The Role of Genetics Loci in Gene Expression

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1 GENETICS OF GENE EXPRESSION

Chromosome map of disease-associated regions

2019 July										
							V (1990)	ww.ebi.a	c.uk/gwas	

"GWAS have so far identified only a small fraction of the heritability of common diseases, so the ability to make meaningful predictions is still quite limited"

Francis Collins, Director of the NIH, Nature, April 2010

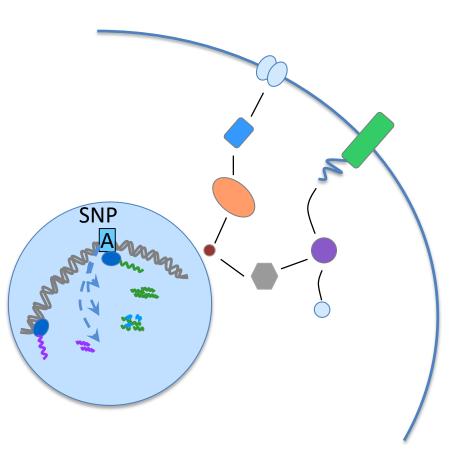
Trait	Heritability (Family base)	Individuals studied	Heritability explained		
Coronary artery disease	40%	86995	10%		
Type 2 Diabetes	40%	47117	10%		
BMI	50%	249796	3%		
Blood pressure	50%	34433	1%		
Circulating lipids	50%	100000	25%		
Height	80%	183727	12.5%		

Motivation

How can we use gene expression and epigenetics to help us understand complex trait genetics?

Majority of trait-associated variation is non-coding.

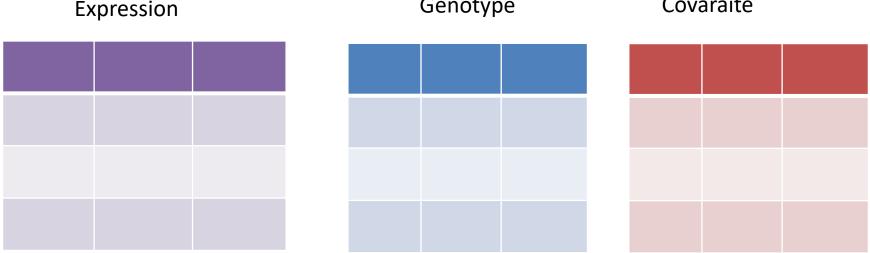
Common hypothesis is that most of these function by altering gene expression.



eQTL analysis Statistics

 Regression: find the coefficients for the effect of expression on genotype when conditioned on the covariates in a linear model and test if they are significant diffetent than 0

gene expression = $\beta_0 + \beta_1$ genotype + β_2 covaraites

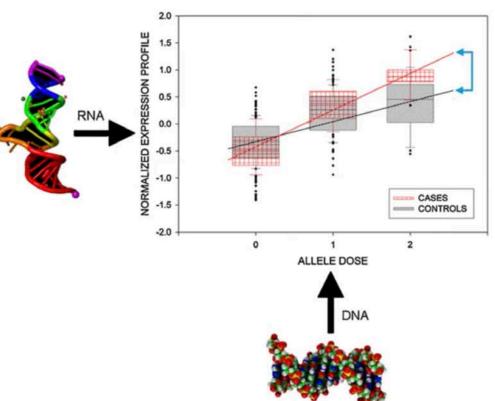


Genotype

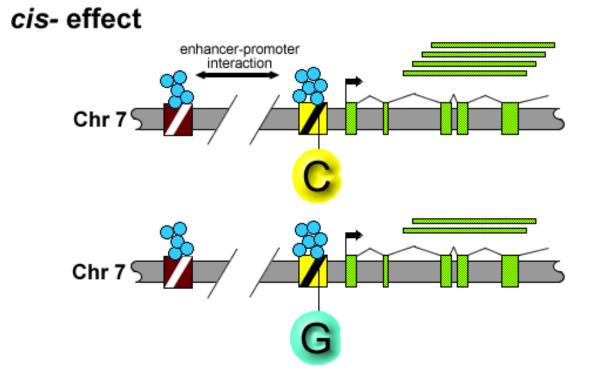
Covaraite

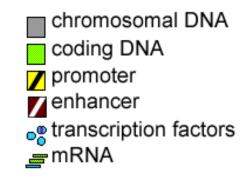
Mapping expression (e)QTL

- RNA expression levels can be treated like any other quantitative trait in QTL mapping.
- 30,000 genes by 10,000 SNPs = 300,000,000 comparisons!
- eQTL studies are sometimes called genetical genomics



Myers, AJ. The age of the "ome": Genome, transcriptome and proteome data set collection and analysis. Brain Research Bulletin Volume 88, Issue 4 2012 294 - 301



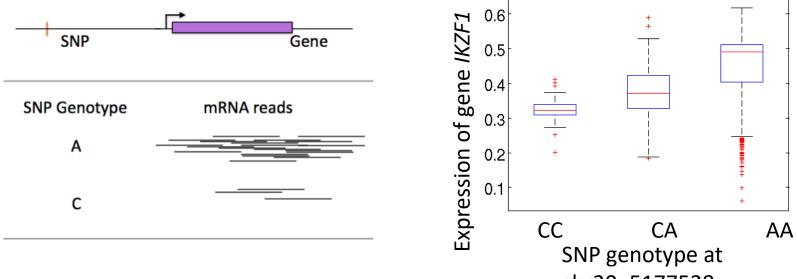


Canonical model

Genetic variants affect gene expression

eQTL (expression Quantitative Trait Locus) analysis:

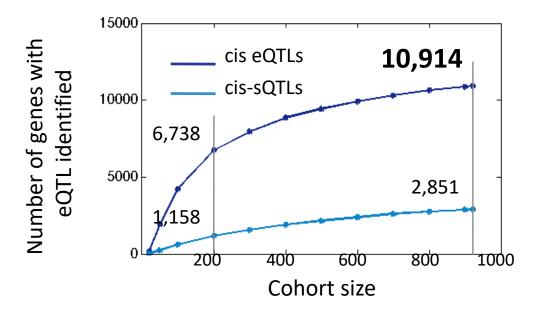
Association between genotype and RNA expression levels



chr20: 5177528

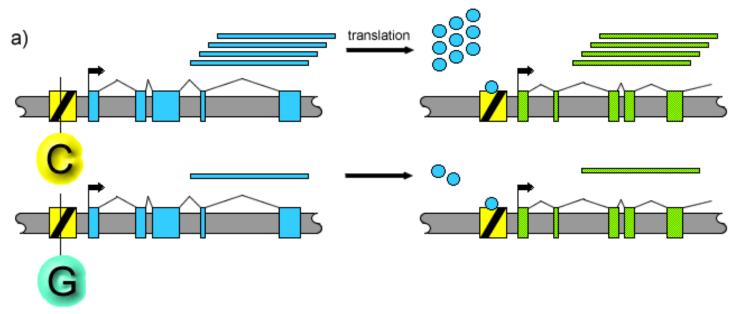
Prevalence of eQTLs

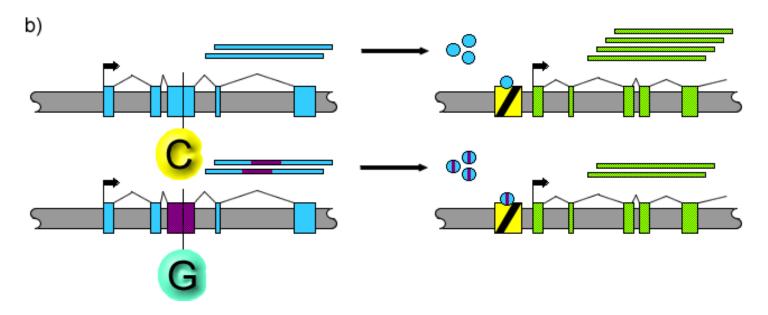
Cis-eQTLs have now been identified for nearly every human gene, with numerous large studies available

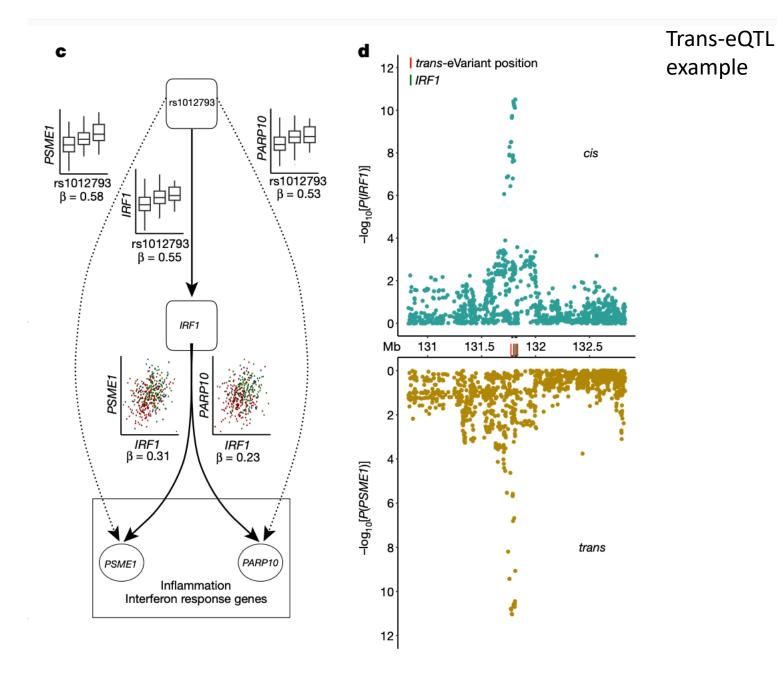


Battle, Genome Research, 2014

trans- effect

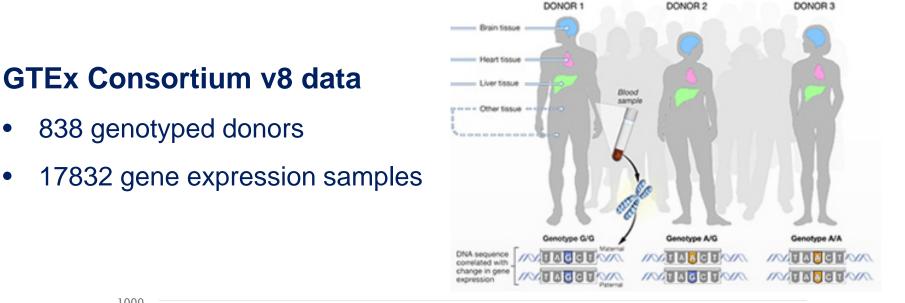


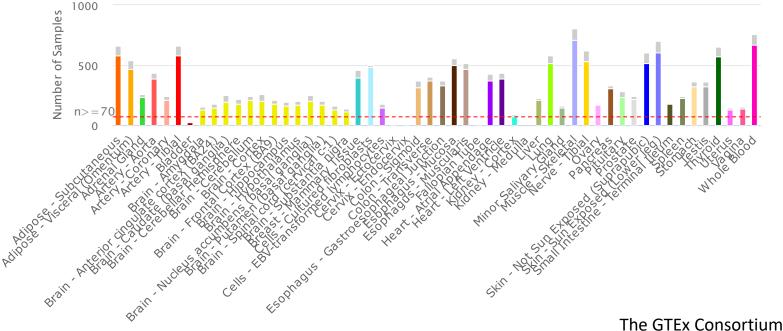




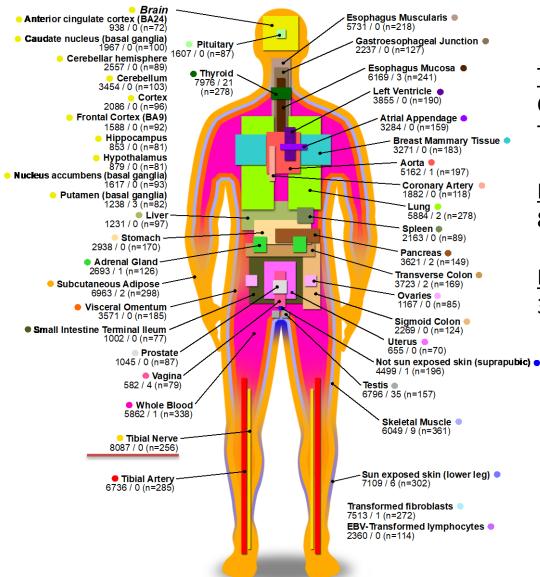
The GTEx Consortium, 2017

GTEx Project





Genetic effects across human tissues



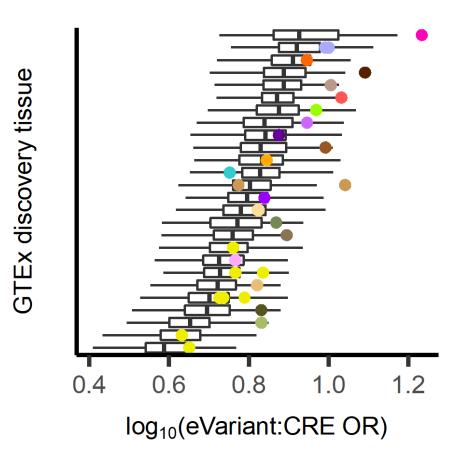
Total unique eQTL genes: Cis: 19,725 (FDR 5%) Trans: 93 (FDR 10%)

<u>Most cis per tissue:</u> 8,087 Tibial nerve (N=256) (脛神經) <u>Most trans per tissue:</u> 35 Testis (N=157)

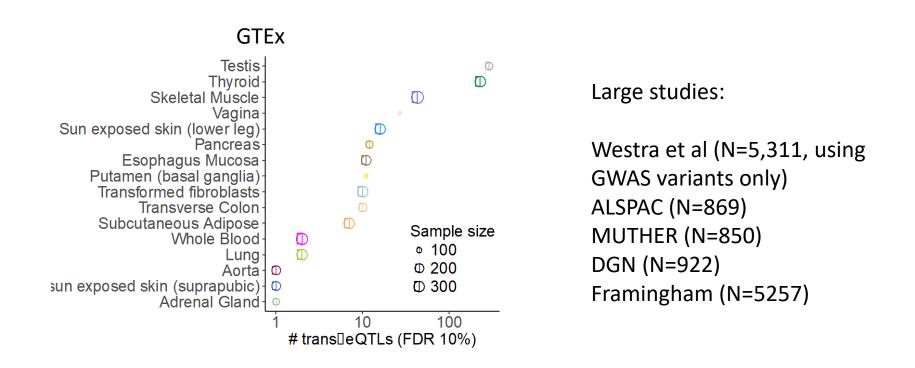
The GTEx Consortium, Nature 2017

Characterizing eQTLs across tissues

• Cis-eQTL variants fall in tissue-specific regulatory elements (from Roadmap Epigenomics)



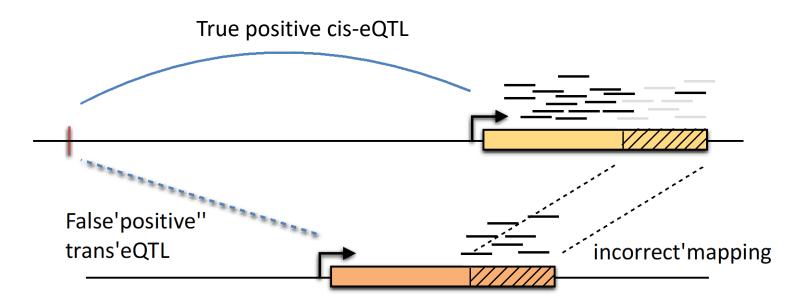
Trans-eQTLs



Studies report wildly different # hits (10s–10000s) Replication and validation remains poor

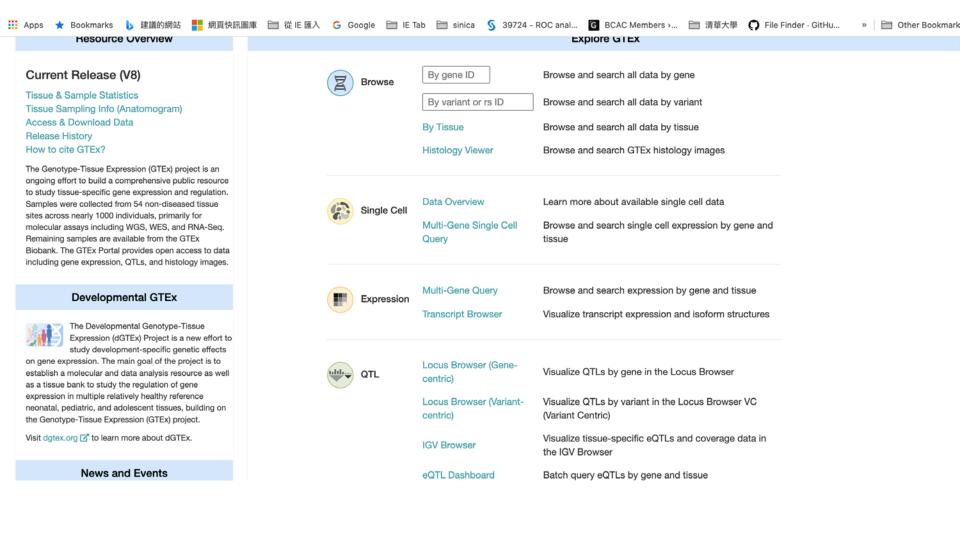
Challenges for trans-eQTL detection

- Power
- False positives from many sources e.g. over and under correcting confounders (Dahl et al, 2017)
- Mapping error (similar to probe cross-hybrid.)



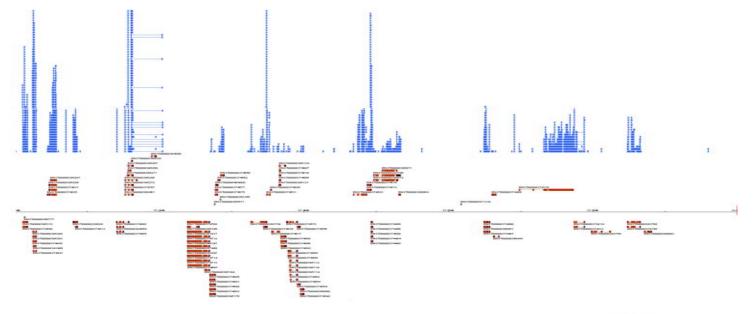
Introduction of GTEx protal

https://gtexportal.org/home/



2 MANY TYPES OF QTLS

Next generation sequencing has increased our ability to survey the transcriptome.

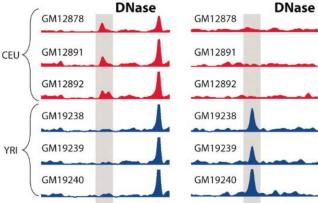


Montgomery, Nature 2010 Pickrell, Nature 2010

RNA-Seq

ChIP-Seq

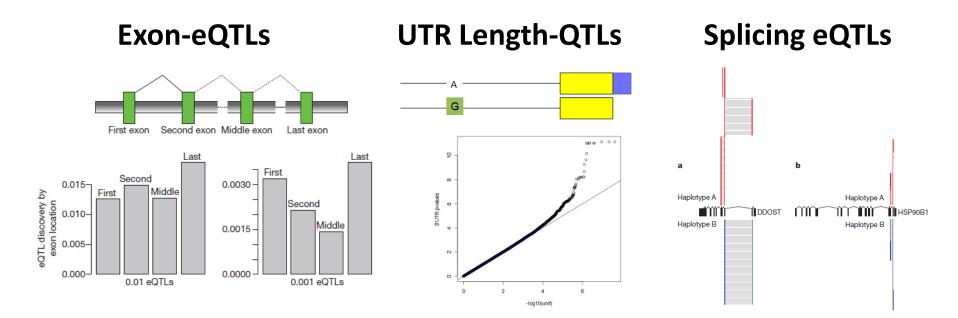
McDaniell, Science 2010



RNA-seq provides resolution of more QTLs

RNA-sequencing in 60 Europeans (HapMap genotypes; LCLs)

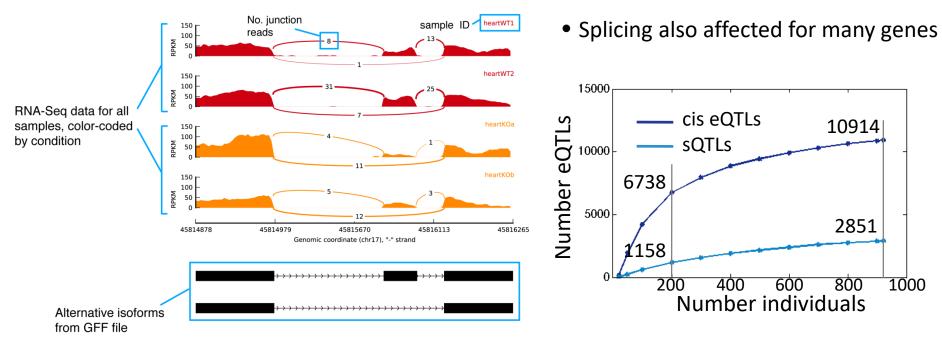
Found 2x more expression Quantitative Trait Loci (eQTLs) and...



Rare eQTLs with allele specific expression-based approaches

Splicing eQTL

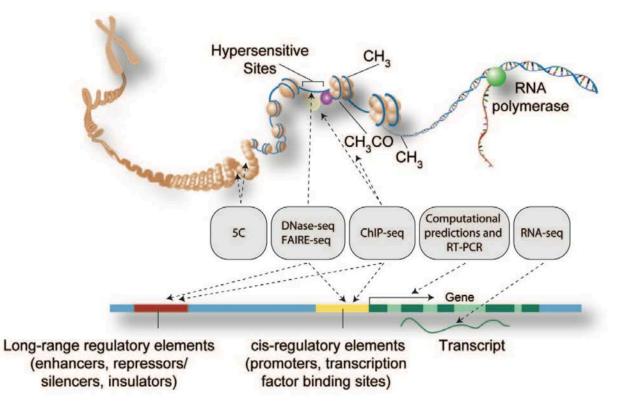
Can investigate relative transcript ratios or reads across junctions.



Katz et al, Nature Methods, 2010

Battle et al, Genome Research, 2014

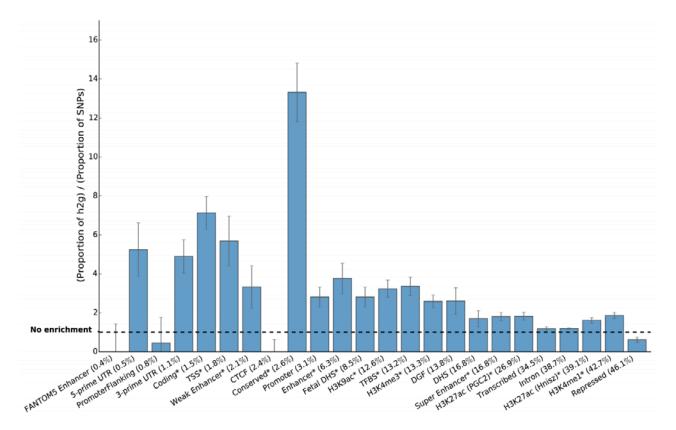
Epigenetic data



- ENCODE, Roadmap Epigenomics
- Regulatory elements: promoters, enhancers
- Transcription factor binding sites
- CpG sites
- ChromHMM

Epigenetic data informs heritability

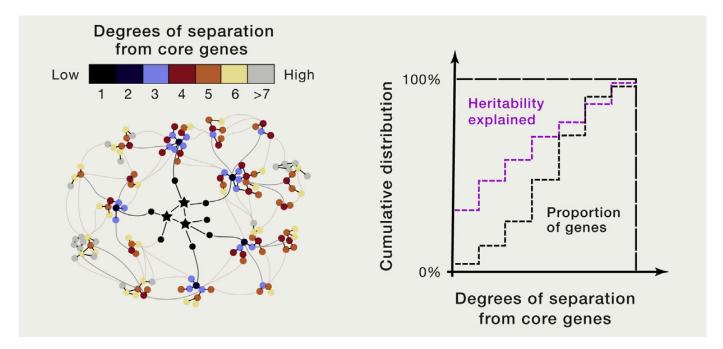
LD score regression, related approaches partition h^2



Large scale epigenetic data (Roadmap, ENCODE) enable analysis, indicate contribution of gene regulation

Ommigenic model

 Most/all expressed genes in disease-relevant cell types affect trait

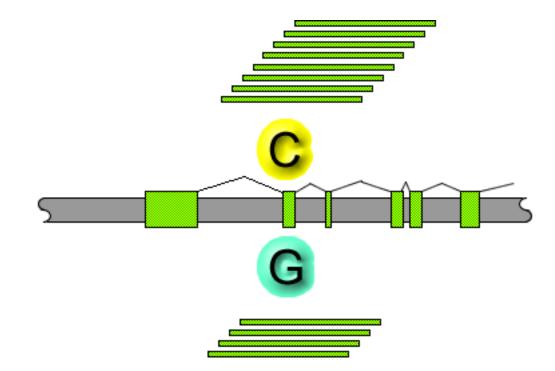


• Highlights potential role of eQTLs, trans effects

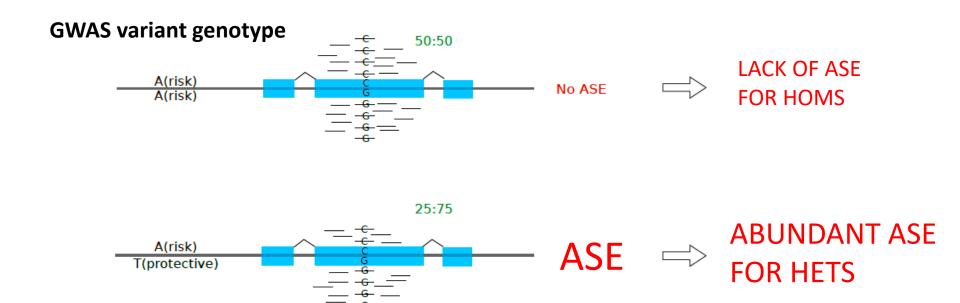
Boyle et al., Cell, 2017

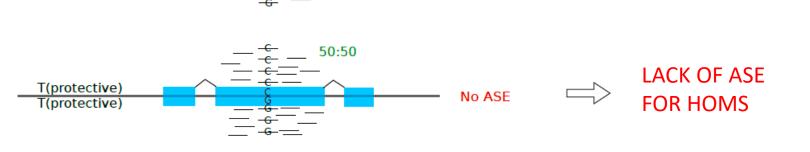
Advantages of ASE

• Test within an individual allelic imbalance, given one has sufficient reads.



Using ASE to detect GWAS signals driven by <u>multiple</u> causal variants





Tests functional differences between alleles in population

Lucia Conde et al, AJHG, 2013

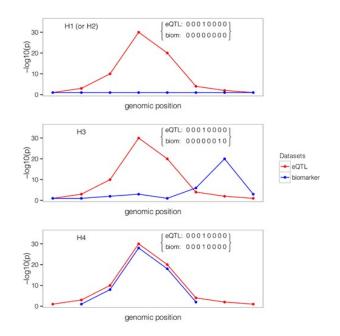
Coloc: A Bayesian test for colocalization of pairs of association signals

H1 is the hypothesis that there is only an eQTL signal at a locus

H2 is the hypothesis that there is only a GWAS signal at a locus.

H3 is the hypothesis that there are two independent eQTL and GWAS signals in linkage.

H4 is the strong hypothesis that the same SNP (not just the locus) is responsible for both the GWAS and eQTL.



Giambartolomei et al (2014) PLoS Genetics 10(5): e1004383

GWAS eQTL colocalization

- Coloc
- eQTpLot
- hypercoloc

Examples of H3 and H4

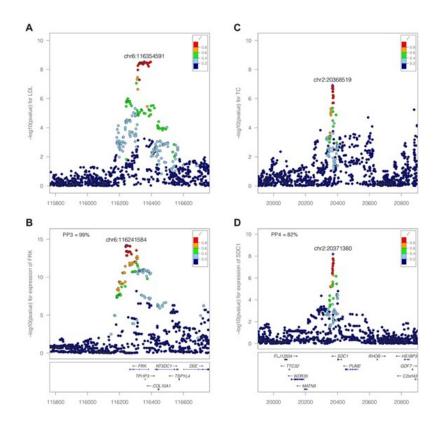
On the left, the profile of association at the FRK locus with LDL (top) is very different from that with *FRK* expression.

H3 is the supported hypothesis.

On the right, even though there are two different peak SNPs, they are in the same strong LD region and the profiles are almost the same for Total Cholesterol and *Soc1* expression.

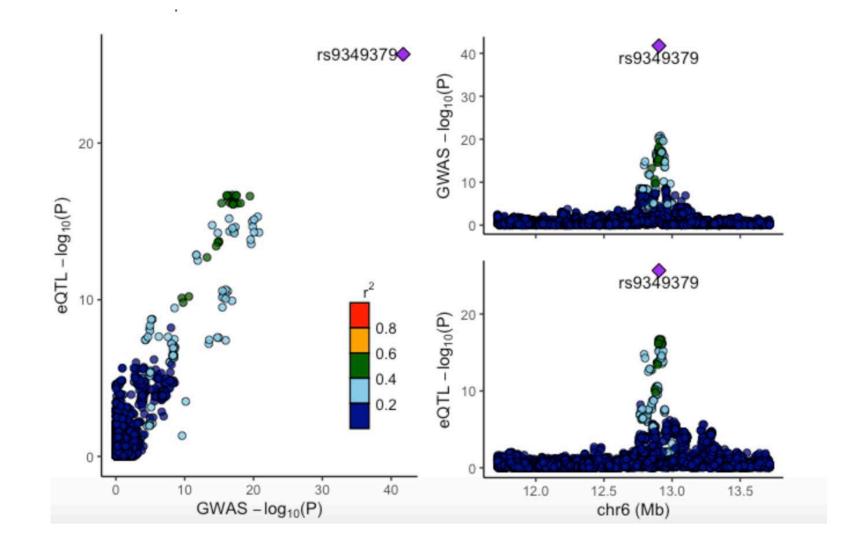
H4 is the supported hypothesis.

Bayesian analysis evaluate each H relative to the other four and generates a confidence level for the most likely one.



Giambartolomei et al (2014) PLoS Genetics 10(5): e1004383

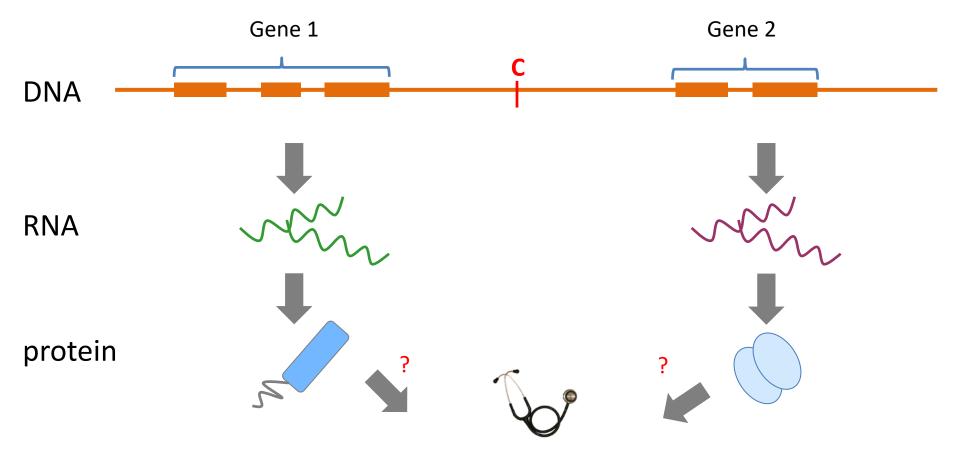
Coloc results



eQTLs and complex disease genetics

Help interpret GWAS variants (especially non-coding):

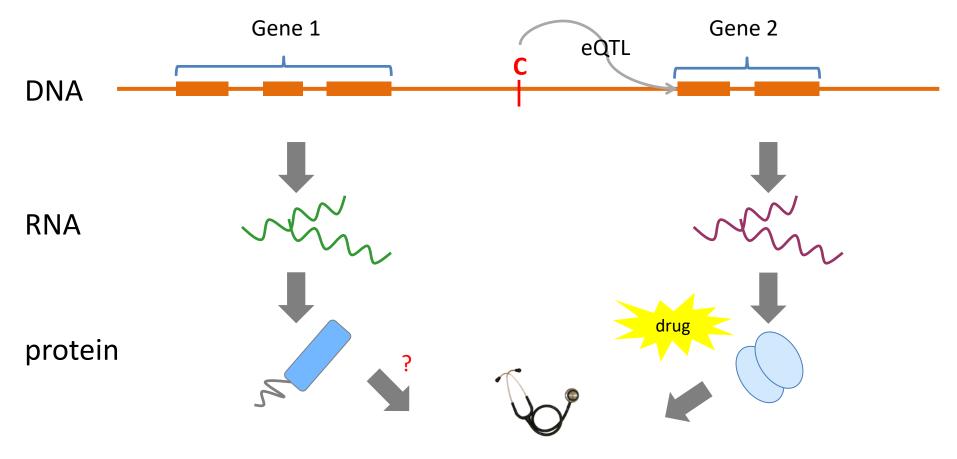
- understand mechanism
- guide interventions



eQTLs and complex disease genetics

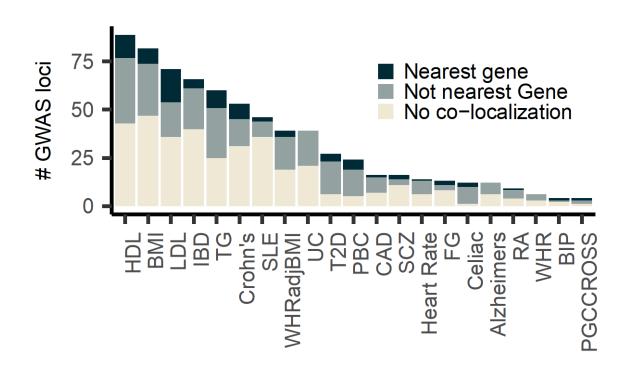
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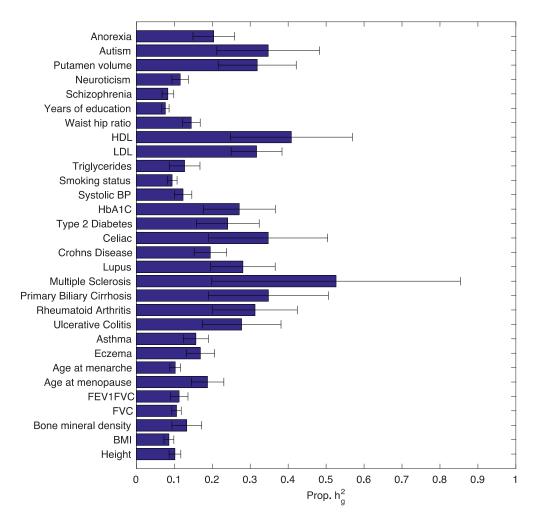
eQTLs and complex disease genetics

52% of genetic variants associated with human disease co-localize with an eQTL



eQTL data informs heritability

GE co-score regression indicates cis-eQTLs explain mean 21% of h^2 across a set of complex traits



O'Connor et al. bioRxiv, 2017

3. Complex effects of genetic variation on gene expression

What are we missing?

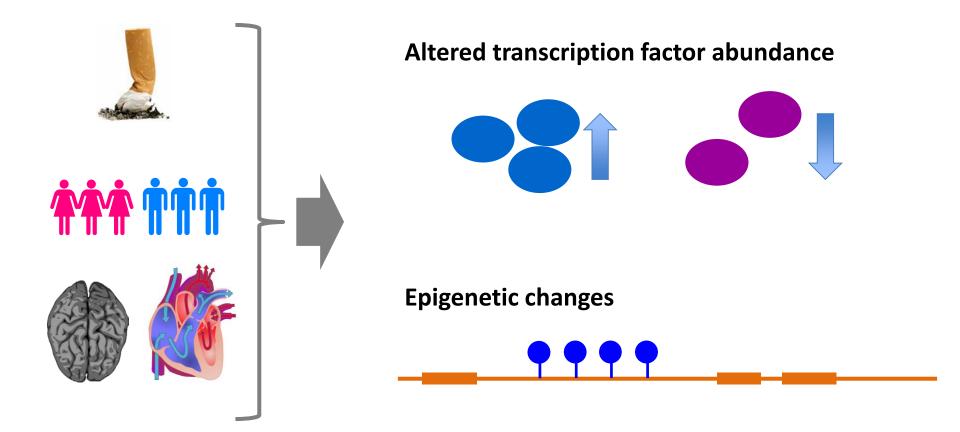
 Most studies are done on steady-state total expression measurements at a single adult or post-mortem time point

 Disease-relevant states include different developmental stages, environmental exposures, cell types

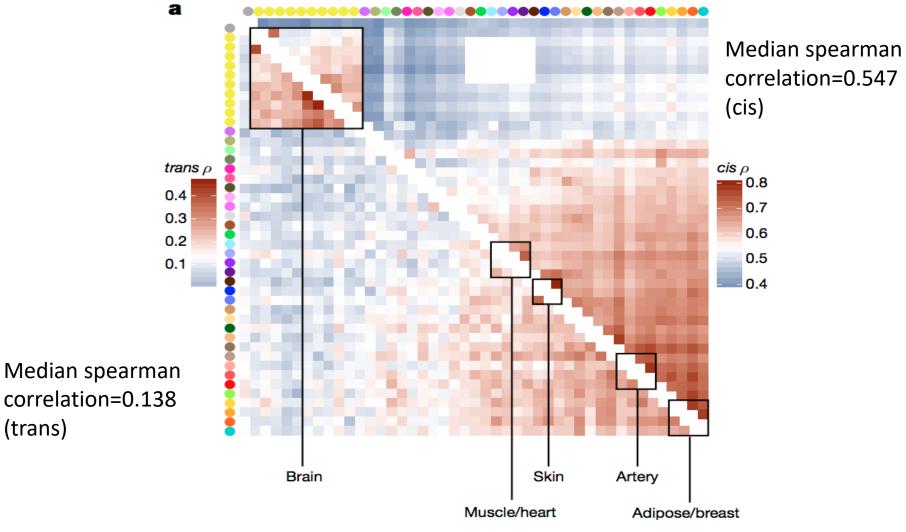
• Other variant classes and regulatory effects

Context-specificity

Many factors can *modulate* regulatory effects



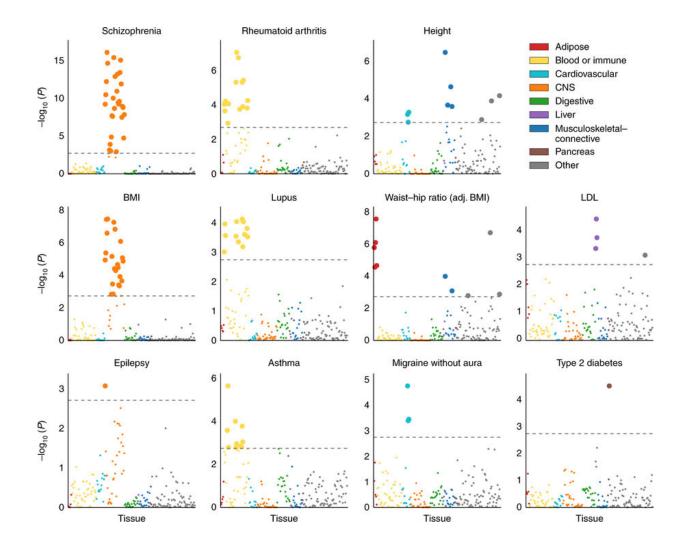
GTEx tissue-specificity of cis and trans



Trans eQTLs appear more highly tissue-specific than cis-eQTLs

(trans)

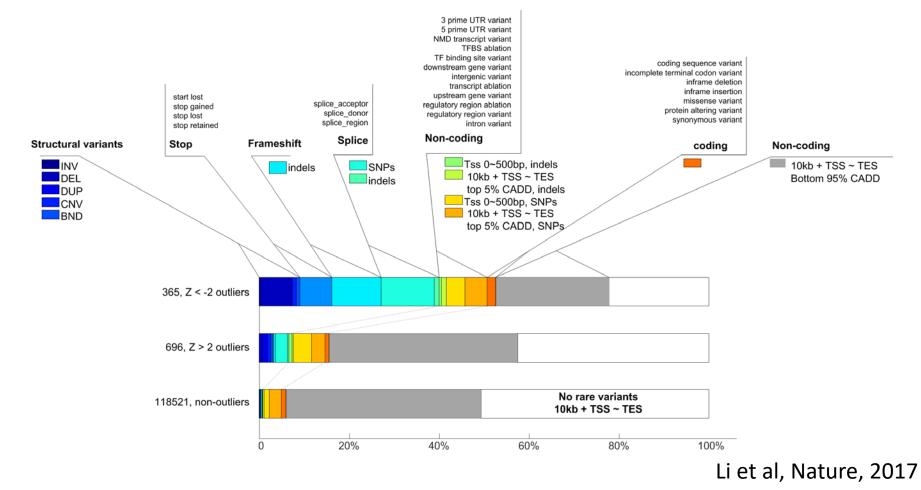
Tissue specificity and heritability



From Finucane et al, NG, 2018

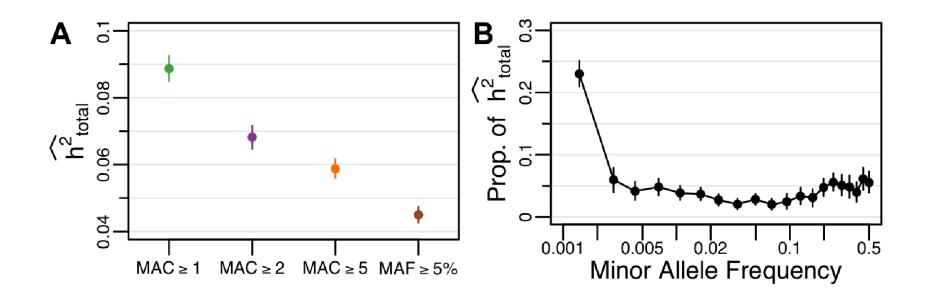
Rare variants

Recent work emphasizes importance of rare variation in driving extreme expression levels



Rare variants

Preprint (Hernandez et al 2017) suggests rare variants explain a large fraction of heritability of gene expression



4. Conclusions

Why delve deeper into expression?

- Help determine when and how much to invest in WGS, expression, epigenetic data
- To continue understanding implicated
 - Genes
 - Tissue and cell types
 - Epigenetic and other regulatory mechanisms
- Challenges and caveats
 - Ambiguity: many variants affect multiple genes
 - Interpretability: missing relevant cell types
 - Power: trans-eQTLs also require large sample sizes

5. eQTL tool demo

matrixEQTL

Statistical Clinics

- 1. It is a special service for free for researchers in Academia Sinica.
- 2. The service is offered at 14:00-16:00, Thursday, in Institute of Statistical Science Building, Room 401.

3. You are welcome to apply for the service. http://disc.stat.sinica.edu.tw/statistical-clinic-service-appointment/



Thank you