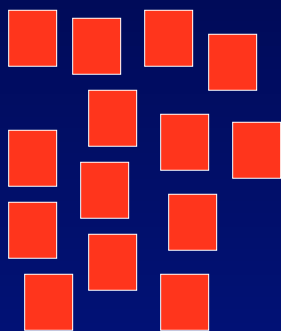
- **Samples**
- **Genotyping**
- **Quality control**
- **Statistical analysis**
- **Replication**

Phenotype

- **Disease (case/control)**
 - Rare
 - Common
- **Quantitative trait**
 - Easy to measure: Weight, height
 - Requires testing: Coronary artery thickness
 - Requires experiment: Gene expression

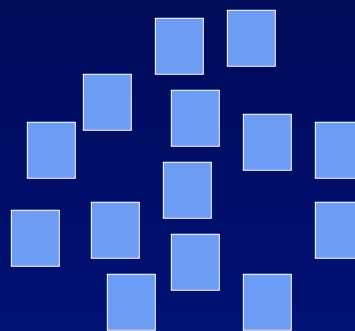
Selection of cases and controls

Cases



Definition of case?

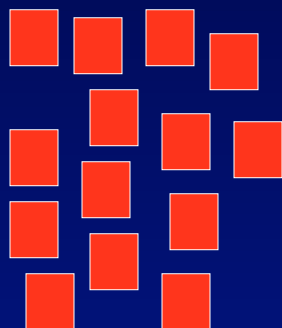
Controls



Definition of control?

Selection of cases

Cases



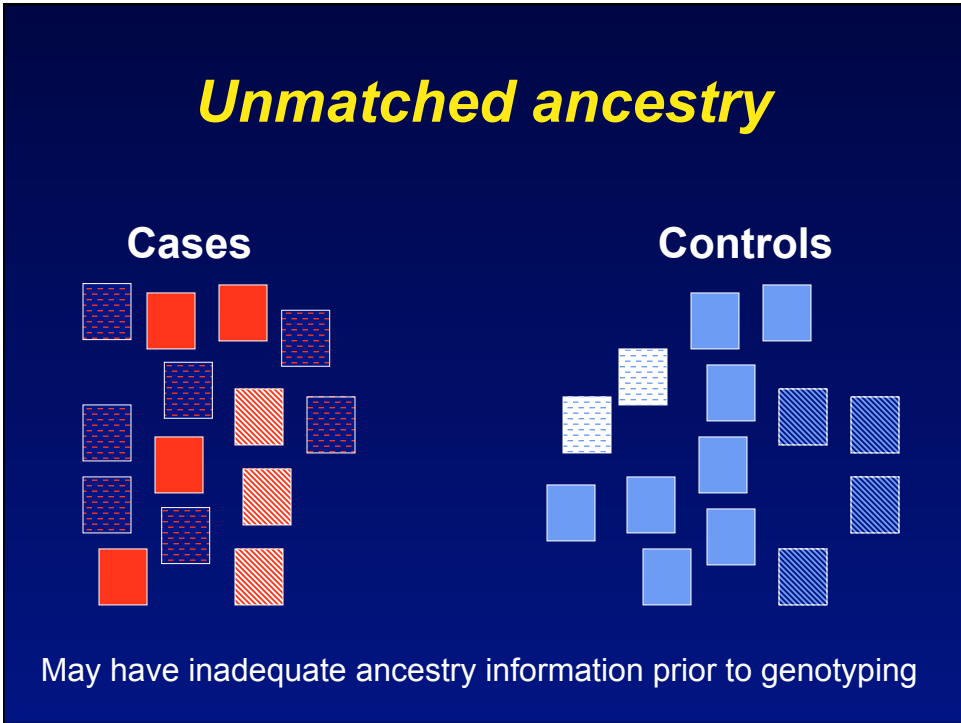
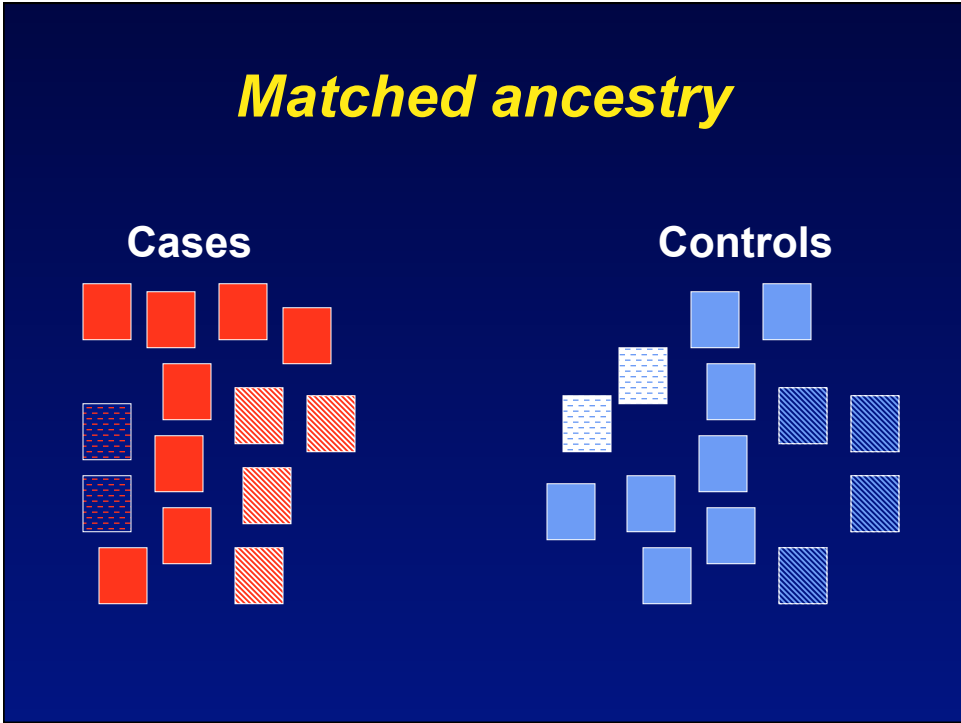
- Potential criteria to enrich genetic effect size
 - More severely affected individuals
 - Require other family member to have disease
 - Younger age-of-disease onset

Selection of controls

- Potential criteria to enrich genetic effect size
 - Low risk of disease rather than population-based samples
 - Same ancestry as cases
 - Matched to cases on age, sex, demographics

Controls



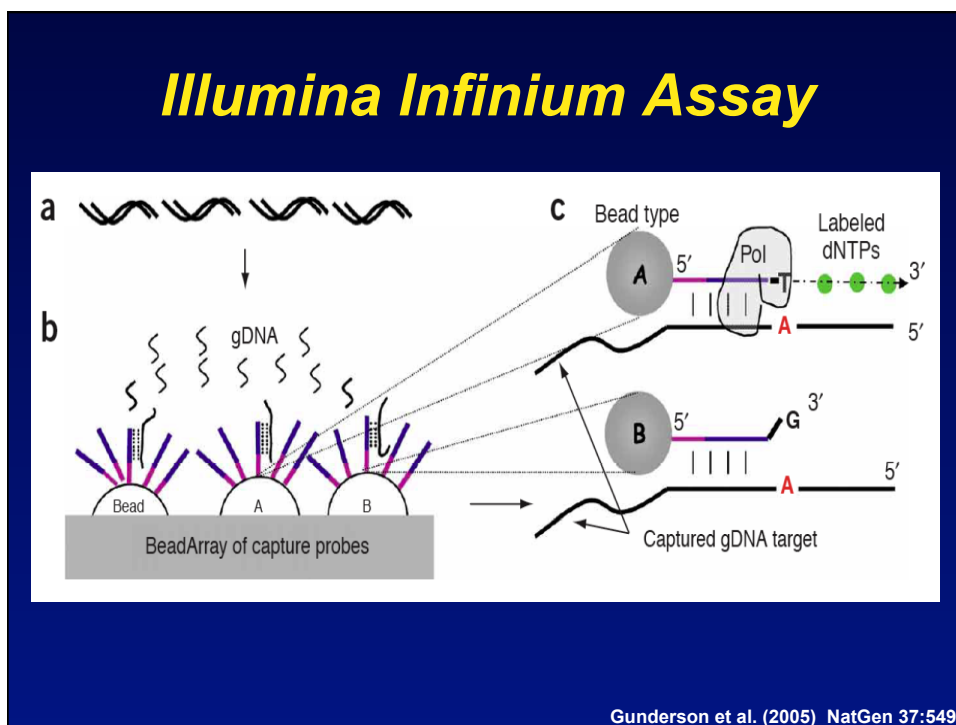
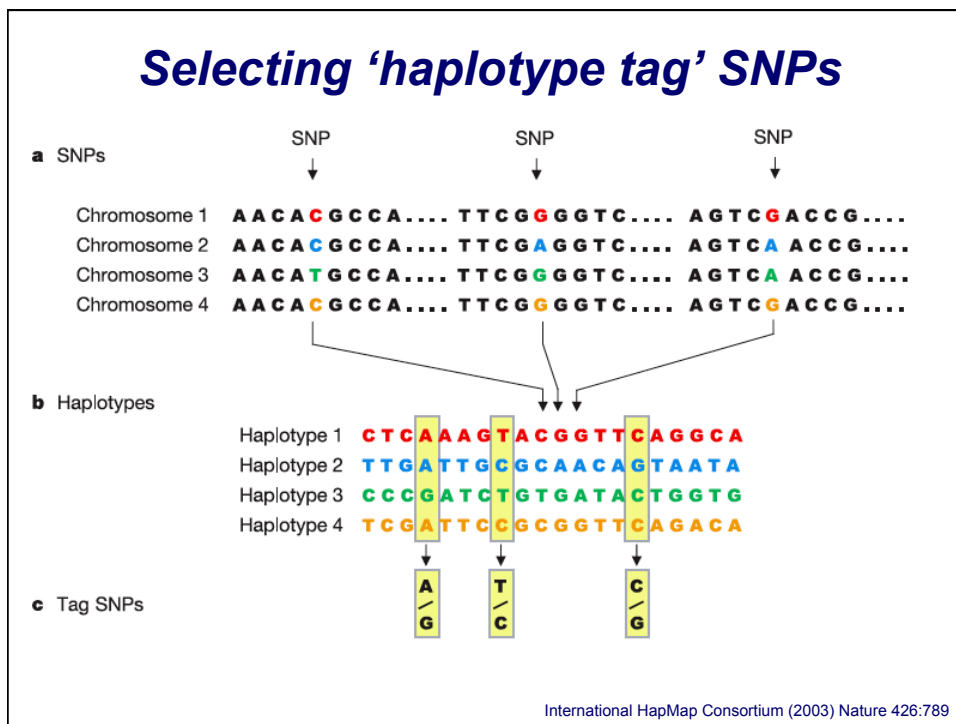


Population stratification and cryptic relatedness

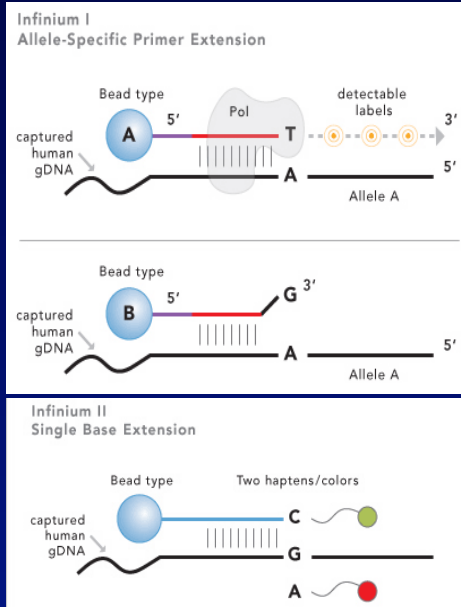
- **Can produce spurious associations in case-control studies**
- **Account for or avoid**
 - **Genomic control**
 - **Principle components**
 - **Family-based study design**

Genome-wide SNP panels

- **10,000 - 1+ million SNPs**
- **Affymetrix, Illumina**
 - **Random SNPs**
 - **Selected haplotype tag SNPs**
 - **Copy number probes**



ILLUMINA Infinium Assays



Infinium Assay BeadChips enable interrogation of ~317,000 to over one million SNPs and offer comprehensive coverage of CNV regions. Shown above, from left to right, are the Human100K-Duo, Human550K-Duo, and Human1M BeadChips.

illumina.com

Affymetrix GeneChip Array

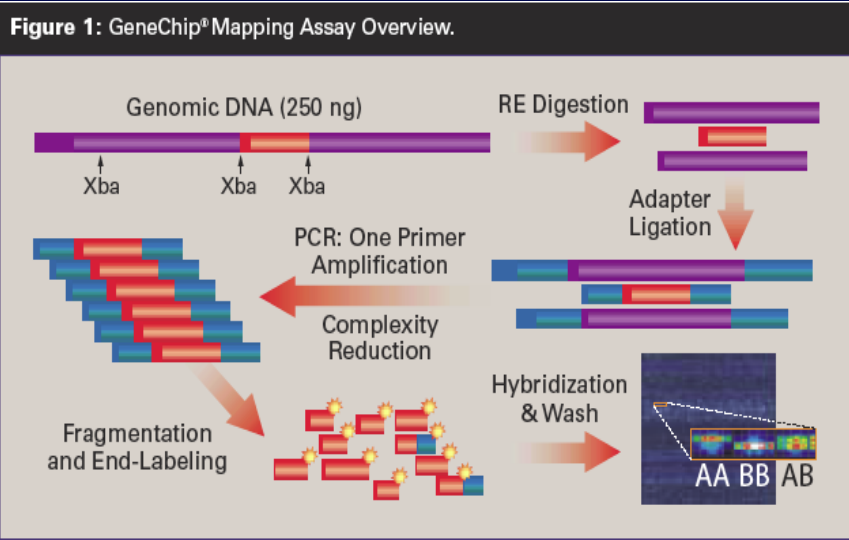
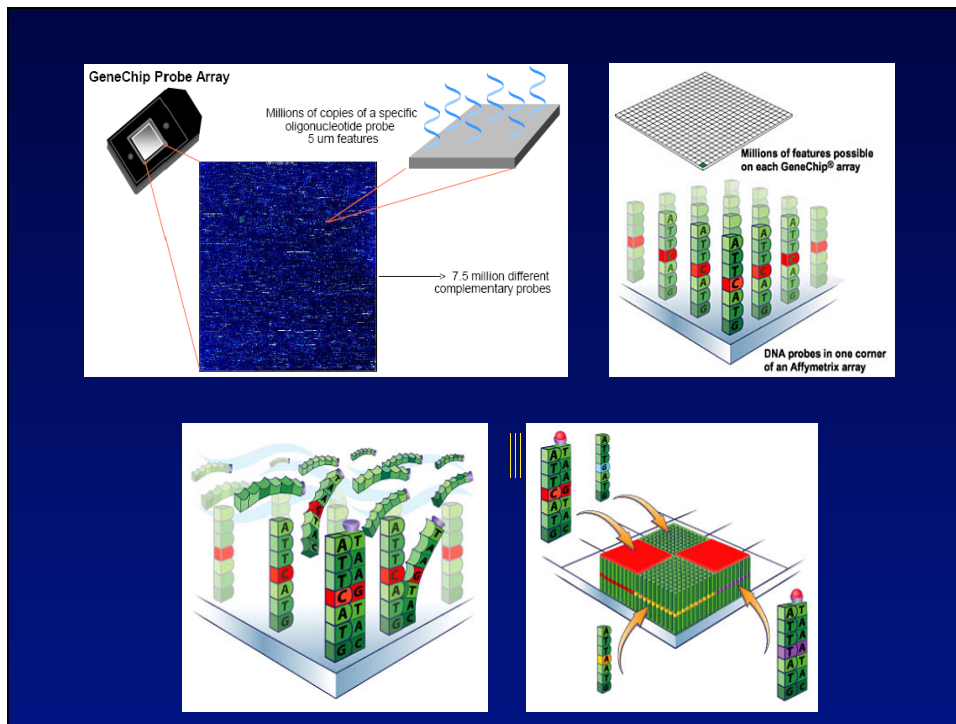


image from affymetrix.com



Global genomic coverage

Table 1 Global coverage (%) by SNP chips

SNP chip	CEU	CHB+JPT	YRI
SNP Array 5.0	64	66	41
SNP Array 6.0	83	84	62
HumanHap300	77	66	29
HumanHap550	87	83	50
HumanHap650Y	87	84	60
Human1M	93	92	68

Li (2008) EJHG 16:625

Local genomic coverage

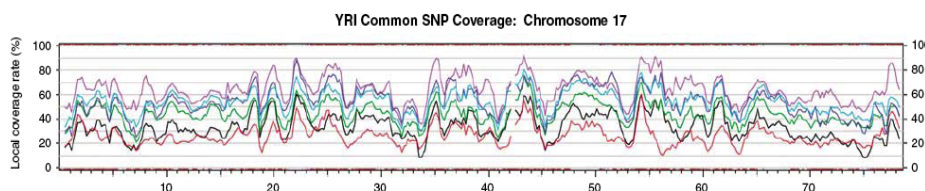


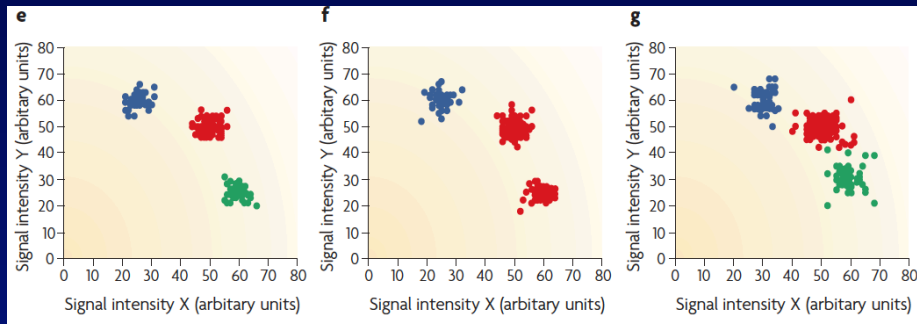
Figure 1 Local coverage map for each HapMap population for chromosome 17. The six SNP chips that were evaluated are SNP Array 5.0 (black), SNP Array 6.0 (blue), HumanHap300 (red), HumanHap550 (green), HumanHap650Y (cyan), and Human1M (purple). The red bars at the top and bottom indicate the transcription regions of known protein coding genes.

Li (2008) EJHG 16:625

Quality control: Identify and remove bad samples

- **Poor quality samples**
 - Sample success rate < 95 %
 - Excess heterozygous genotypes
- **Sample switches**
 - Wrong sex
- **Unexpected related individuals**
 - Pair-wise comparisons of genotype similarity
 - Duplicates
- **Ancestry different from the rest of sample**

Quality control: Identify and remove bad SNPs



Ideal genotyping plot

Clusters mis-called

Clusters overlap

McCarthy (2008) Nat Rev Gen 9:356

Quality control: Identify and remove bad SNPs

- Genotyping success rate < 95%
- Different genotypes in duplicate samples
- Expected proportions of genotypes are not consistent with observed allele frequencies
- Non-Mendelian inheritance in trios
- Differential missingness in cases and controls

Test for association

- Differences between cases & controls

	AA	AC	CC
Case			
Control			

- Ex. Cochran-Armitage test for trend
- Covariates (age, sex, ...)
- Other genetic models

Odds ratio

- Surrogate measure of effect of allele on risk of developing disease

Allele	A	C	Total
Case	860	1140	2000
Control	1000	1000	2000
Total	1860	2140	4000

Odds of C allele given case status = $\frac{\text{Case C}}{\text{Case A}}$

Odds of C allele given control status = $\frac{\text{Control C}}{\text{Control A}}$

$$\text{Odds Ratio} = \frac{\frac{\text{Case C}}{\text{Case A}}}{\frac{\text{Control C}}{\text{Control A}}} = \frac{1140 / 860}{1000 / 1000} = 1.33$$

Multiple testing

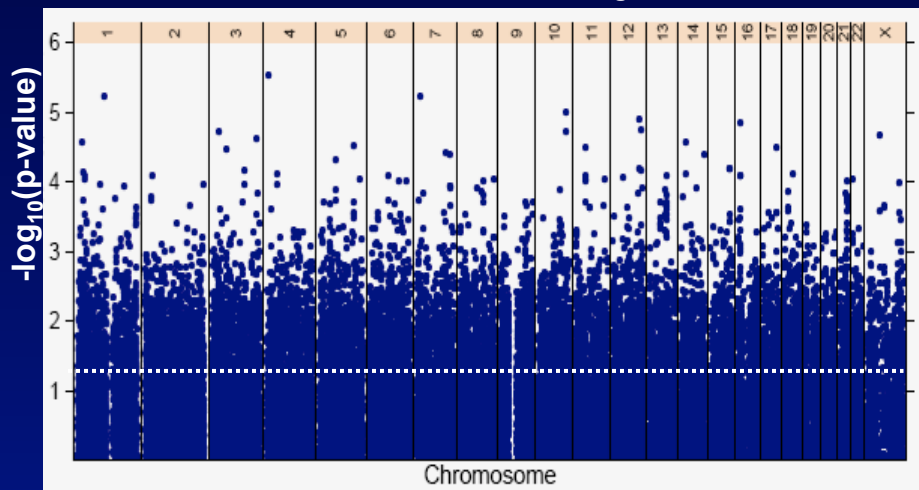
- Genotype and test > 300K – 1M SNPs
- Correct for the multiple tests

$$\frac{.05 \text{ P-value}}{1 \text{ million SNPs}} = 5 \times 10^{-8}$$

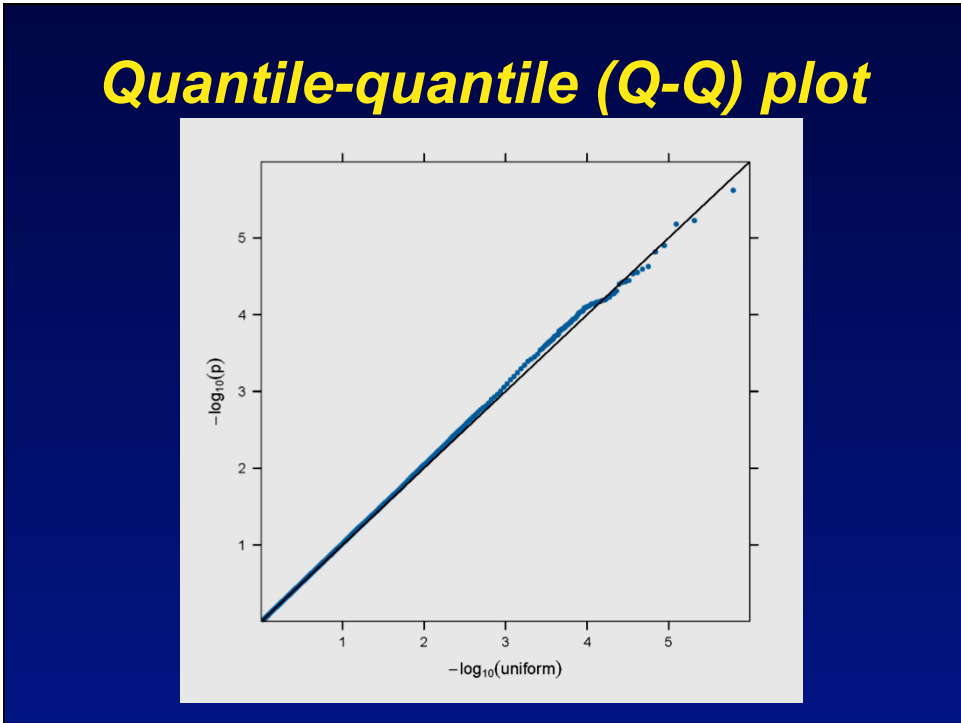
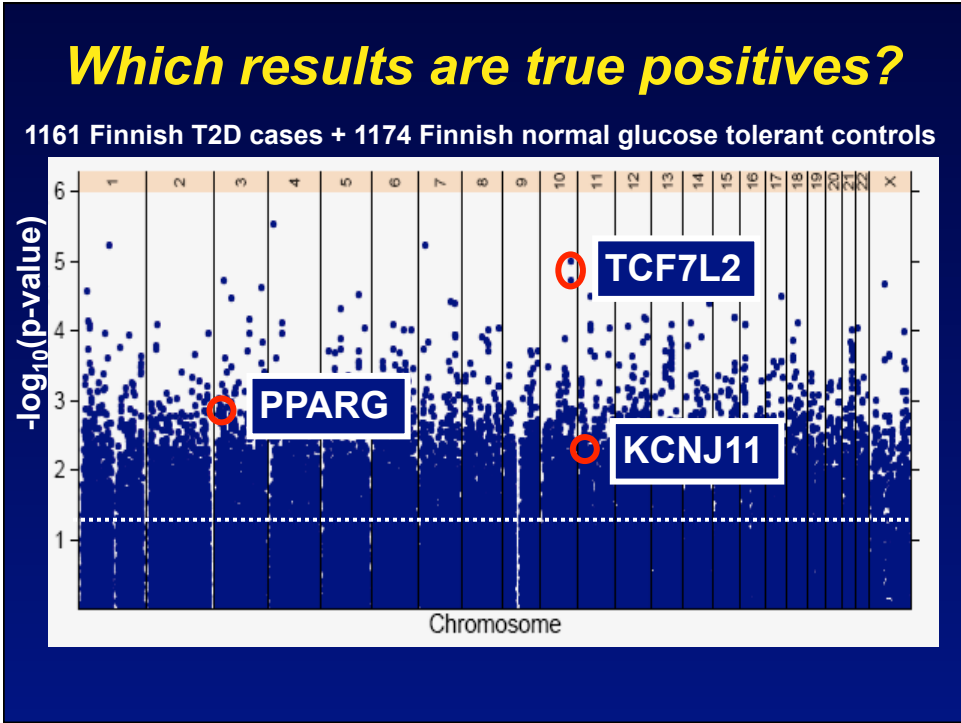
- Need large effect or large sample size

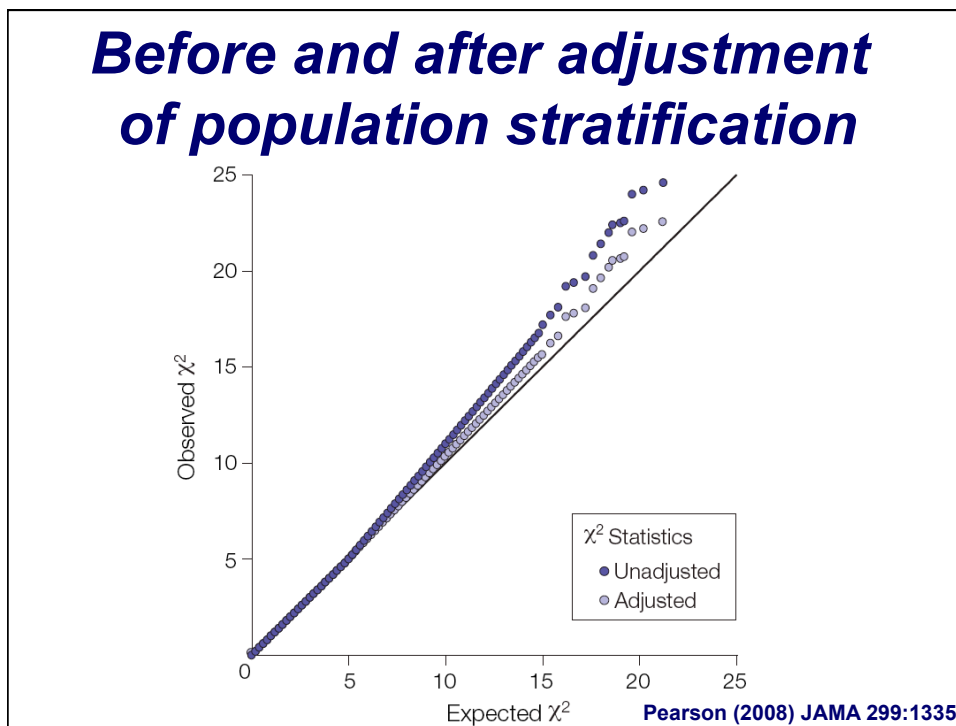
Type 2 diabetes association results

1161 Finnish T2D cases + 1174 Finnish normal glucose tolerant controls



Logistic regression using additive model adjusted for age, gender, birth province





Power to detect association

Table 1 Power of GWASs to discover several recently defined associations

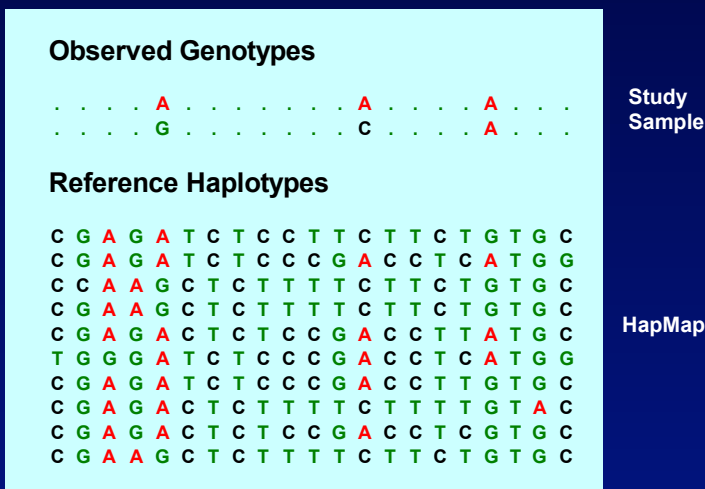
Gene	Disease	Power in a 'typical' GWAS (1,000 cases/1,000 controls)			Sample size required for 90% power, $P < 10^{-8}$	RAF	RR
		1.0×10^{-2}	1.0×10^{-4}	1.0×10^{-8}			
<i>ATG16L1</i>	CD	>0.99	>0.99	0.74	2,430	0.5	1.5
<i>IRGM</i>	CD	0.67	0.19	<0.01	10,902	0.075	1.4
<i>PTPN2</i>	T1D, CD	0.37	0.05	<0.01	19,754	0.17	1.2
<i>IL2</i>	T1D	0.11	<0.01	<0.01	54,600	0.26	1.1
9p21	MI	0.97	0.87	0.09	5,066	0.47	1.25
9p21	T2D	0.36	0.05	<0.01	20,220	0.83	1.2
<i>CDKAL1</i>	T2D	0.35	0.04	<0.01	20,700	0.31	1.15

Altshuler (2007) Nat Gen 7:813

Gain power through collaboration

- Each study performs GWA
- Combine data from all studies by performing a meta-analysis
- Potential issues:
 - Different genotyping and analysis strategies
 - Case definitions are different

Imputation: Observed genotypes



Li (2009) Ann Rev Genomics Hum Genet 10:387

Gonçalo Abecasis

Identify match among reference

Observed Genotypes

. A A A
 G C A

Reference Haplotypes

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

Li (2009) Ann Rev Genomics Hum Genet 10:387

Gonçalo Abecasis

Phase chromosomes, impute missing genotypes

Observed Genotypes

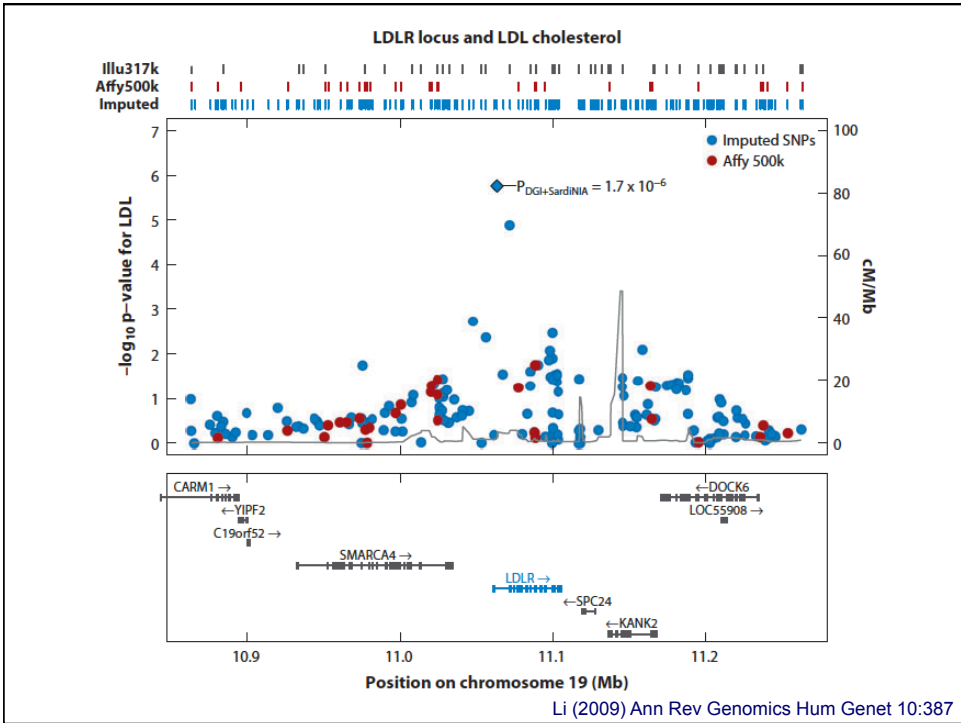
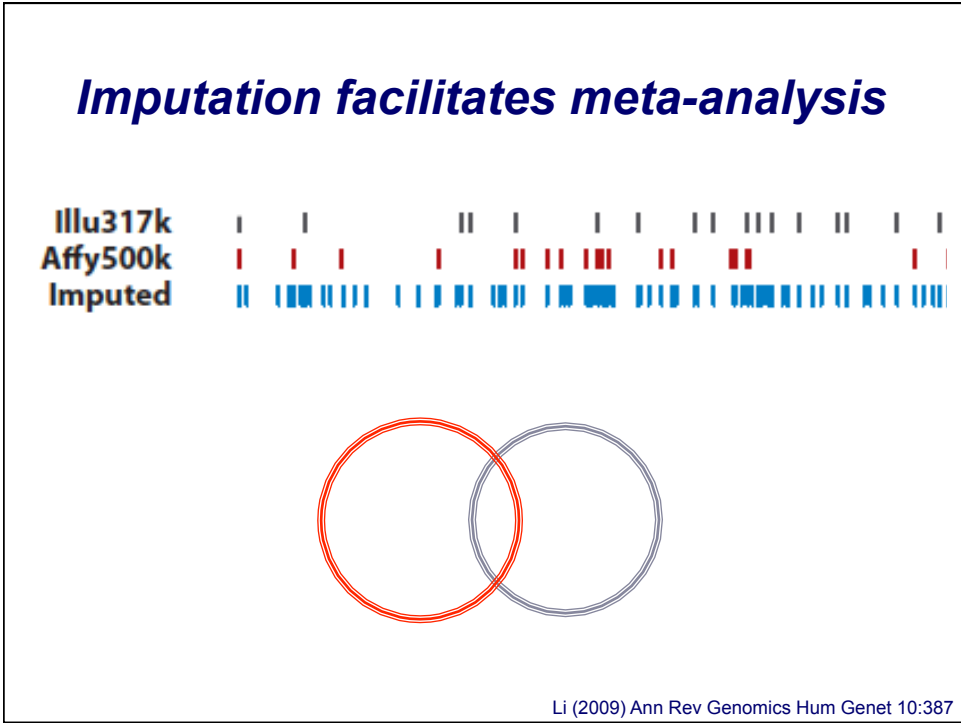
c g a g A t c t c c c g A c c t c A t g g
 c g a a G c t c t t t t C t t t c A t g g

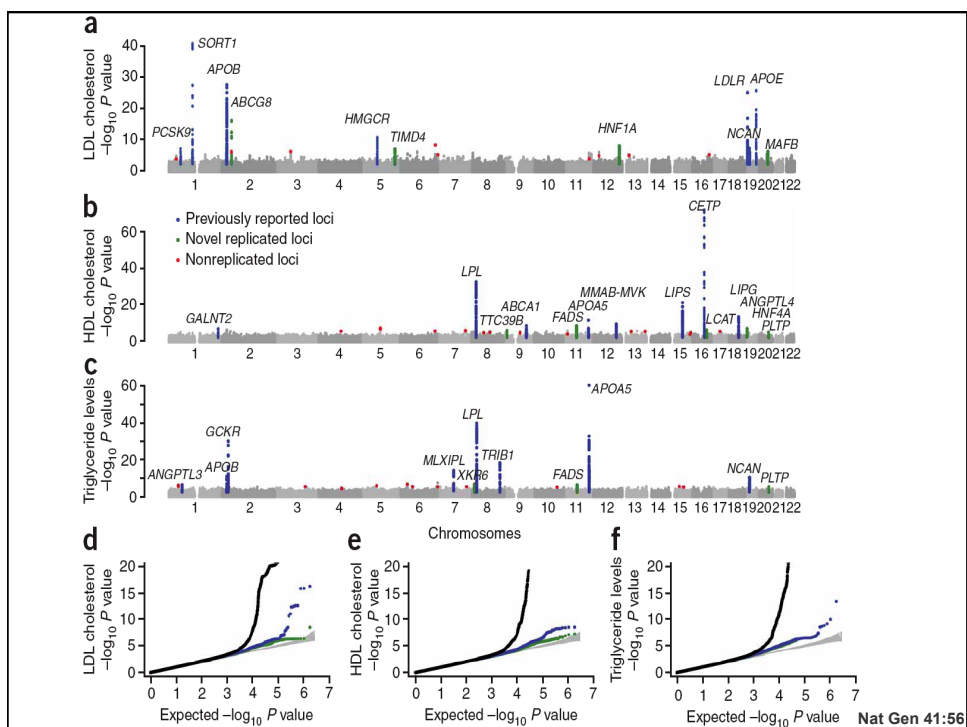
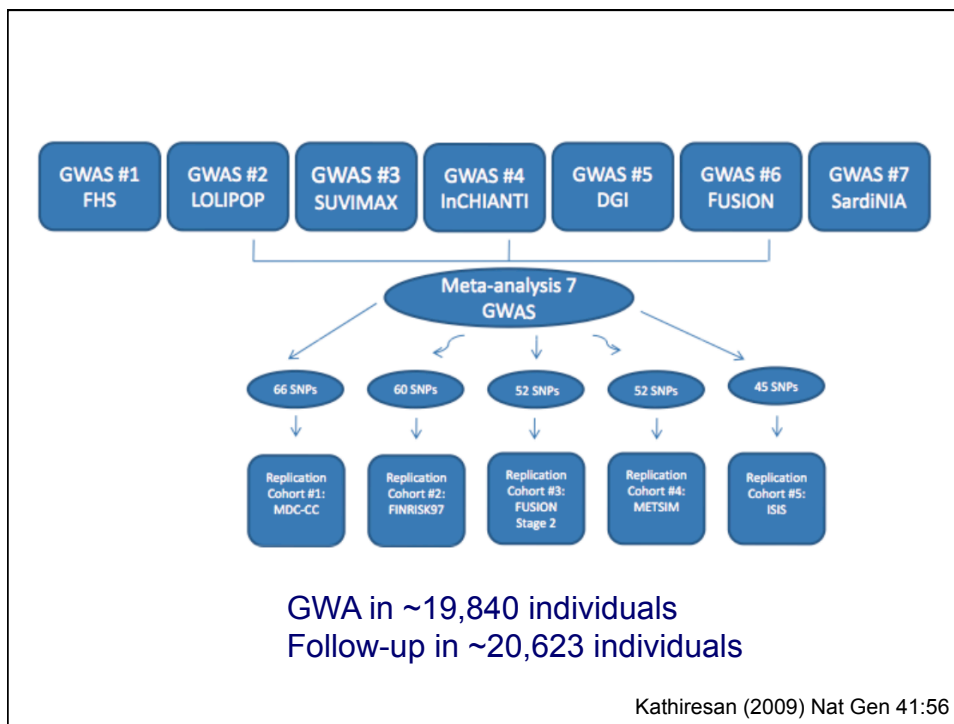
Reference Haplotypes

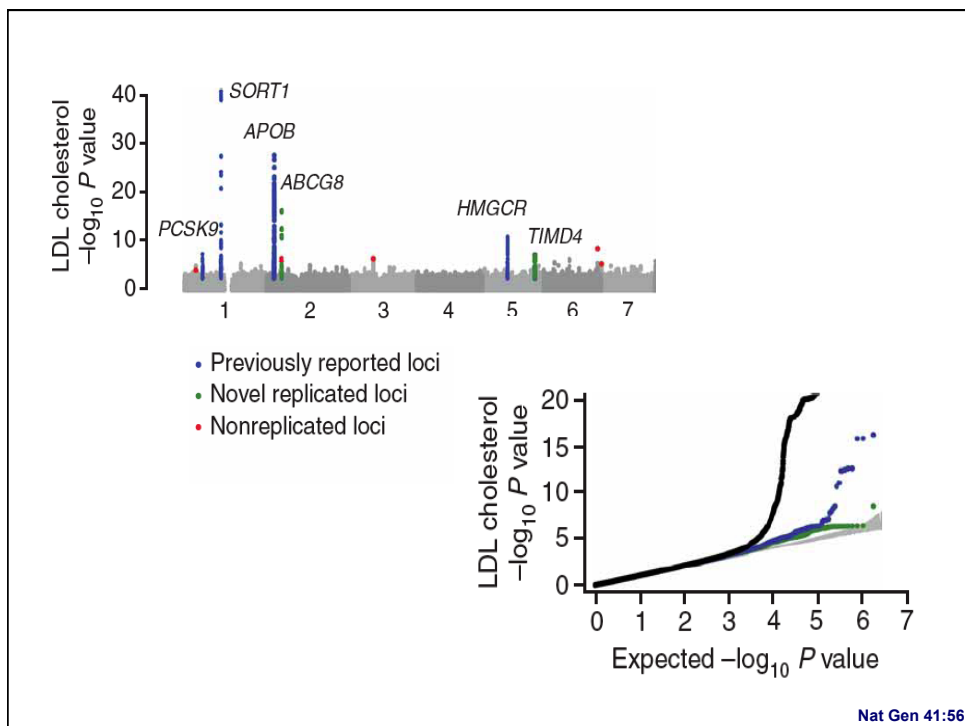
C G A G A T C T C C T T C T T C T G T G C
 C G A G A T C T C C C G A C C T C A T G G
 C C A A G C T C T T T T C T T C T G T G C
 C G A A G C T C T T T T C T T C T G T G C
 C G A G A C T C T C C G A C C T T A T G C
 T G G G A T C T C C C G A C C T C A T G G
 C G A G A T C T C C C G A C C T T G T G C
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

Li (2009) Ann Rev Genomics Hum Genet 10:387

Gonçalo Abecasis



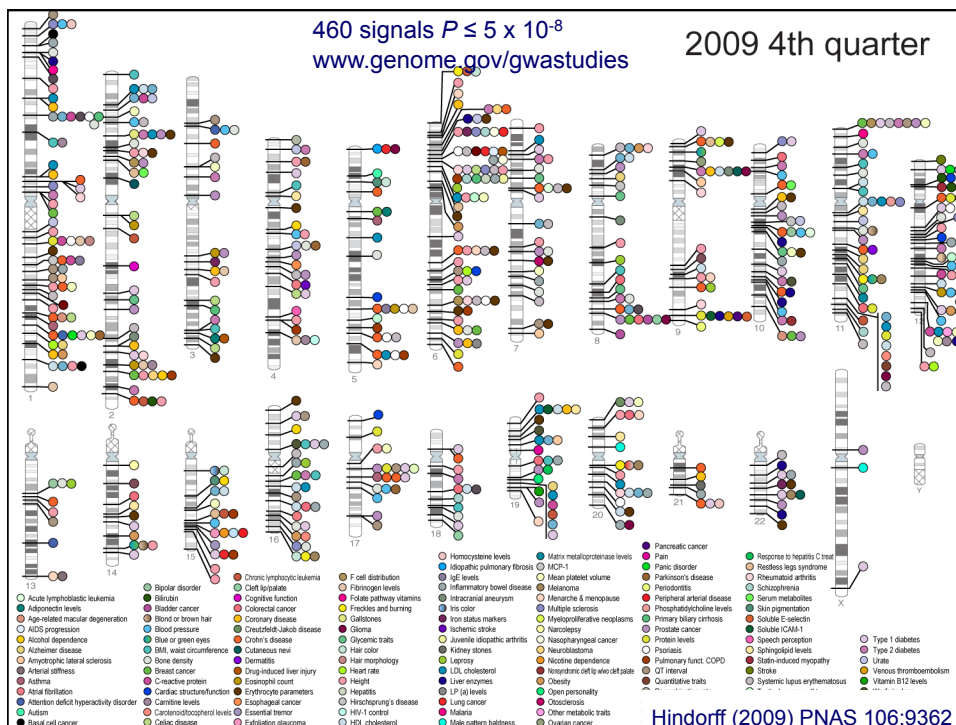


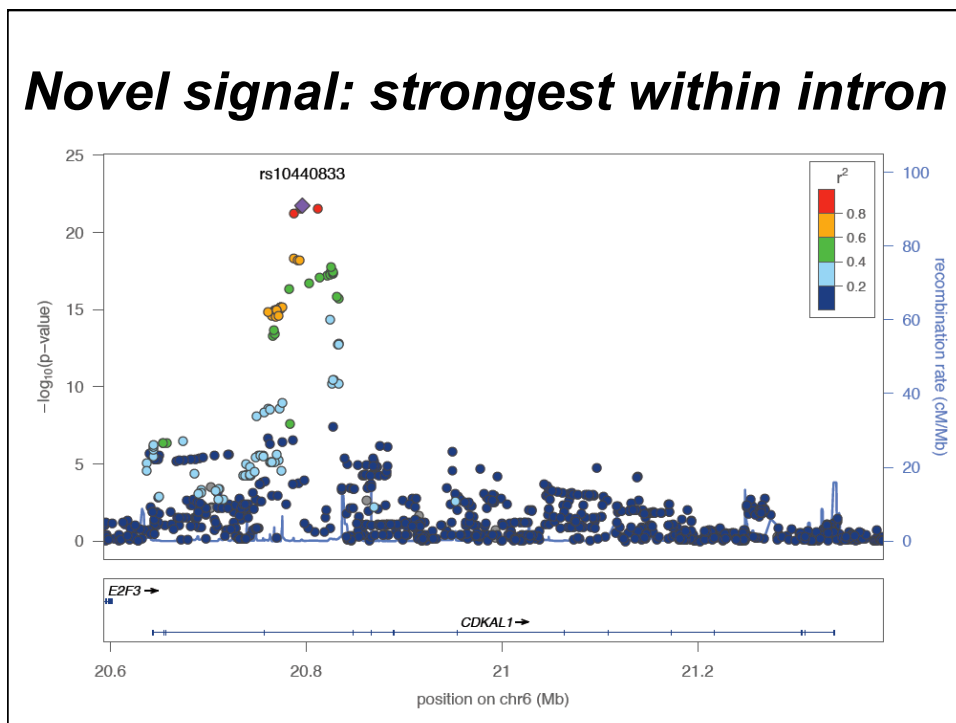
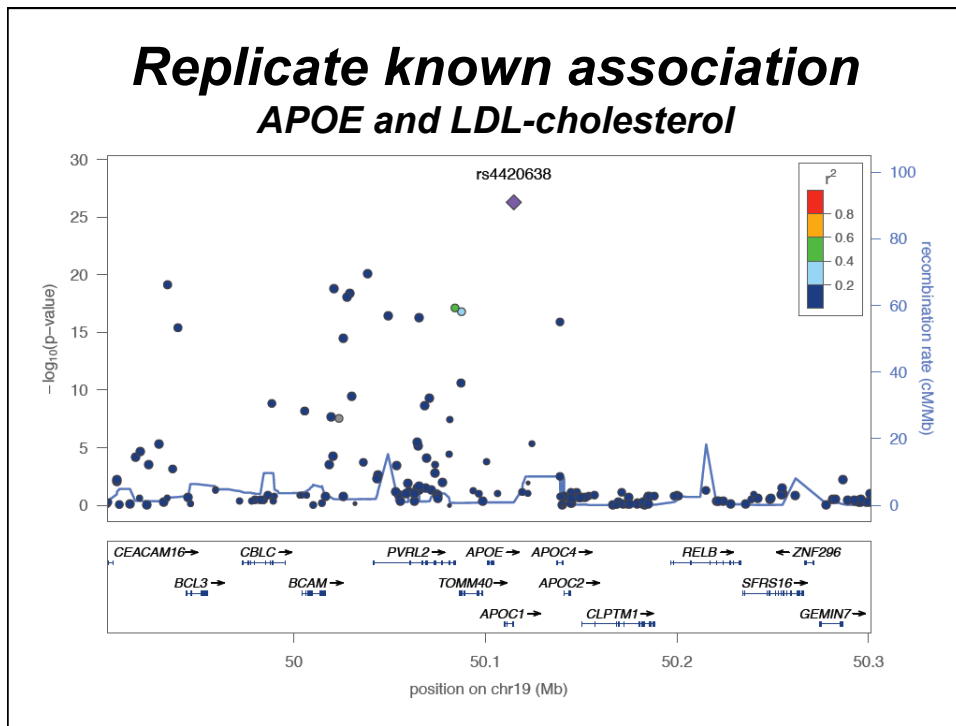


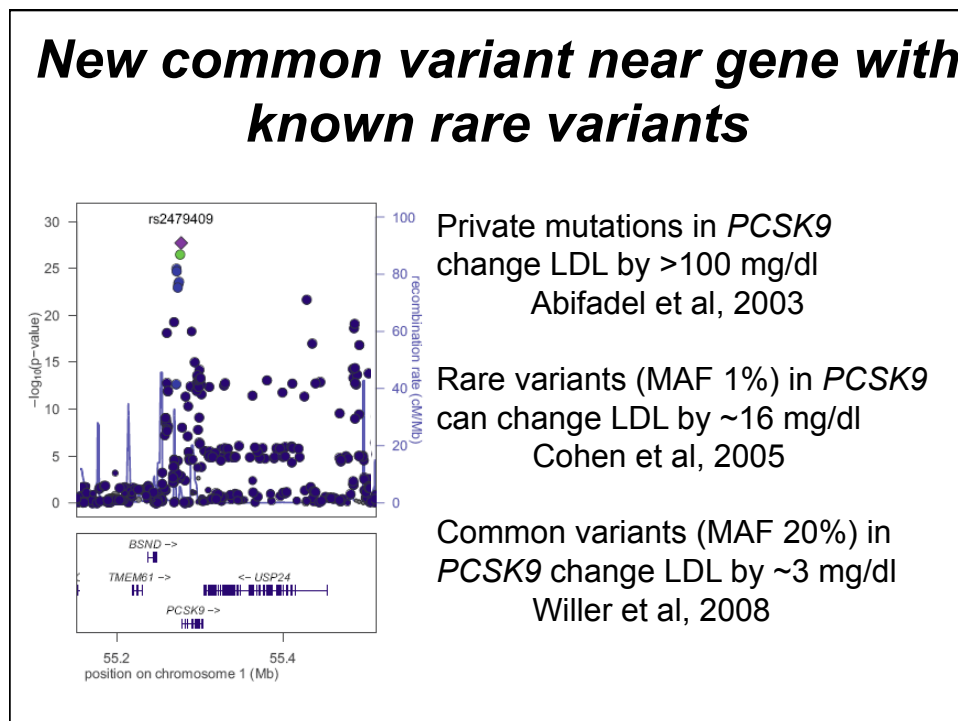
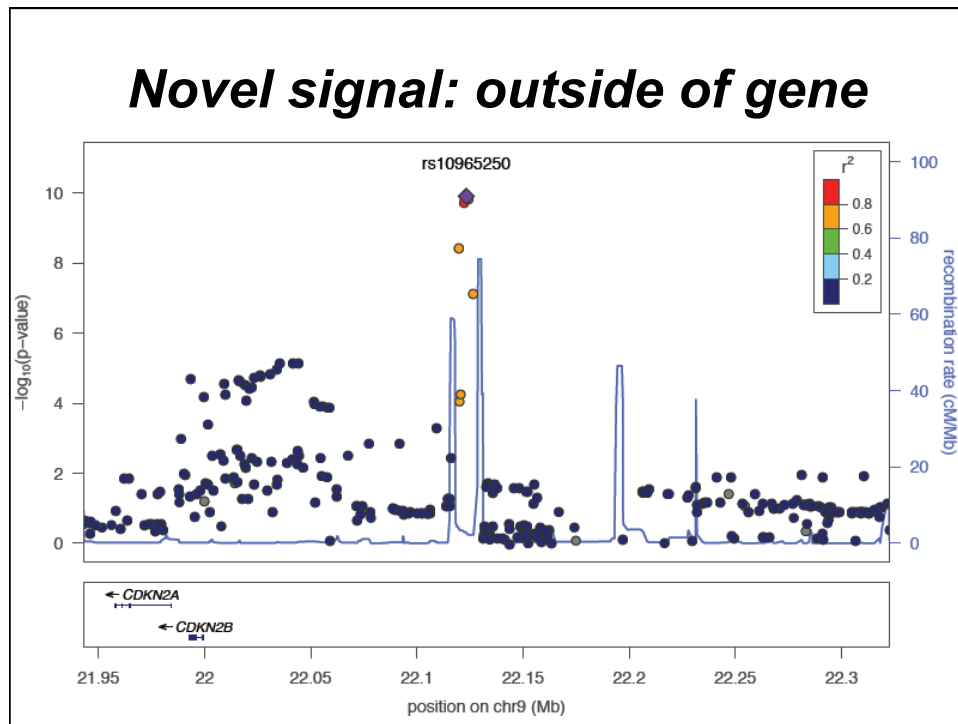
Heterogeneity

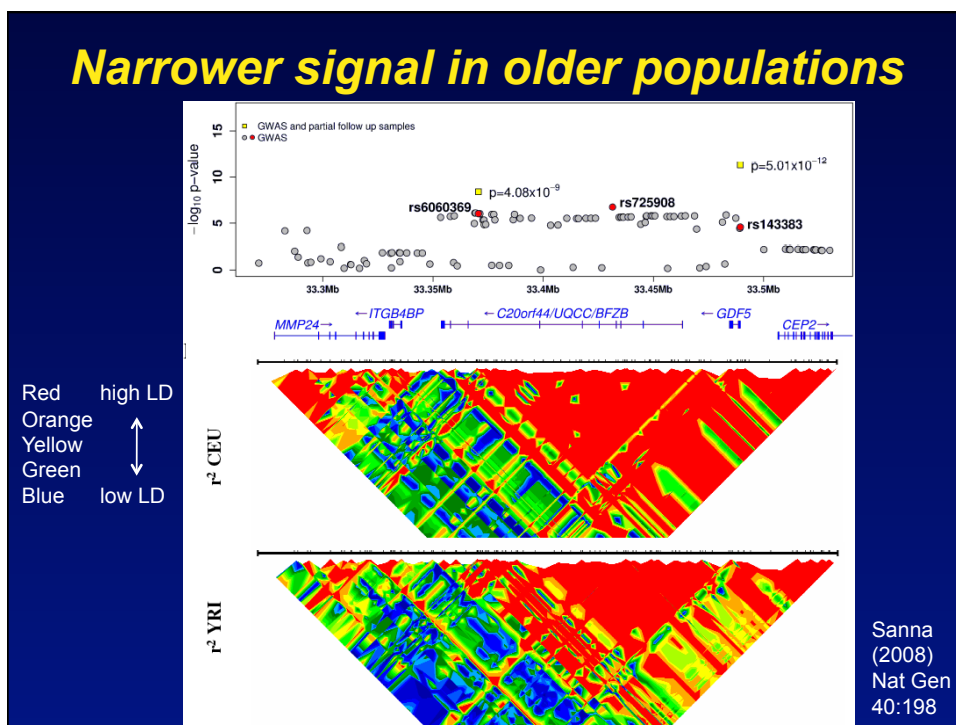
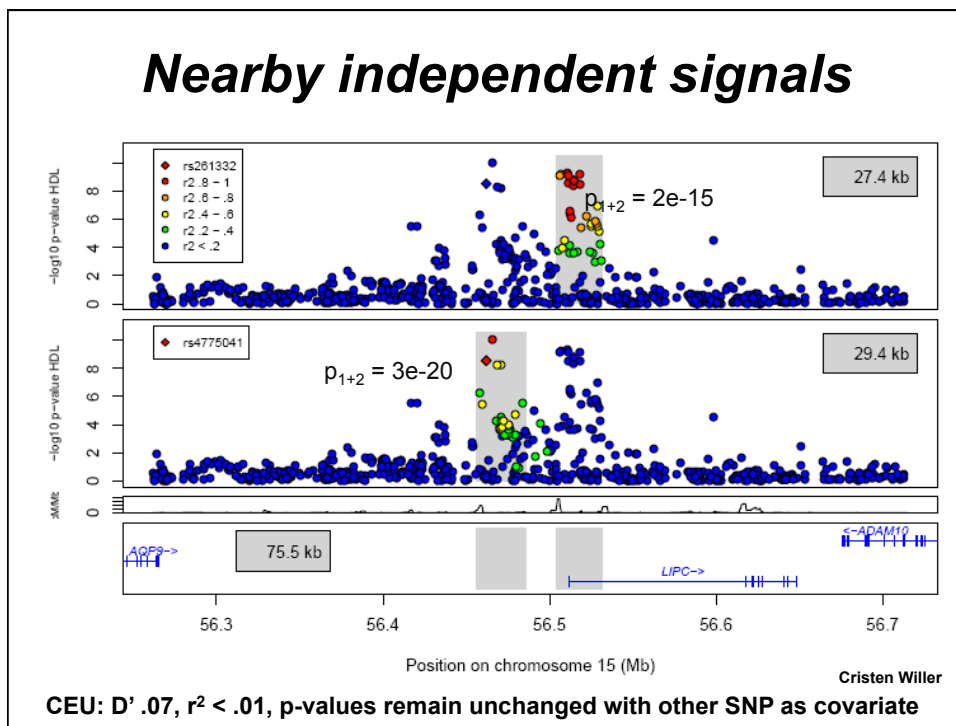
- *FTO* associated with type 2 diabetes in the Wellcome Trust Case-Control Consortium
- Mostly not observed in other diabetes studies
- WTCCC cases more obese than controls
- Diabetes signal abolished when adjust for BMI
- ID of heterogeneity source led to BMI gene

Frayling 2007 Science 316:889







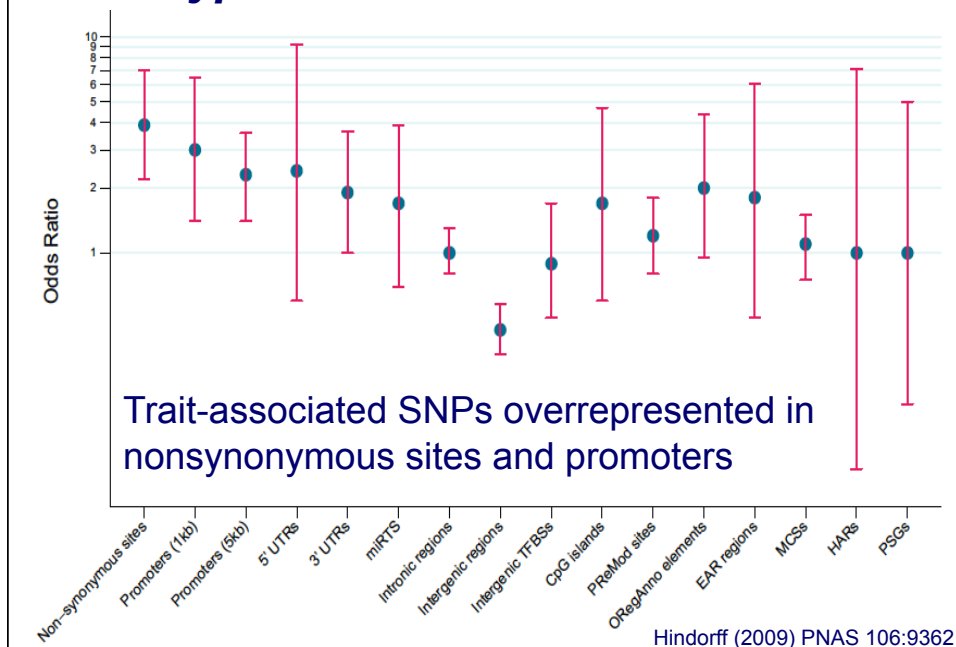


Signals associated with ≥ 2 traits

Attributed genes	Associated traits reported in catalog
<i>PTPN22</i>	Crohn's disease, type 1 diabetes, rheumatoid arthritis
<i>FCER1A</i>	Serum IgE levels, select biomarker traits (MCP1)
<i>BCL11A</i>	Fetal hemoglobin, F-cell distribution
<i>GCKR</i>	CRP, lipids, waist circumference
<i>HLA / MHC region</i>	Systemic lupus erythematosus, lung cancer, psoriasis, inflammatory bowel disease, ulcerative colitis, celiac disease, rheumatoid arthritis, juvenile idiopathic arthritis, multiple sclerosis, type 1 diabetes
<i>CDKAL1</i>	Crohn's disease, type 2 diabetes
<i>IRF4</i>	Freckles, hair color, chronic lymphocytic leukemia
<i>TNFAIP3</i>	Systemic lupus erythematosus, rheumatoid arthritis
<i>JAZF1</i>	Height, type 2 diabetes*
<i>Intergenic</i>	Prostate or colorectal cancer, breast cancer
<i>CDKN2A, CDKN2B</i>	Type 2 diabetes, intracranial aneurysm, myocardial infarction

Hindorf (2009) PNAS 106:9362

Types of associated variants



Small proportion of variability currently explained by common variants

Table 1 | Estimates of heritability and number of loci for several complex traits

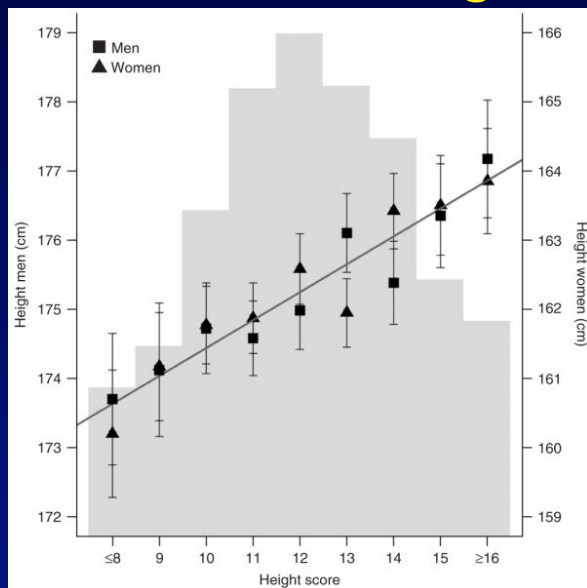
Disease	Number of loci	Proportion of heritability explained
Age-related macular degeneration ⁷²	5	50%
Crohn's disease ²¹	32	20%
Systemic lupus erythematosus ⁷³	6	15%
Type 2 diabetes ⁷⁴	18	6%
HDL cholesterol ⁷⁵	7	5.2%
Height ¹⁵	40	5%
Early onset myocardial infarction ⁷⁶	9	2.8%
Fasting glucose ⁷⁷	4	1.5%

* Residual is after adjustment for age, gender, diabetes.

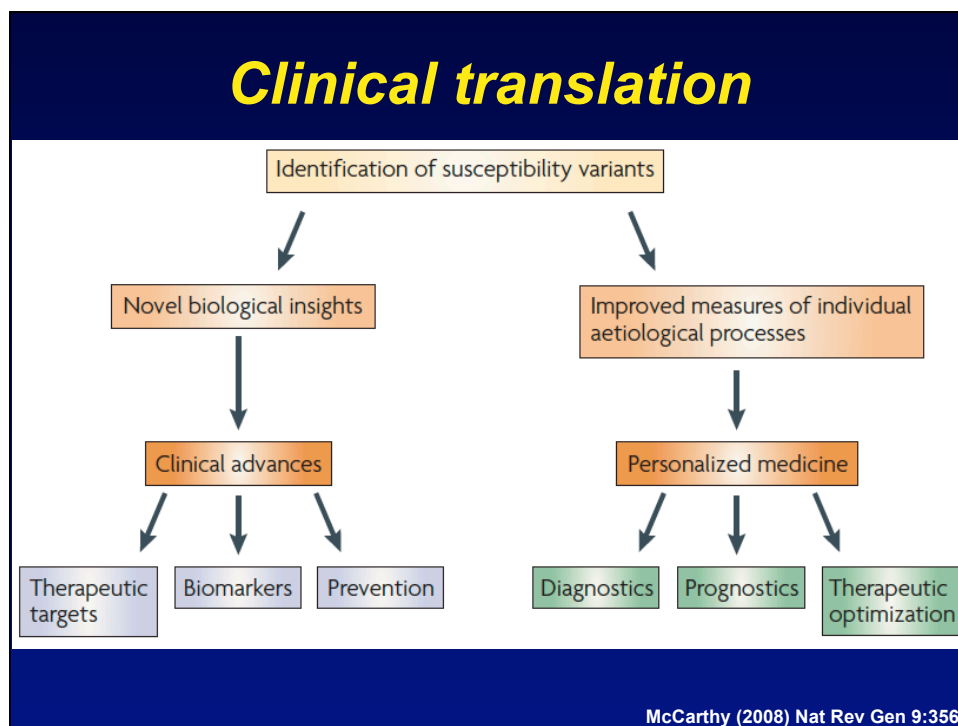
Use of the current information in clinical practice will be disease dependent

Manolio (2009) Nature 46: 747

Prediction of height



Lettre et al. (2008) Nat Gen 40:584-591



Summary

- Need careful attention to design and QC
- Need large samples to find small signals
- 460 signals ($P \leq 5 \times 10^{-8}$) and counting
- Finding an association signal does not immediately yield information on the underlying biology or clinical utility
- Time to changes in medical care based on GWA results may be many years

Future of GWA

- **More and more loci identified**
- **Larger meta-analyses**
- **Deeper follow-up of GWA signals**
- **Larger GWA panels with lower frequency**
- **More diverse populations**
- **Other sequence variants**
- **New phenotypes**
- **Gene-gene and -environment interactions**
- **Molecular and biological mechanisms**