

Transcriptome and genome sequencing uncovers functional variation in humans

Reproducibility of high-throughput mRNA and small RNA sequencing across laboratories

Aim of study

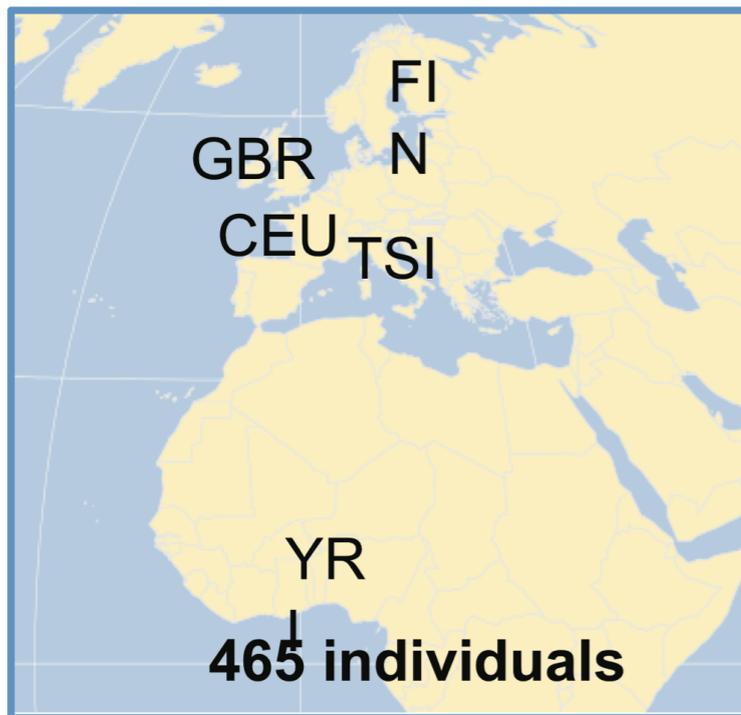
Characterize functional variation in human genomes

- catalogue novel **loci** with regulatory variation
- discover & characterize **molecular properties** of **causal functional variants**

What do GWAS variants actually do in the cell?

Setup

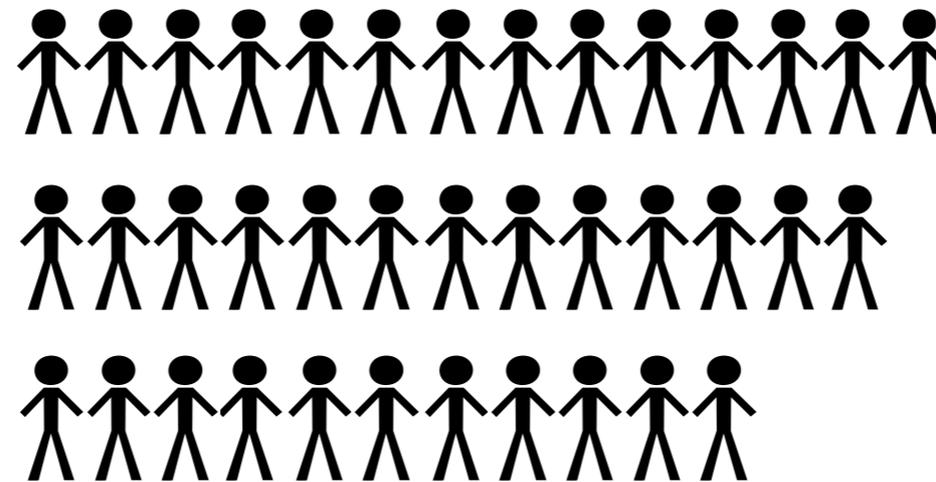
Lymphoblastoid cell lines
Samples from 1000 genomes project
mRNA and sRNA sequencing



Populations

Europe, Finns, British,
Toscani, Nigeria

Data (after control)



462 - mRNA

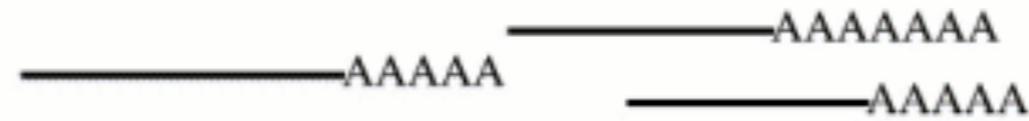
452 - sRNA

421 - in 1000

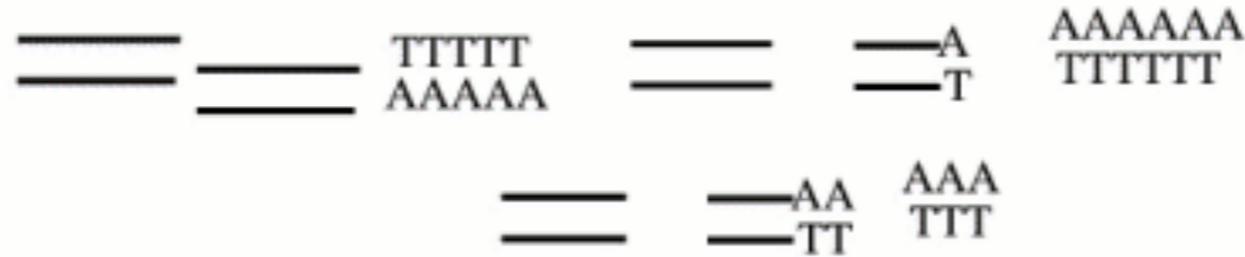
Genome Phase 1
(rest SNP array)

RNA Seq

extraction of poly-A RNAs



conversion into ds-cDNA
and shearing

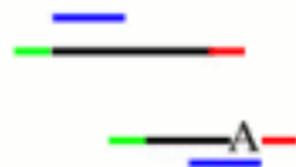


amplification and
adapter ligation

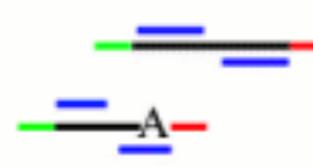


sequencing

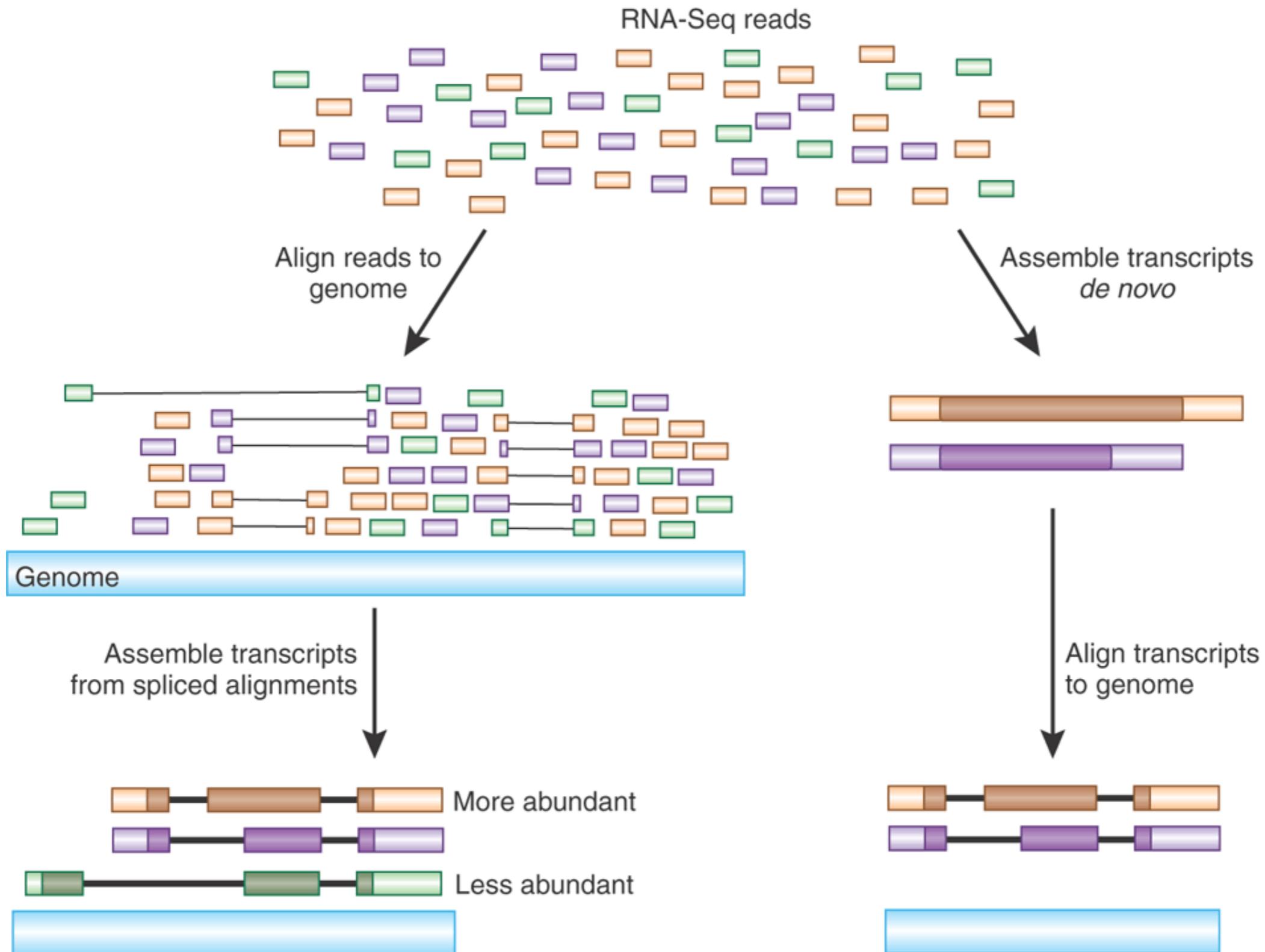
single end (SET)



paired-end (PET)



RNA Seq

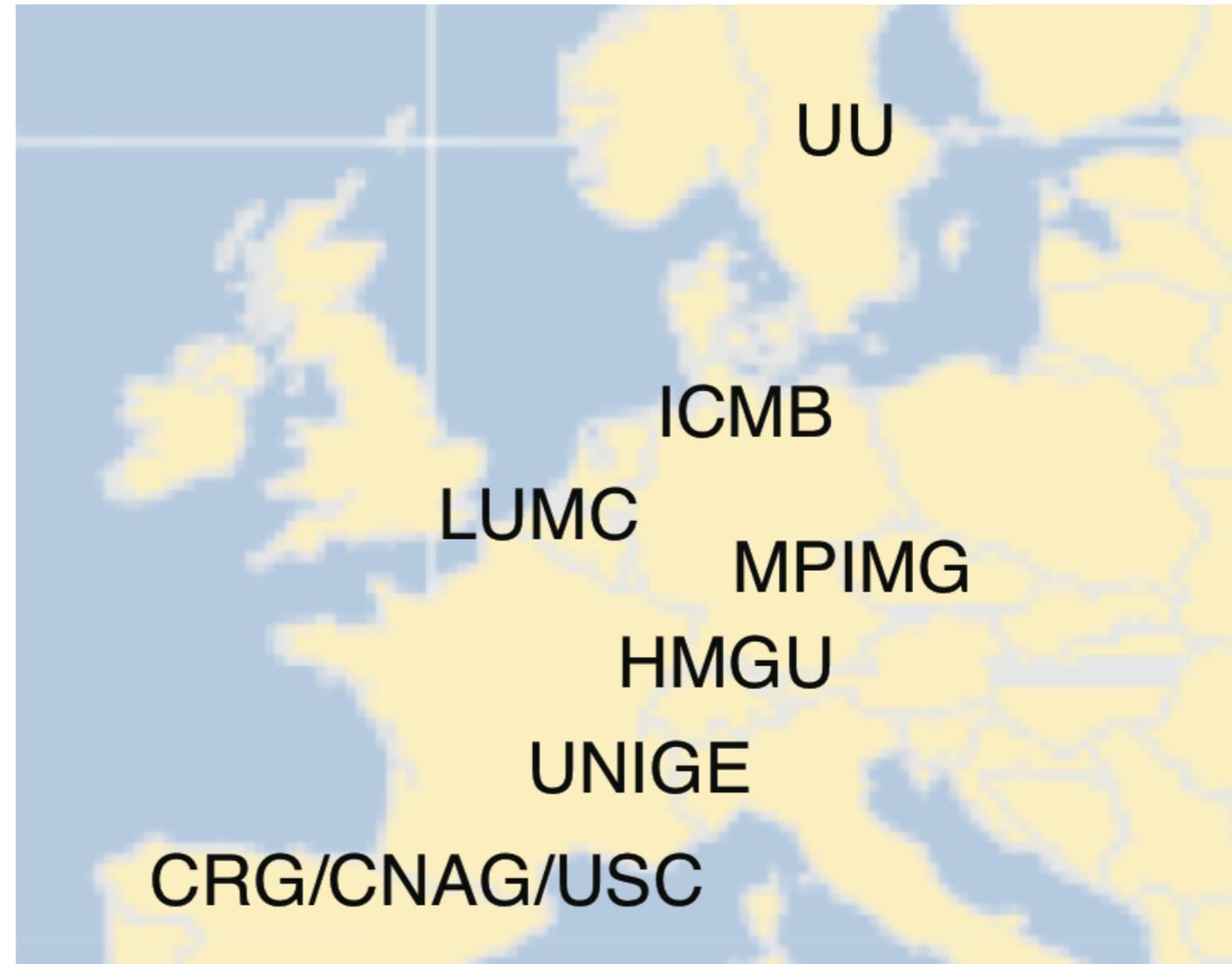


Sequencing labs

Transcriptome sequencing



7 laboratories
48-116 samples each



Quantitative trait locus (QTL) analysis

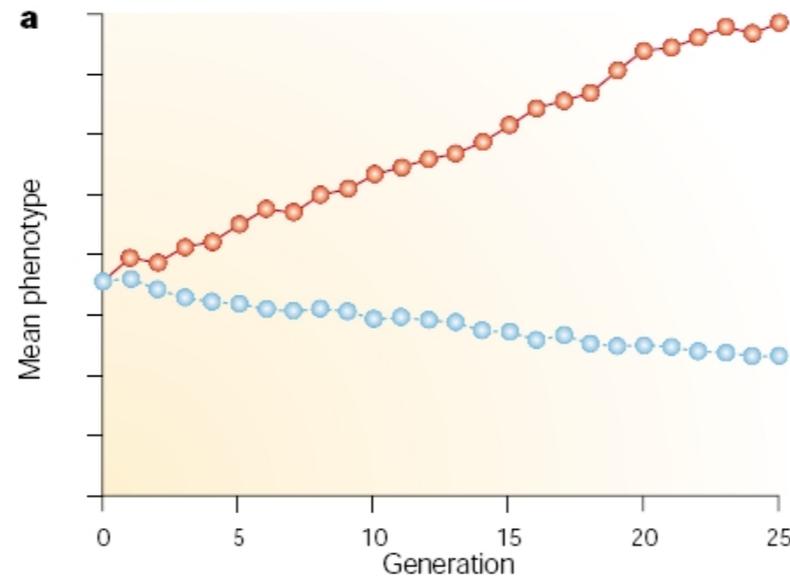
Links phenotypic and genotypic data

Stretches of DNA containing or linked to the **genes** that underlie a **quantitative trait** (phenotypes; can be attributed to polygenic effects)

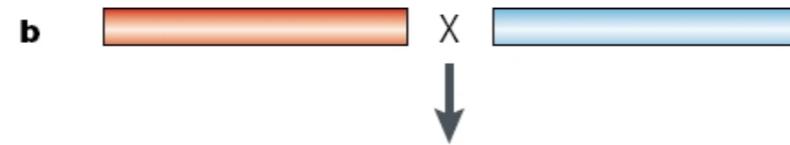
Expression QTLs (**eQTL**):

cis- and *trans*-controlling elements for the **expression** of **genes**

Quantitative trait locus (QTL) analysis



Traits



Parental lines

Analysis

Mapped *cis*- QTLs to transcriptome traits of protein-coding and miRNA genes

Separately for European and Yoruba (Nigeria)

***cis*-regulatory element**

region of DNA or RNA that regulates the expression of genes located on that same molecule of DNA

***trans*-acting element**

is usually a DNA sequence containing a gene; resulting protein (or microRNA) is used in the regulation of another target gene

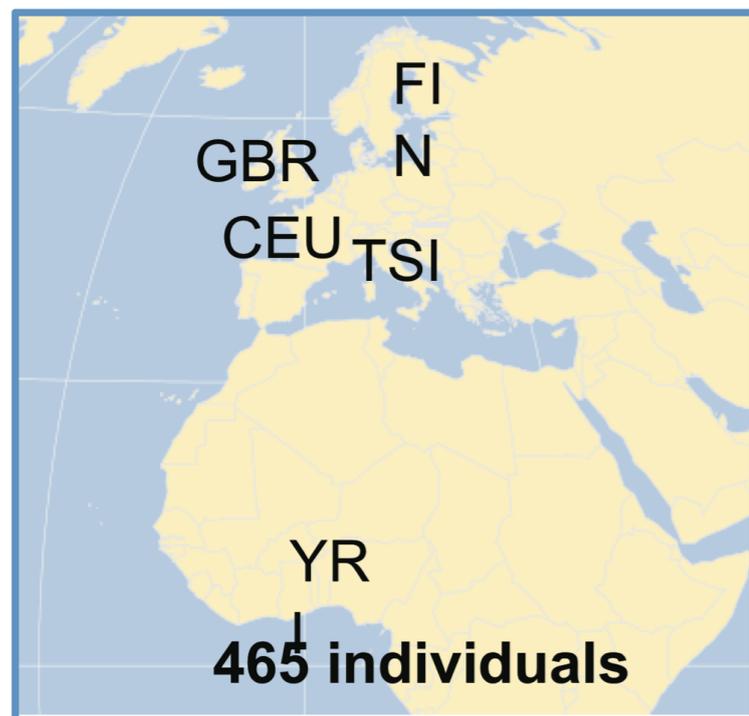
Transcriptome variation in populations

Shown in

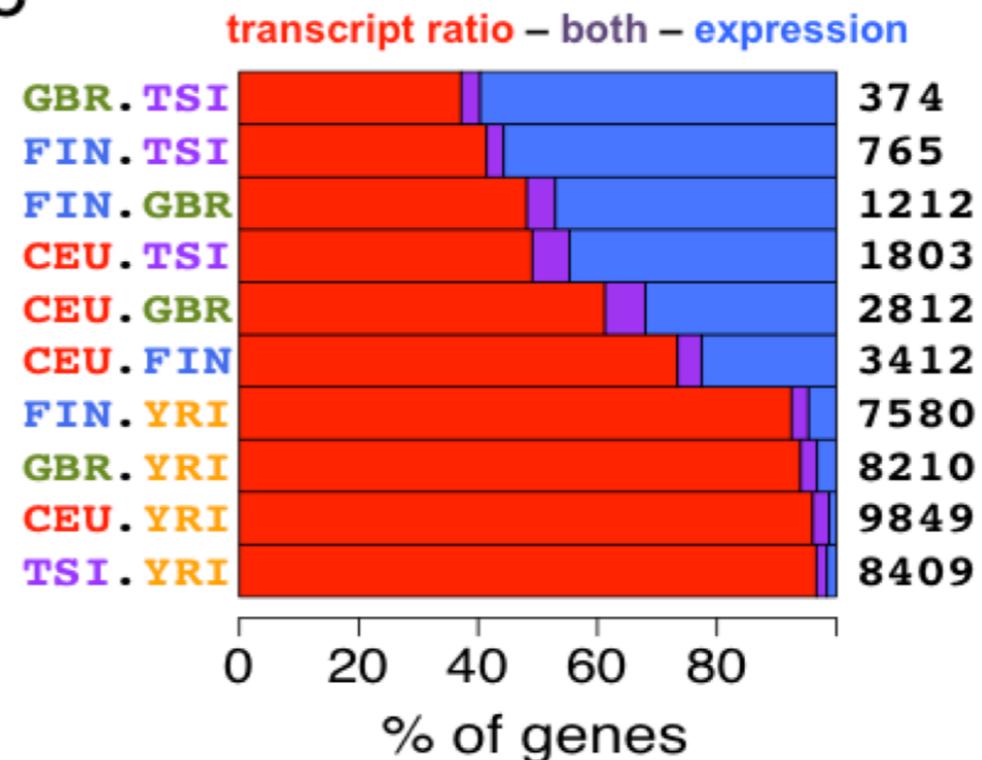
- overall expression levels
- relative abundance of transcripts from the same gene

263 - 4,379 genes with **differential expression** or **transcript ratios** between population pairs

Continental tr differences higher than within Europe



C



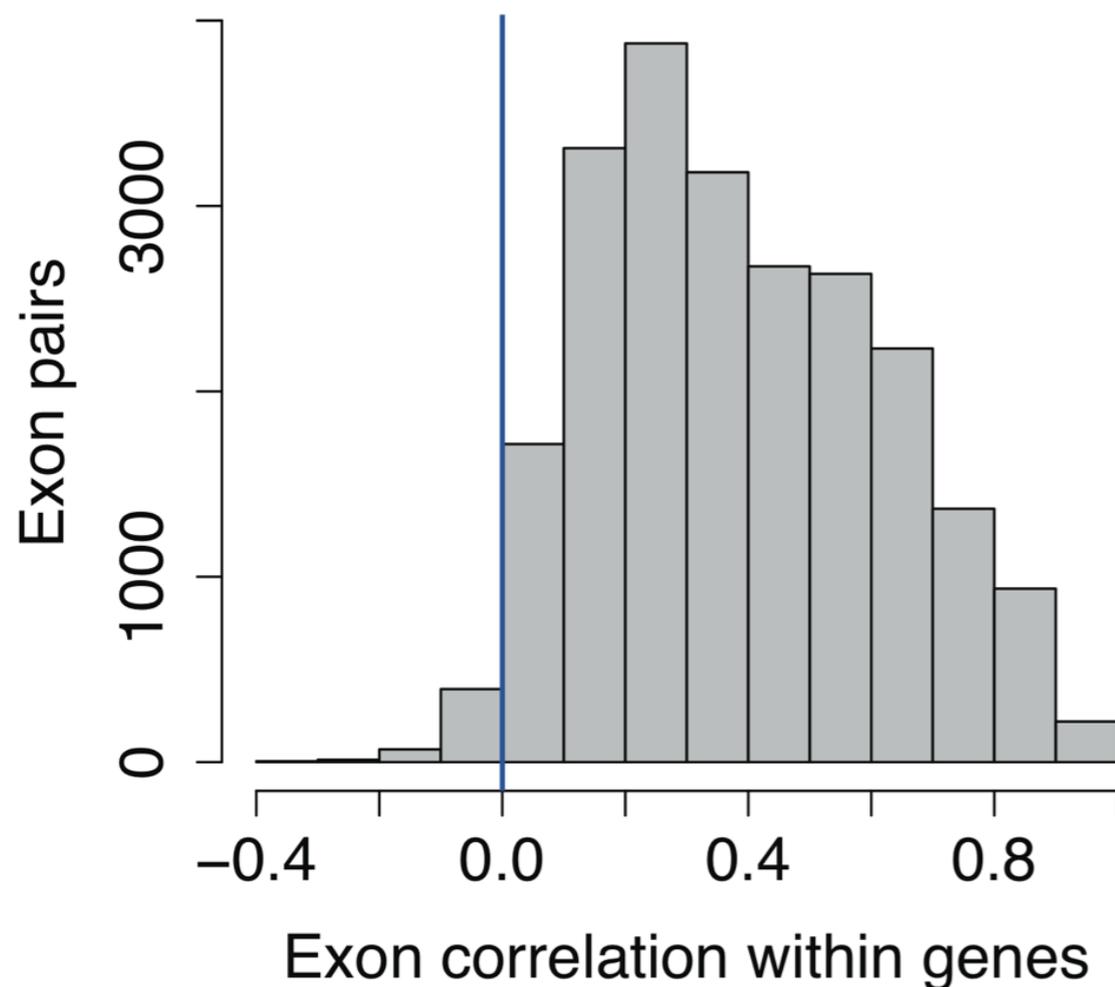
The proportion of differentially expressed and differentially spliced genes between population pairs using the DEXSeq method.¹⁰

eQTL analysis

eQTL analysis of **protein-coding** and **lincRNA** genes

-> **3,773** genes having "classical" eQTL

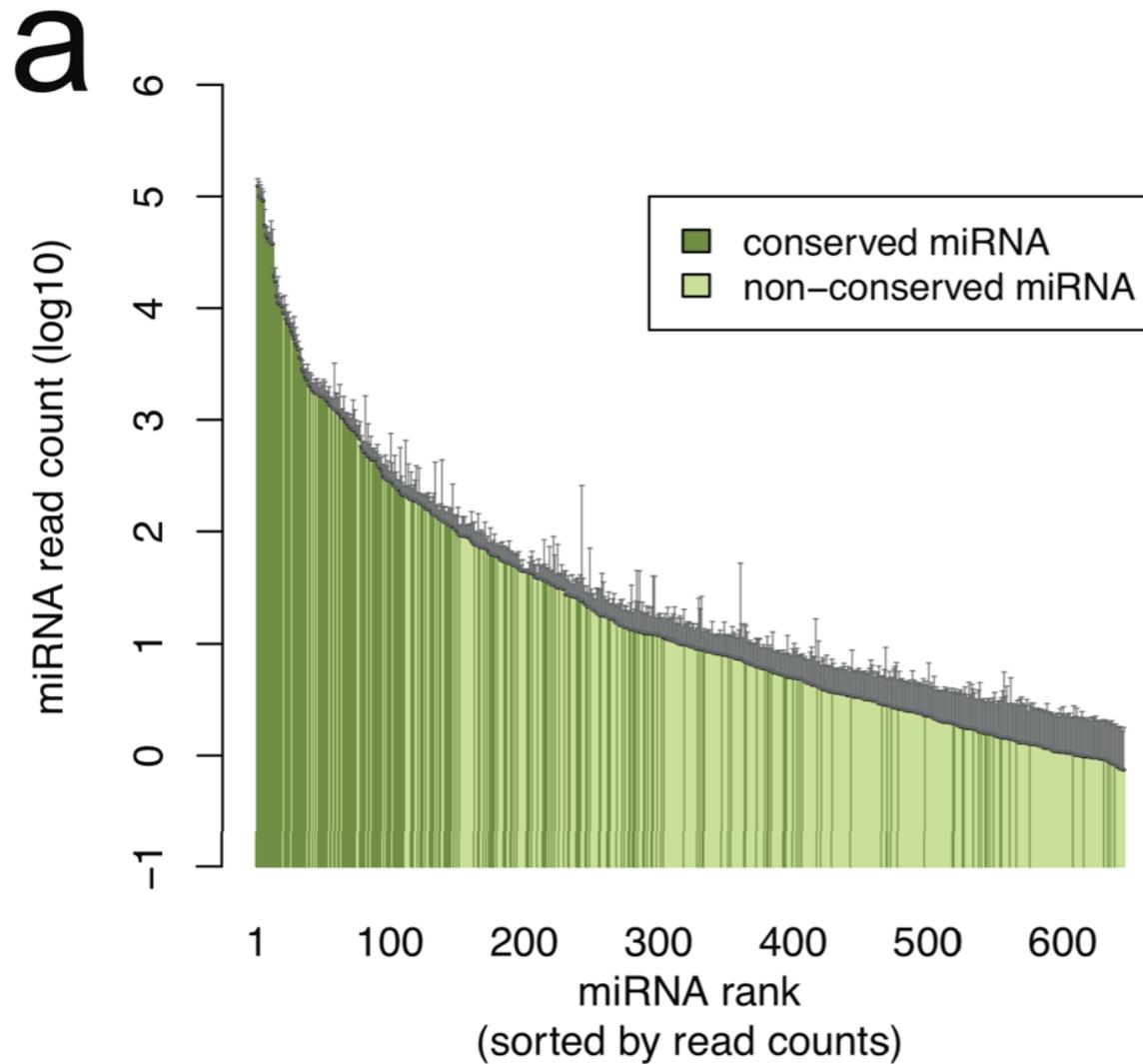
mapped eQTLs for exon quantifications that can capture **gene expression & splicing variations** -> **7,825** genes



Correlation between quantifications of exons from the same gene for chr20 in EU dataset.

Often correlation is not very high
-> frequent splicing variation

miRNA expression study



Expression of 644 autosomal miRNAs
detected in 452 individuals

60 (out of 644) have
significant *cis*-eQTLs

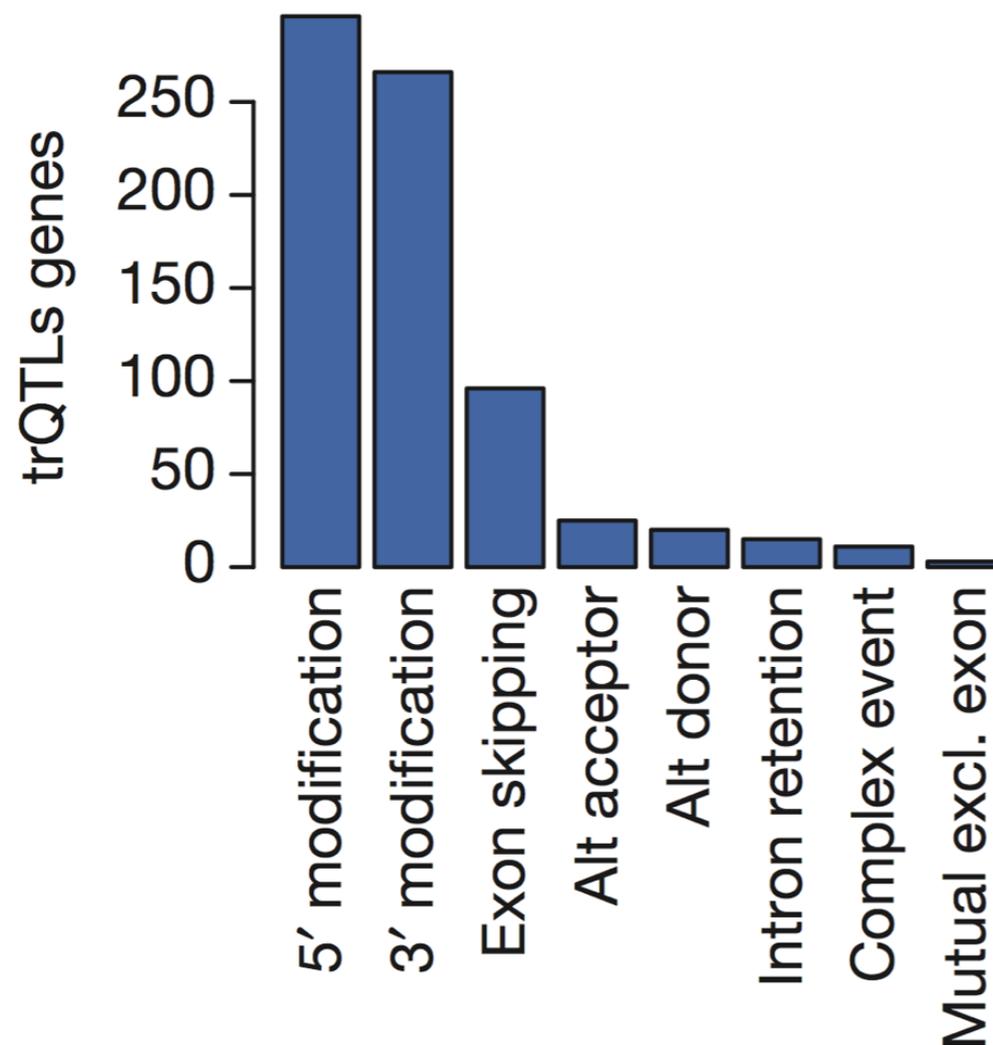
miRNA - mRNA effect

32 (out of 100): expression
correlated with target exons
-> miRNA downregulate
genes

transcript ratio QTLs (trQTLs)

Investigate genetic effects on **splicing**

Discovered **639** genes with transcript ratio QTLs (trQTLs) affecting the ratio of each **transcript** to the **gene total**



Classification of changes
caused by trQTLs

transcript ratio QTLs (trQTLs)

Investigate genetic effects on **splicing**

Discovered **639** genes with transcript ratio QTLs (**trQTLs**) affecting the ratio of **transcript** to the **gene total**

Identified **279** genes where **eQTL** and **trQTL** causal variants are **independent** (in most genes) -> transcriptional activity and transcript usage are controlled by different regulatory elements

Properties of regulatory variants

Compared properties of **top** (most significant) **eQTL** variant per gene to a **null of non-eQTL** variants

InDels enriched -> more likely to have effect (vs SNPs)

eQTLs

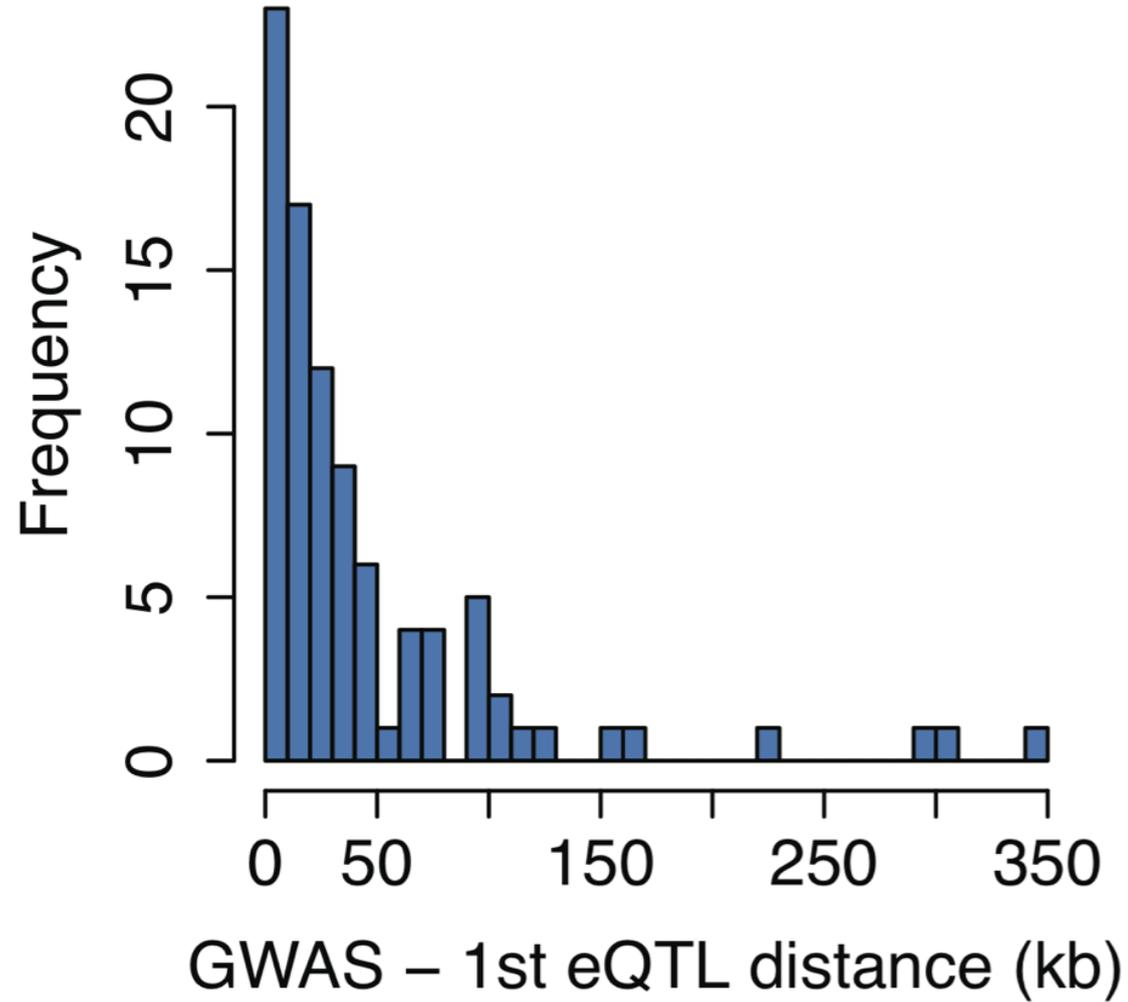
- non-coding elements | transcription factor peaks
- Dnase1 hypersensitive sites
- chromatin states of active promoters & enhancers
- splice sites
- non-synonymous enrichments

trQTLs

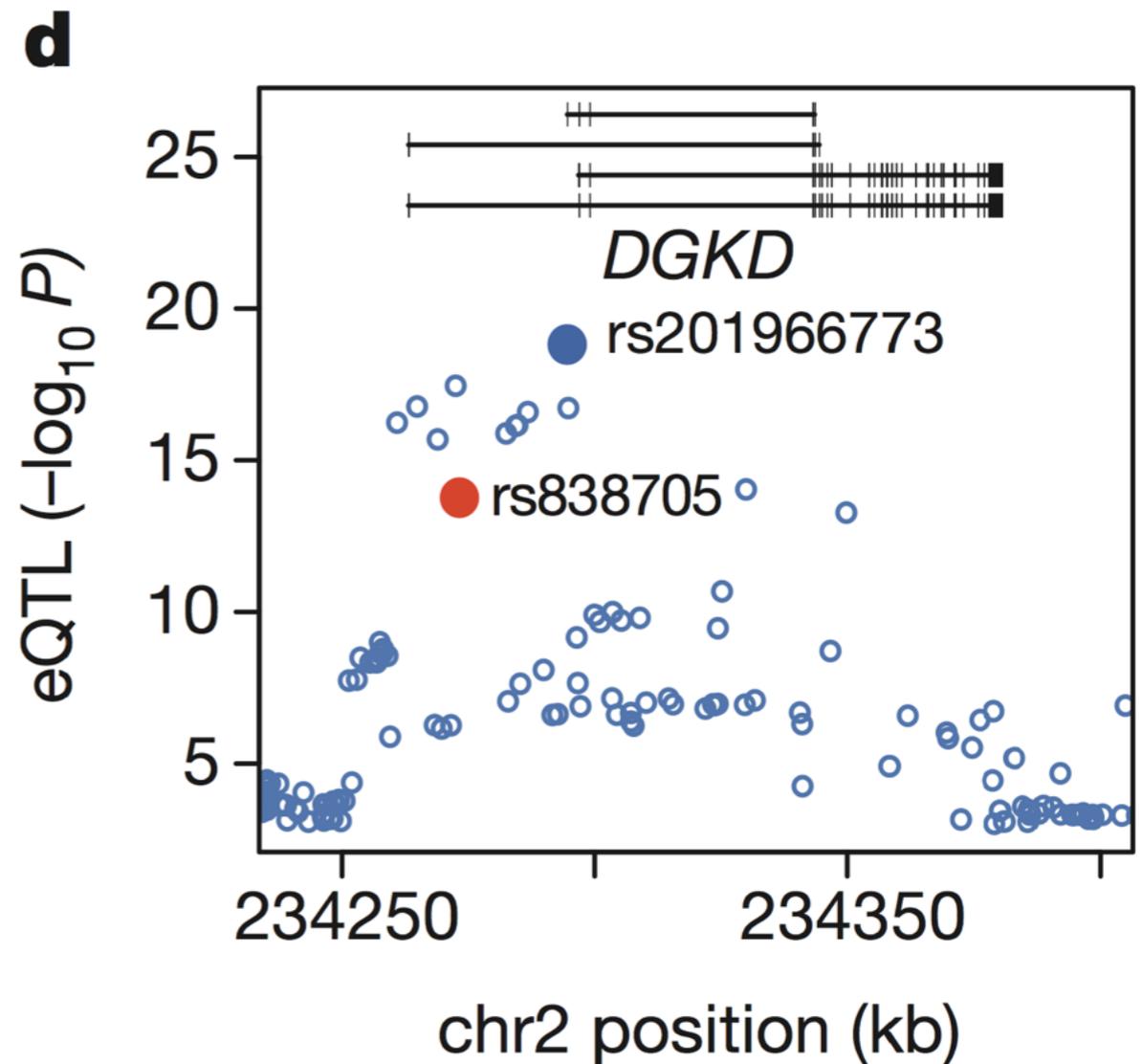
- splices sites | 3' UTRs | promoters

GWAS SNPs - eQTL

78 eQTL regions that are likely causal signals for 91 GWAS SNPs



How close the GWAS variant is to best EUR eQTL



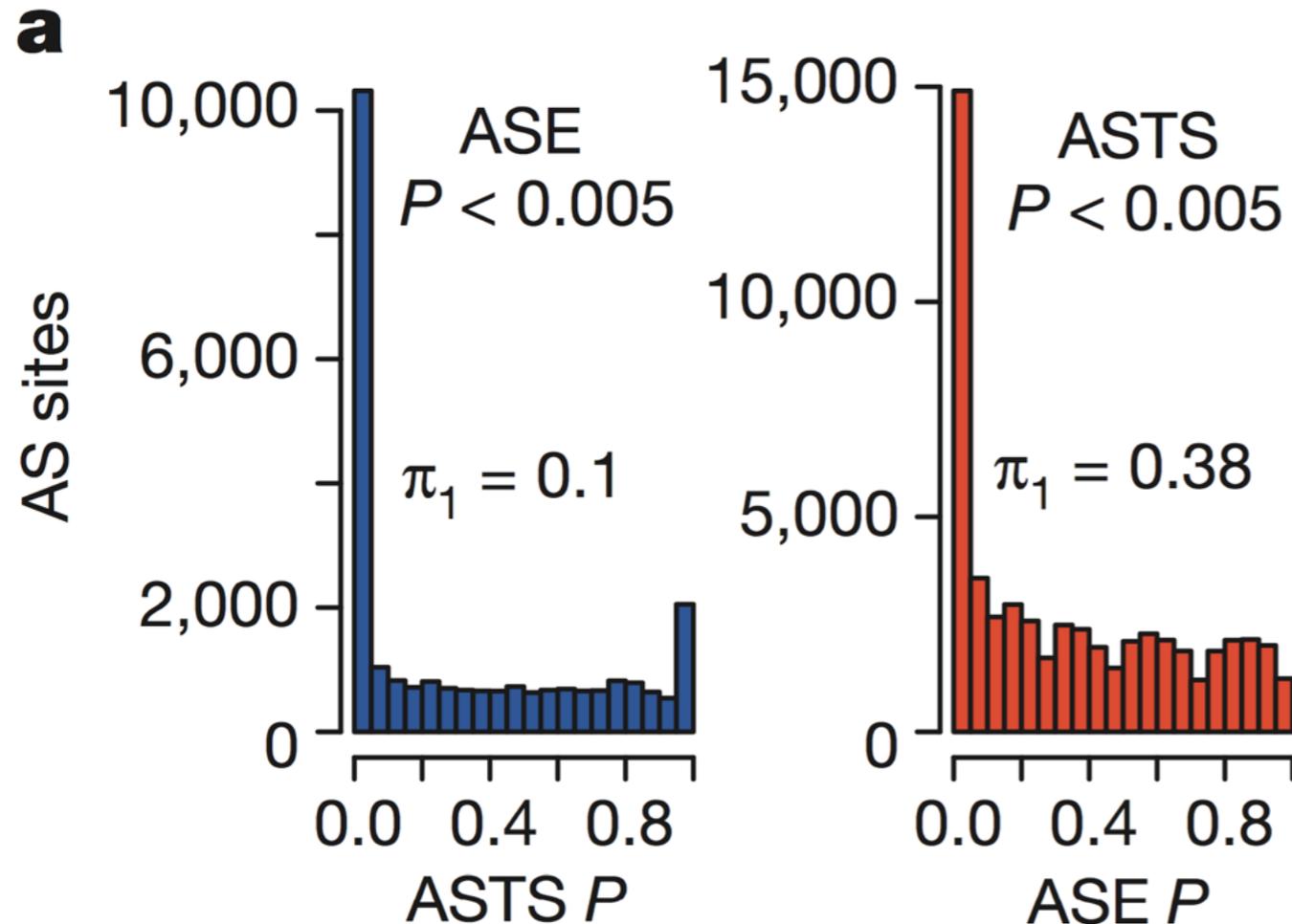
Intronic SNP is associated with Ca levels (red), and the top eQTL variant (blue) is a very likely causal variant

Allelic effects

ASE equally common as ASTS

ASE seems to be genetically driven rather than epigenetic effects

ASE may be driven by ASTS



Distribution of ASTS p-value of sites with significant ASE and vice versa

Summary

Genetic regulation is very common

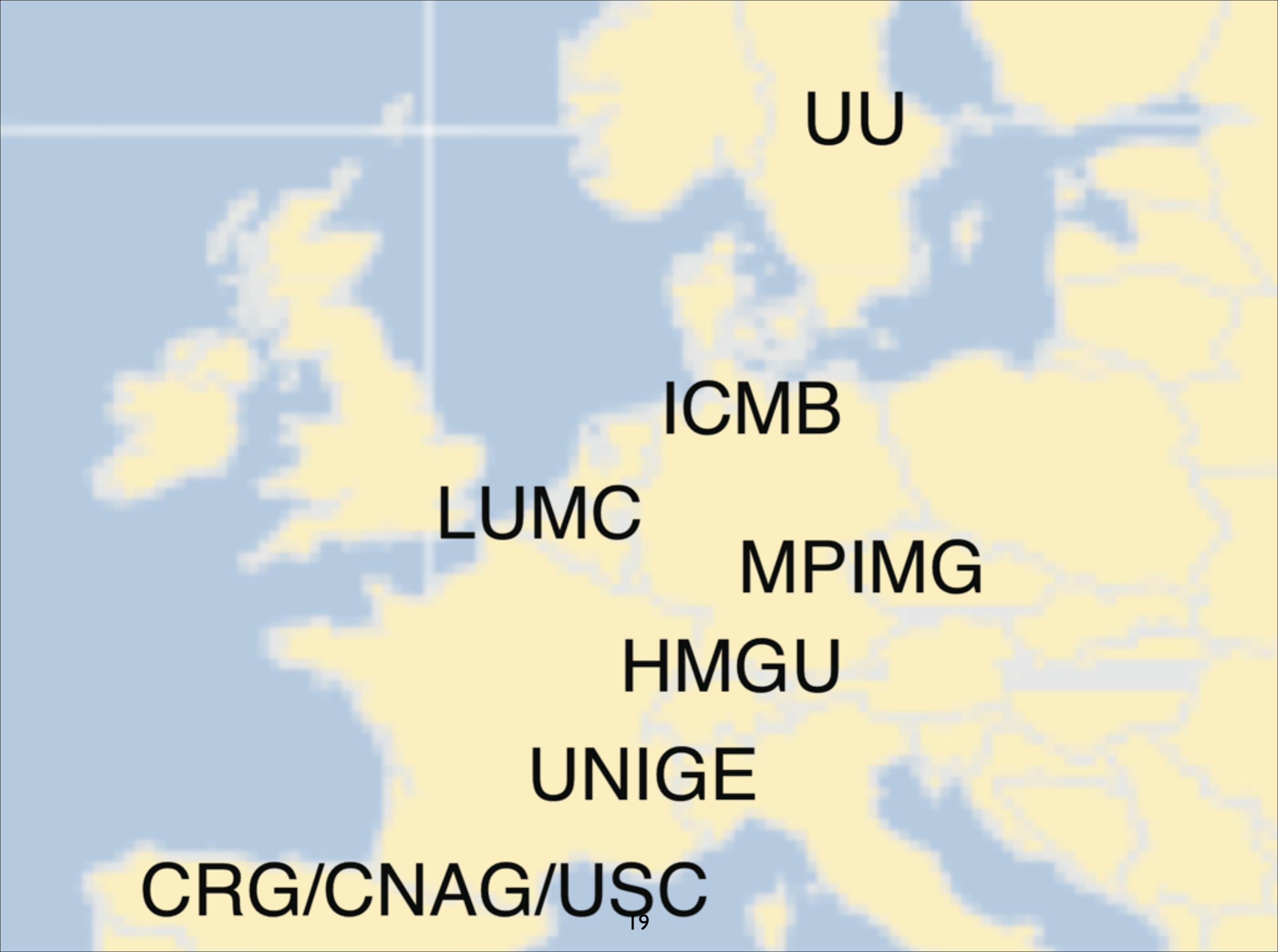
- over **one half of genes** affected by **eQTLs**
- major determinant of allelic expression

16% of known **GWAS** variants are **eQTLs**

- but by chance overlap is 11%

However, with **statistical methods** **eQTL** can be used to **identify causal GWAS** variant

eQTLs and **ASE** appear common but independent



UU

ICMB

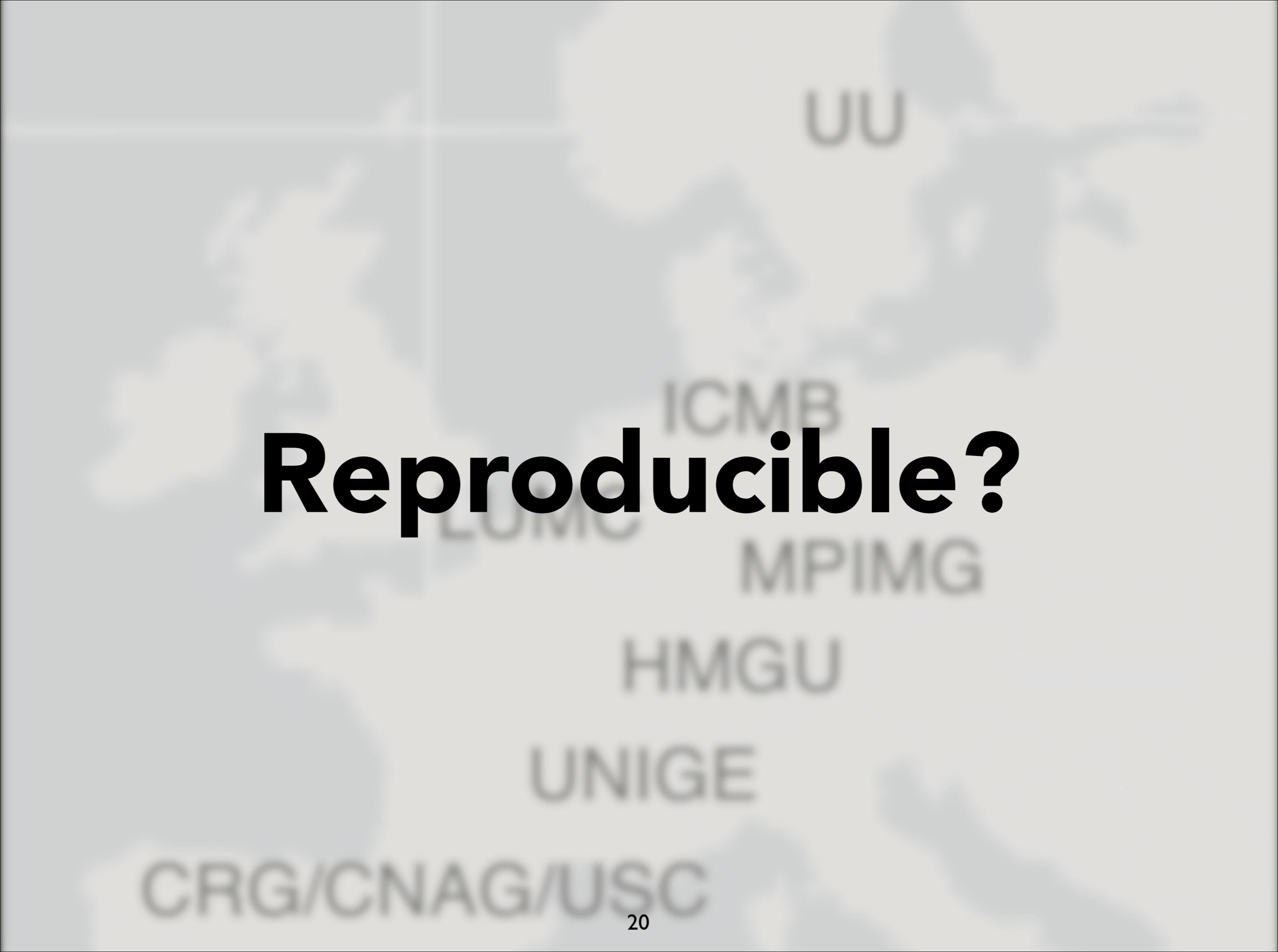
LUMC

MPIMG

HMGU

UNIGE

CRG/CNAG/USC



Reproducible?

UU

ICMB

LUMIC

MPIMG

HMGU

UNIGE

CRG/CNAG/USC

Setup

Illumina **HiSeq 2000**

Illumina's TruSeq kits for preparation

mRNA - 75bp PE

sRNA - 36bp SE

5 RNA samples sequenced **at all sites**

→ for comparison

Software

GEM (mRNA)

miraligner (sRNA)

Quality check

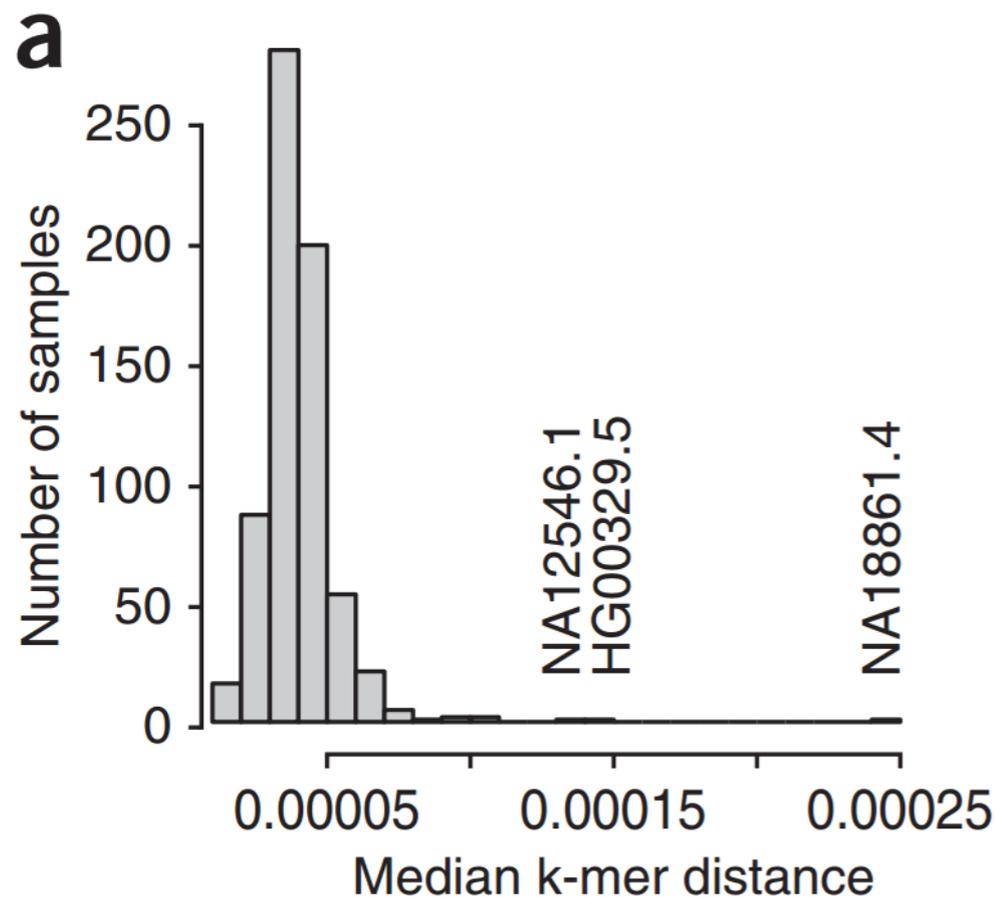
k-mer distance using abundance in samples

promising & does not require mapping

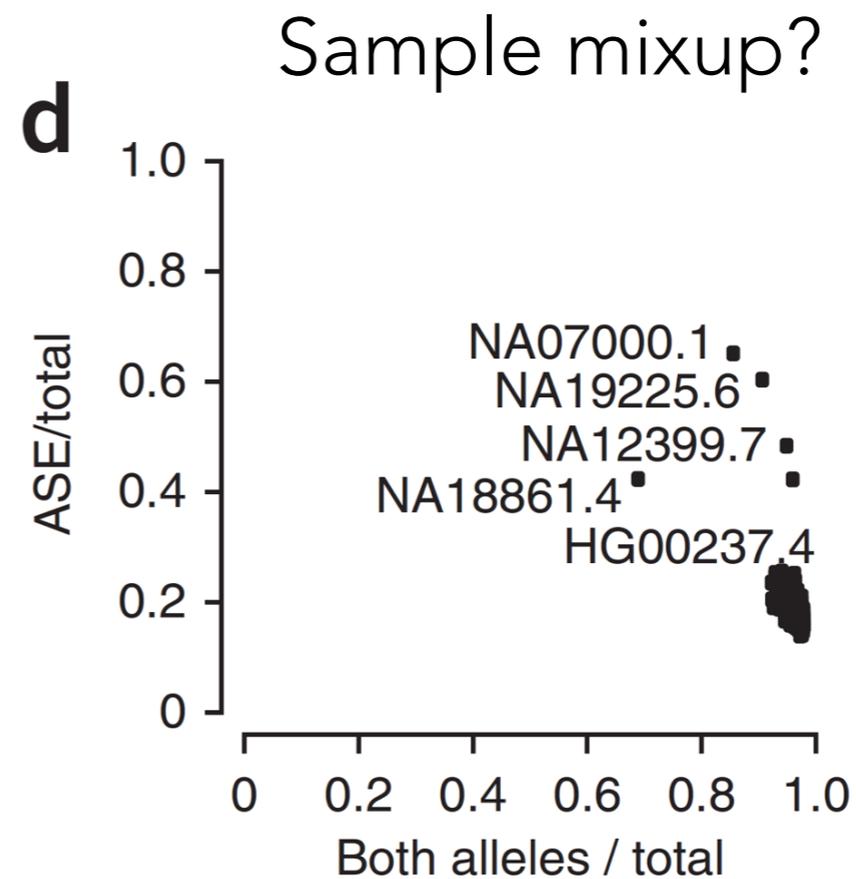
Exon/transcript expression correlation between samples

→ low correlation = outlier

Allele-specific expression



with $k=9$; pair wise distance



For all heterozygous sites: proportion of het SNPs where both alleles were observed vs proportion of ASE

Sources of variation

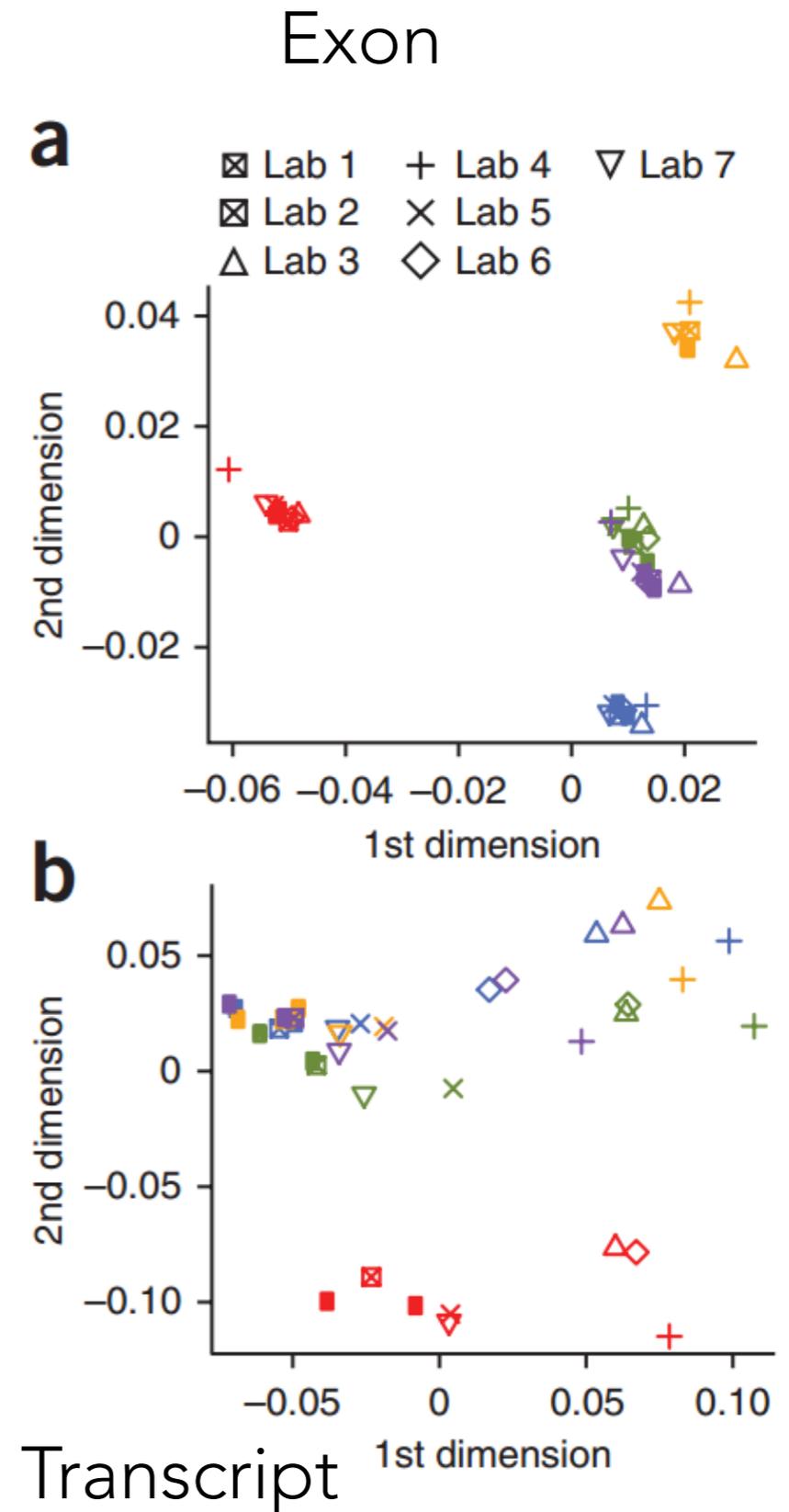
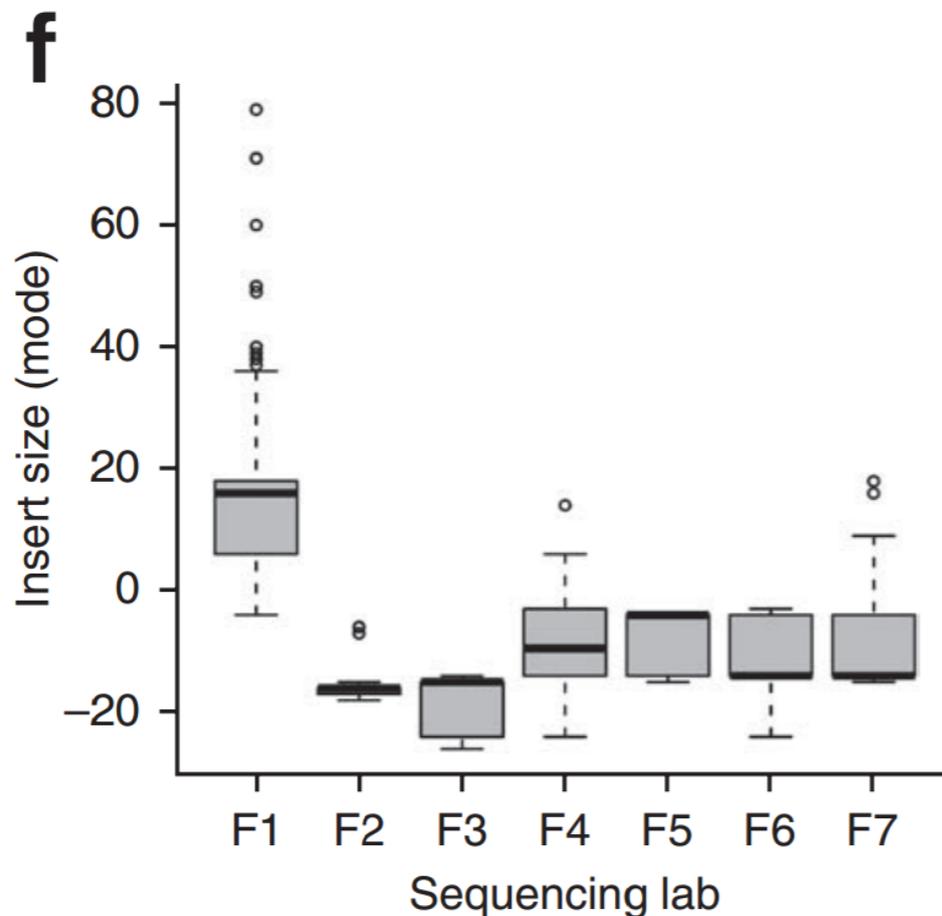
Samples clustered by sample and not lab

Clustering for **exon** quantification

stronger than for **transcript**

Lib prep difference in **GC percentage**

Smaller inserts than targeted



Correction for variation in mRNA

Used PEER

takes quantification of genes (or transcripts) and uses factor analysis-based methods to infer factors explaining variance.

After PEER

clustering by lab **less pronounced**

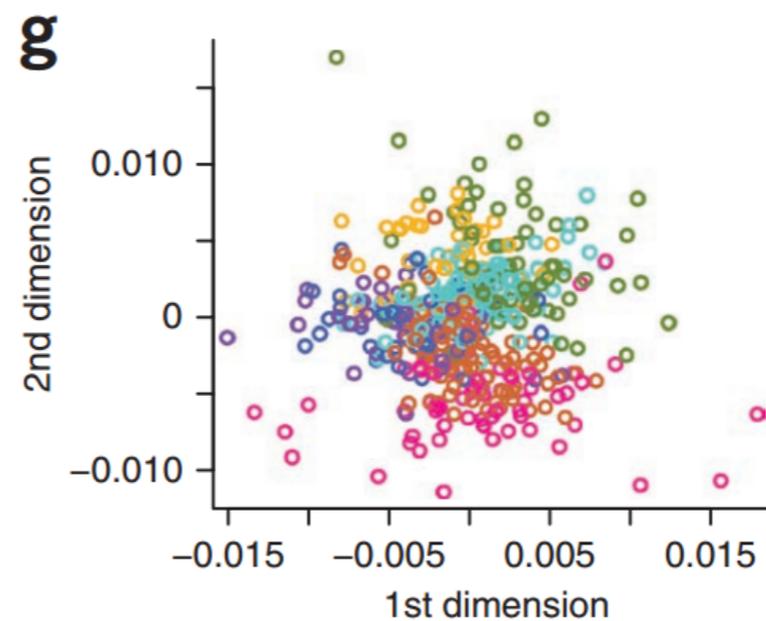
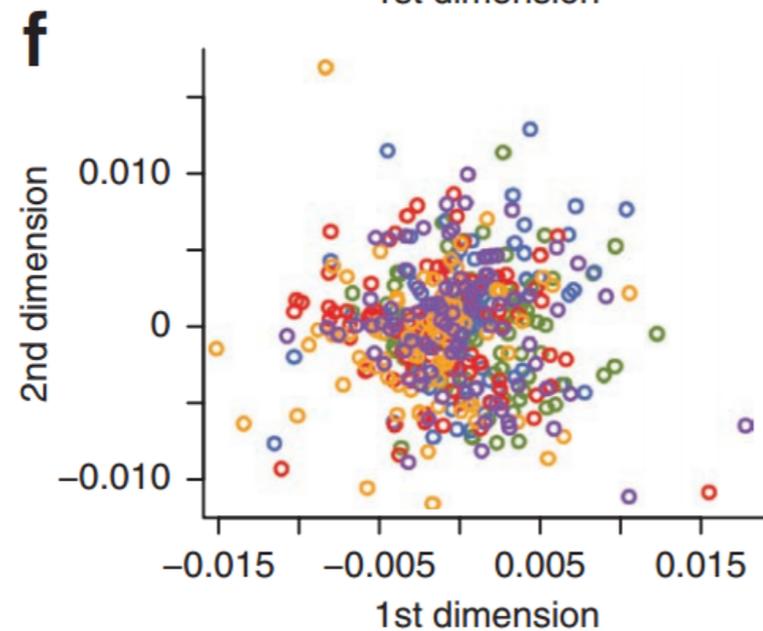
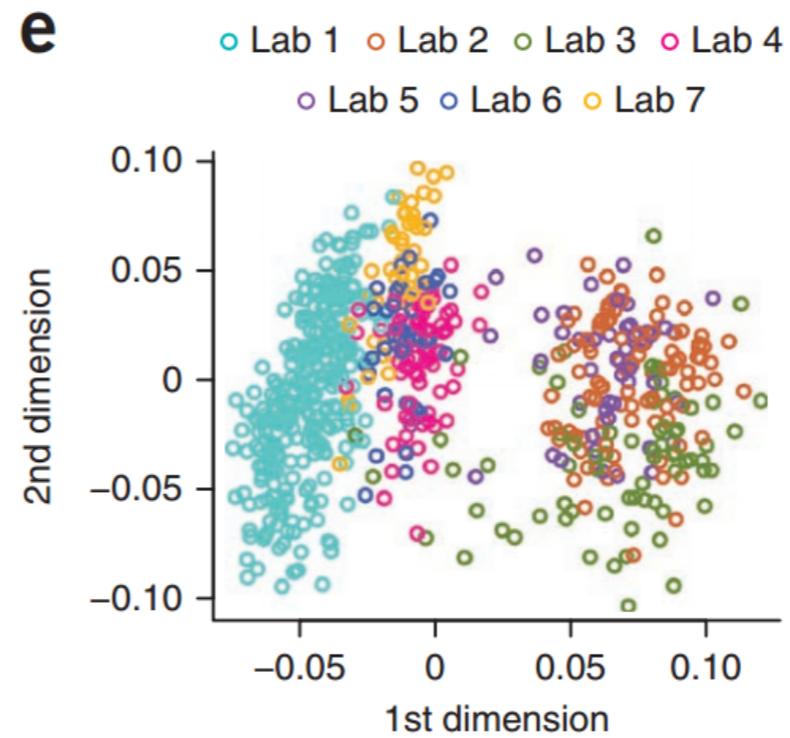
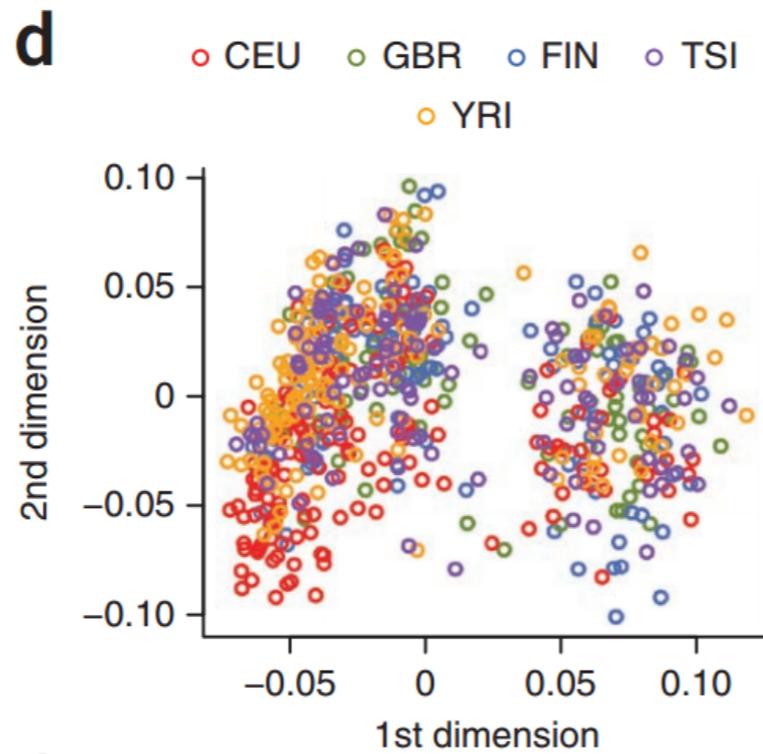
technical variation can be properly accounted for

Stegele et al. Using probabilistic estimation of expression residuals (PEER) to obtain increased power and interpretability of gene expression analyses. Nat Protoc 2012

Plots of transcript quantification

By population

By laboratory



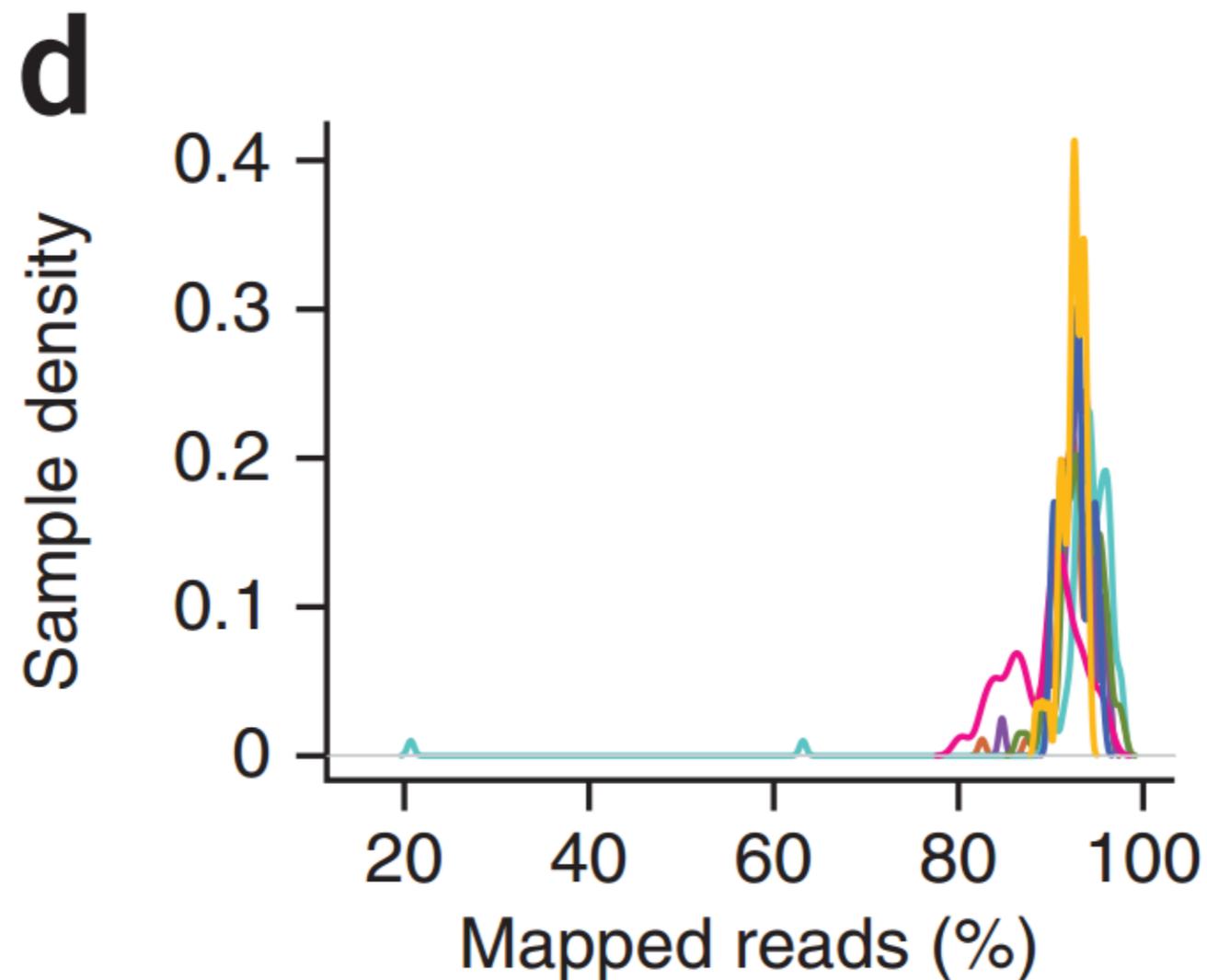
Before PEER

After PEER

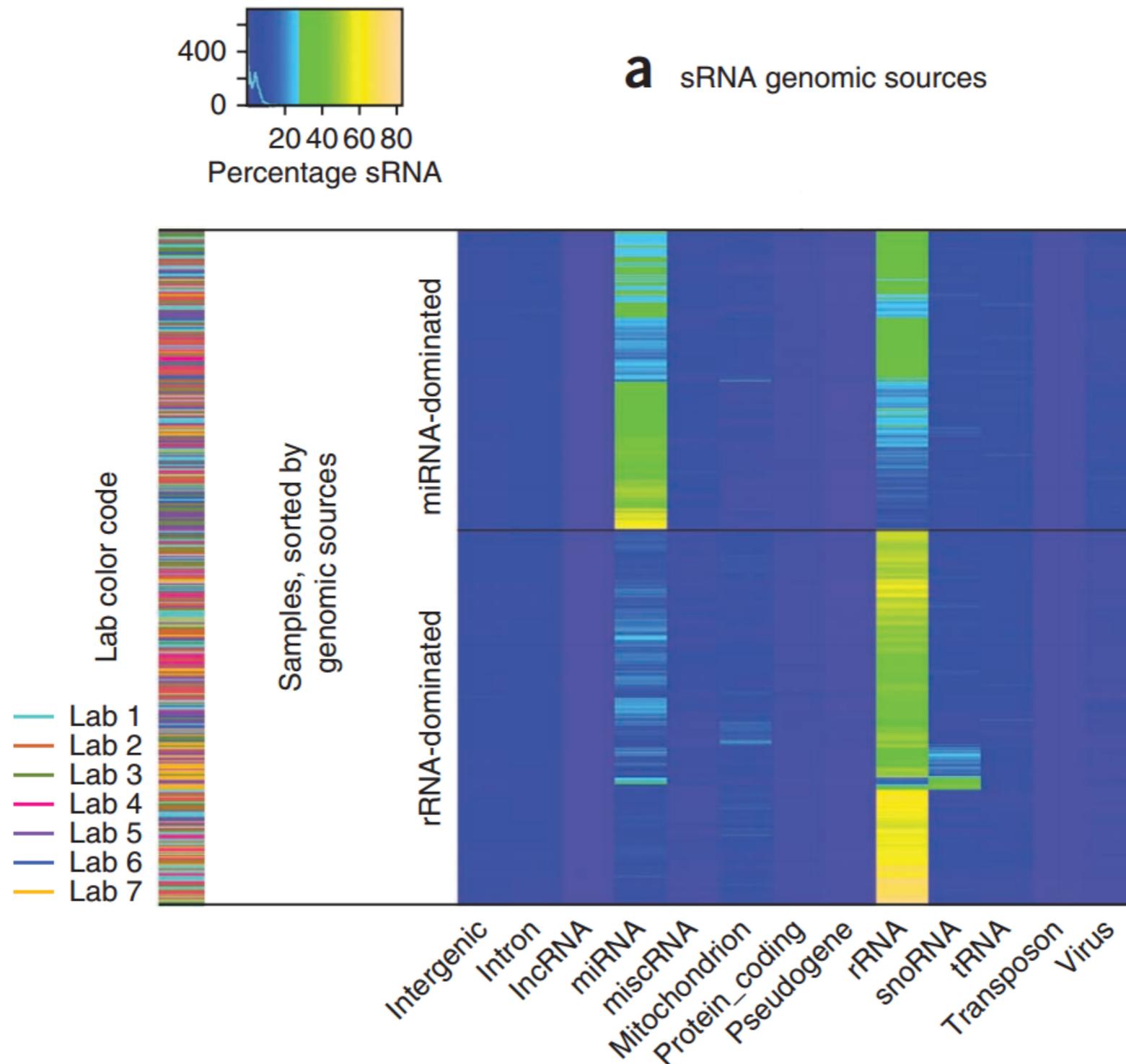
sRNA sequencing

Removed reads <18bp

Sequencing quality & mapping quality high



Heatmap clustered by expression of sRNA sources



sRNA sequencing

Removed reads <18bp

Sequencing quality & mapping quality high

Origins from miRNA and rRNA

differences in sRNA likely due to RNA extraction
(not labs)

Corrected by **PEER**

GC% biggest source of variation

Summary

Technical **variation** in RNA-seq is **small** (when)

- using exact **same protocols** for sample prep and sequencing kits

Slight variations in **GC%** and **insert size**

PEER algorithm

- account for & reduce technical factors (large studies needed)

Quality checks

Proposed quality checks

mRNA & sRNA sequencing

Distribution of **base quality** scores

Average and width of the distribution of **GC content**

Percentage of **reads mapping** to the genome

Checks for **sample swaps** and **contaminations**

Outlier detection: **pairwise correlations** in **expression** quantification between samples

Proposed quality checks

mRNA

Mean & s.d. of insert size

Percentage of reads mapping to annotated exons (at least 60 %)

5'–3' trends in coverage across transcripts

sRNA

Length distribution after adaptor clipping

Percentage of reads mapping to known sRNA genes

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Reproducibility of high-throughput mRNA and small RNA sequencing across laboratories