

# Genetic architecture of transcriptome regulation and orthogonal tissue decomposition

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2015-05-19 16:14:53 <sup>1</sup>Department of Medicine, University of Chicago, <sup>2</sup>Committee on Genetics, Genomics, and Systems Biology, University of Chicago, <sup>3</sup>Division of Genetic Medicine, Vanderbilt University

## Abstract

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## Introduction

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## Results

### Local genetic variation explains a large proportion of gene expression variance

We estimated the heritability of gene expression in whole blood from the Depression Genes and Networks (DGN) cohort (n=922) [1] using a mixed-effects model (see Materials and Methods) and calculated variances using restricted maximum likelihood as implemented in GCTA [2]. We fit a joint model with a local and a global genetic relationship matrix (GRM). The local GRM was derived from SNPs within 1 Mb of each gene and the global GRM was derived from SNPs that are located on non-gene chromosomes and are eQTLs in the Framingham Heart Study (FHS) cohort (n=5257, FDR < 0.05) [3]. The mean local  $h^2$  was 0.130 and 54.6% of genes had a positive 95% confidence interval (CI), while the mean global  $h^2$  was 0.076 and just 4.2% of genes had a positive CI (Fig 1). The maximum local  $h^2$  was 0.93 with a standard error (SE) of 0.009 while the maximum global  $h^2$  was 0.91 with a SE of 0.16. Similar results were observed for the 1194 genes with *trans*-eQTLs (FHS FDR < 0.05) when the global GRM was limited to known *trans*-eQTLs (Fig 2). That is,

the mean local  $h^2$  was 0.133 and 61.3% of genes had a positive 95% confidence interval (CI), while the mean *trans*  $h^2$  was just 0.021 and 4.2% of genes tested had a positive CI.

## The effect of local genetic variation on gene expression is sparse rather than polygenic

## Cross-tissue and tissue-specific gene expression by orthogonal tissue decomposition

## Citations

The relationship was first described by Reference 4. However, there are also opinions that the relationship is spurious [5]. We used R for our calculations [6], and we used package `knitcitations` [7] to make the bibliography.

## Discussion

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## Methods

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## Equations

The deterministic part of the model is defined by this **in-line equation** as  $\mu_i = \beta_0 + \beta_1 x$ , and the stochastic part by the **centered equation**:

$$\frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu_i)^2/(2\sigma^2)}$$

## Tables

## Warning: package 'knitr' was built under R version 3.1.3

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.02	0.10	0.26	0.8
x	2.01	0.09	21.46	0.0

Table 1: This is a GLM summary table.

## Figures

## Supplemental Figures

## References

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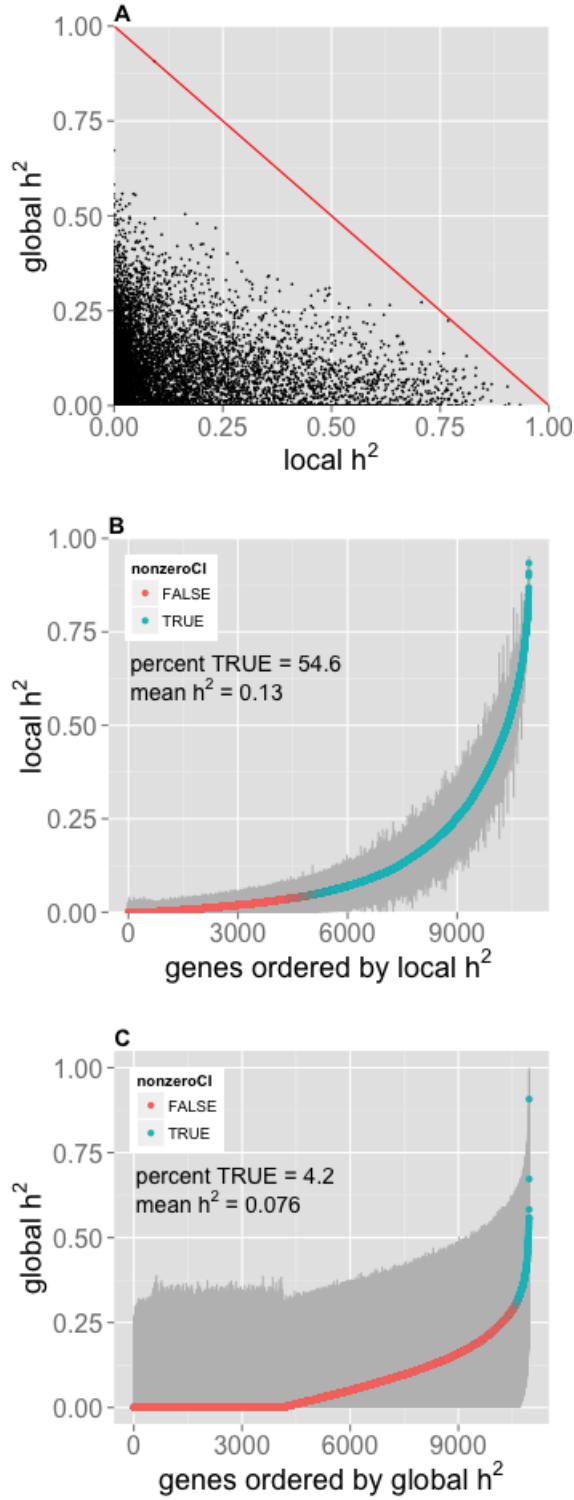


Figure 1: DGN whole blood expression joint heritability ( $h^2$ ). Local (SNPs within 1 Mb of each gene) and global (SNPs that are eQTLs in the Framingham Heart Study on other chromosomes [FDR < 0.05])  $h^2$  for gene expression were jointly estimated. **(A)** Global  $h^2$  compared to local  $h^2$  per gene. **(B)** Local and **(C)** global gene expression  $h^2$  estimates ordered by increasing  $h^2$ . The 95% confidence interval (CI) of each  $h^2$  estimate is in gray and genes with a lower bound greater than zero are in blue.

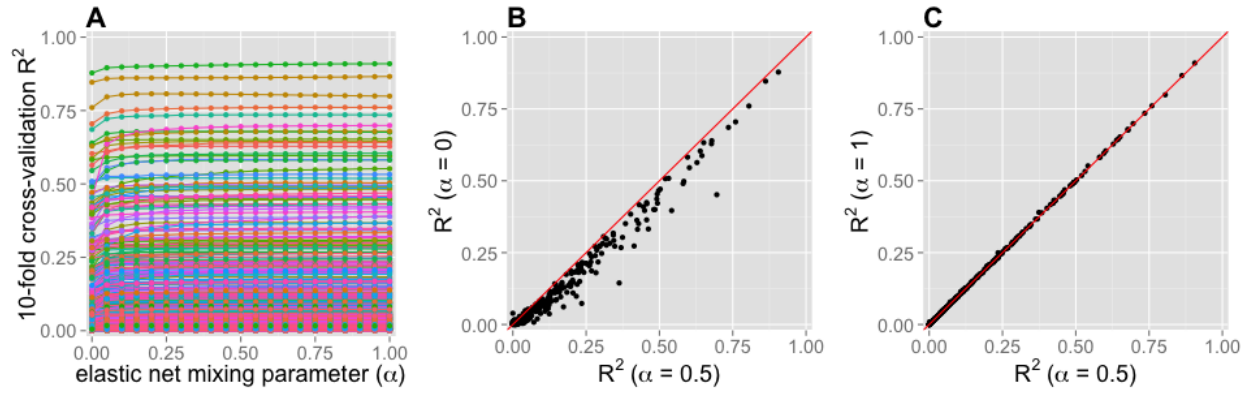


Figure 2: Cross-validated predictive performance across the elastic net. **(A)** 10-fold cross-validated  $R^2$  of predicted vs. observed expression in DGN whole blood compared to a range of elastic net mixing parameters ( $\alpha$ ) for 341 genes on chromosome 22. **(B)** Predictive  $R^2$  for  $\alpha = 0$  (ridge regression) compared to  $\alpha = 0.5$ . **(C)** Predictive  $R^2$  for  $\alpha = 1$  (lasso) compared to  $\alpha = 0.5$ .

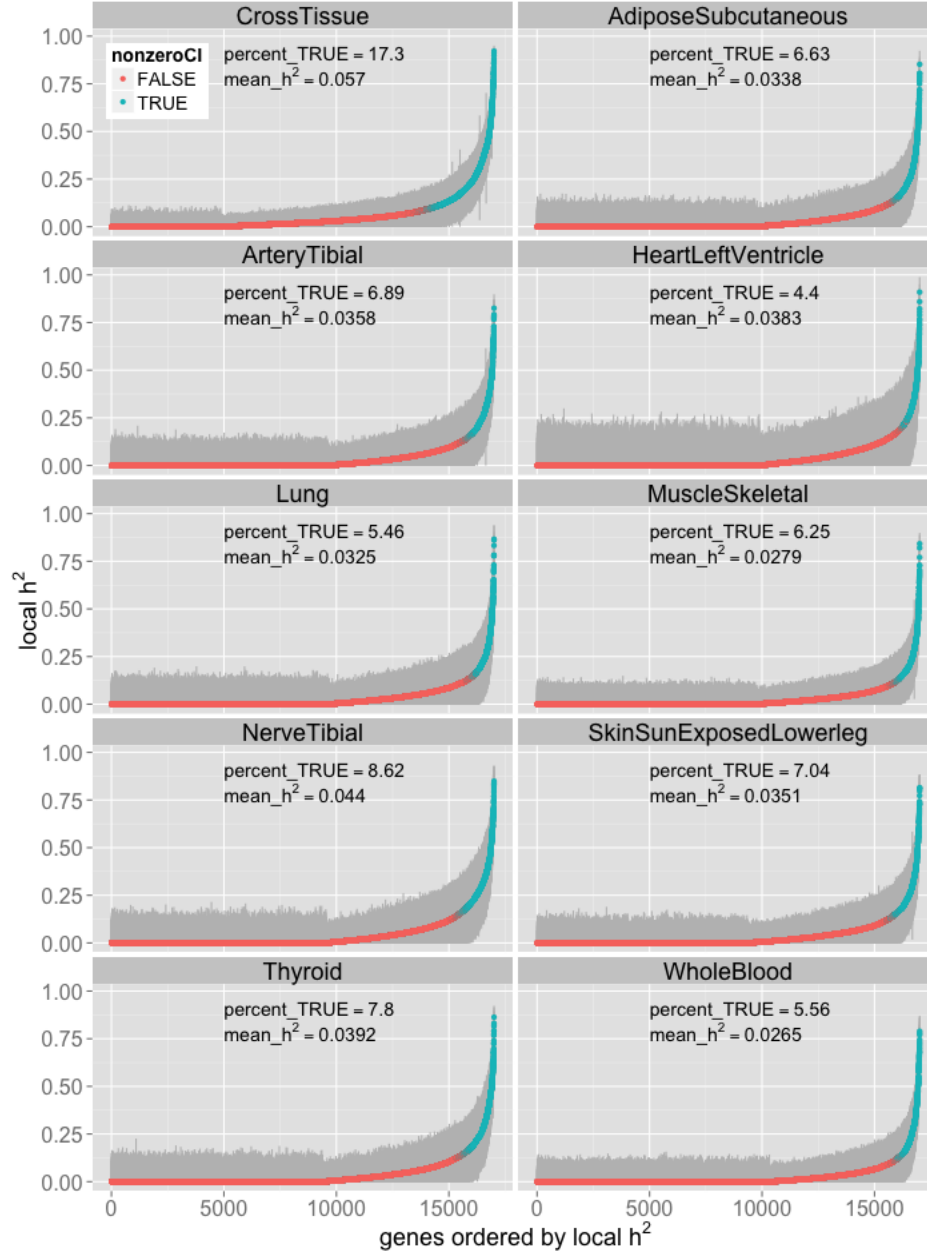


Figure 3: Cross-tissue heritability ( $h^2$ ) compared to tissue-wide  $h^2$ . Cross-tissue local  $h^2$  is estimated using the cross-tissue component (random effects) of the mixed effects model for gene expression and SNPs within 1 Mb of each gene. Tissue-wide local  $h^2$  is estimated using the measured gene expression for each respective tissue and SNPs within 1 Mb of each gene.

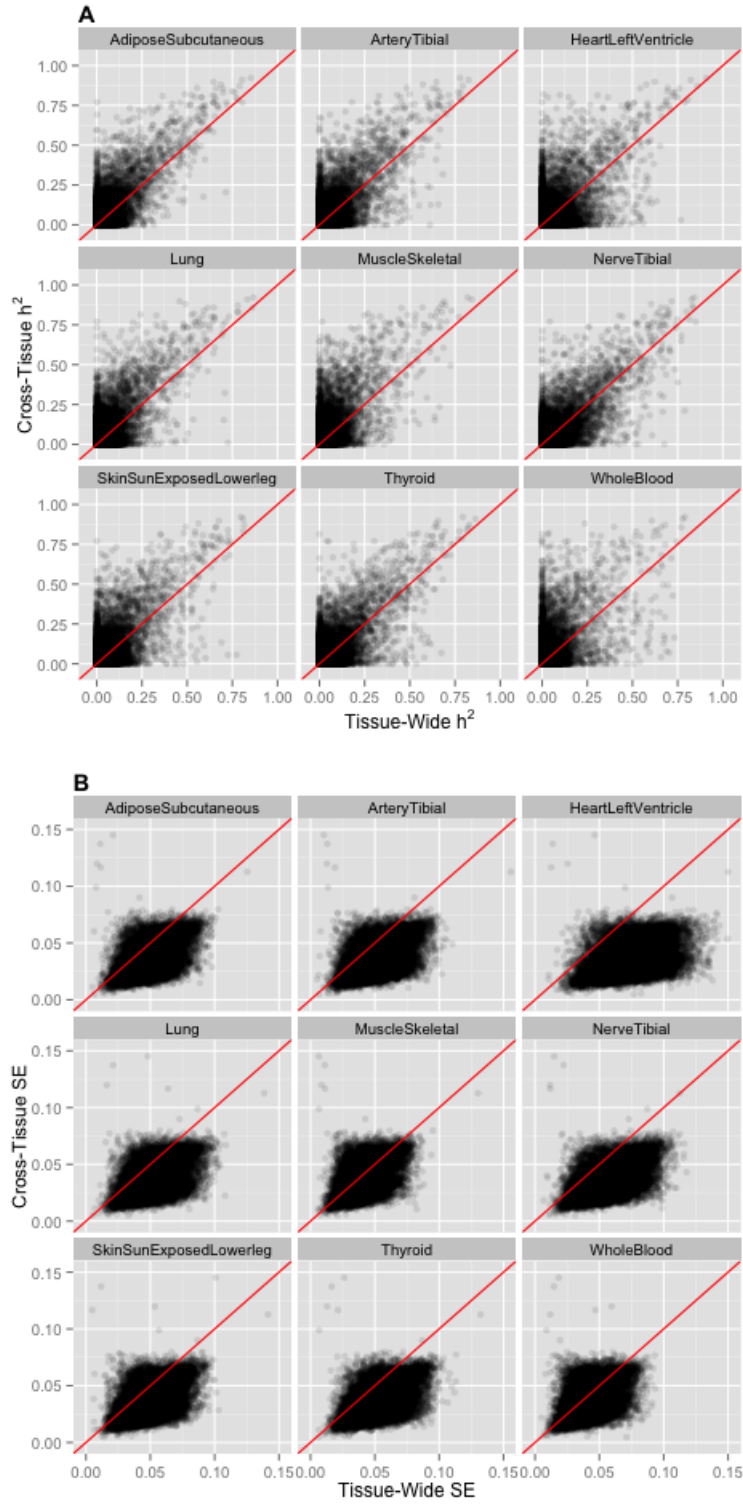


Figure 4: Cross-tissue and tissue-wide comparison of heritability ( $h^2$ , **A**) and standard error (SE, **B**). Cross-tissue local  $h^2$  is estimated using the cross-tissue component (random effects) of the mixed effects model for gene expression and SNPs within 1 Mb of each gene. Tissue-wide local  $h^2$  is estimated using the measured gene expression for each respective tissue and SNPs within 1 Mb of each gene.

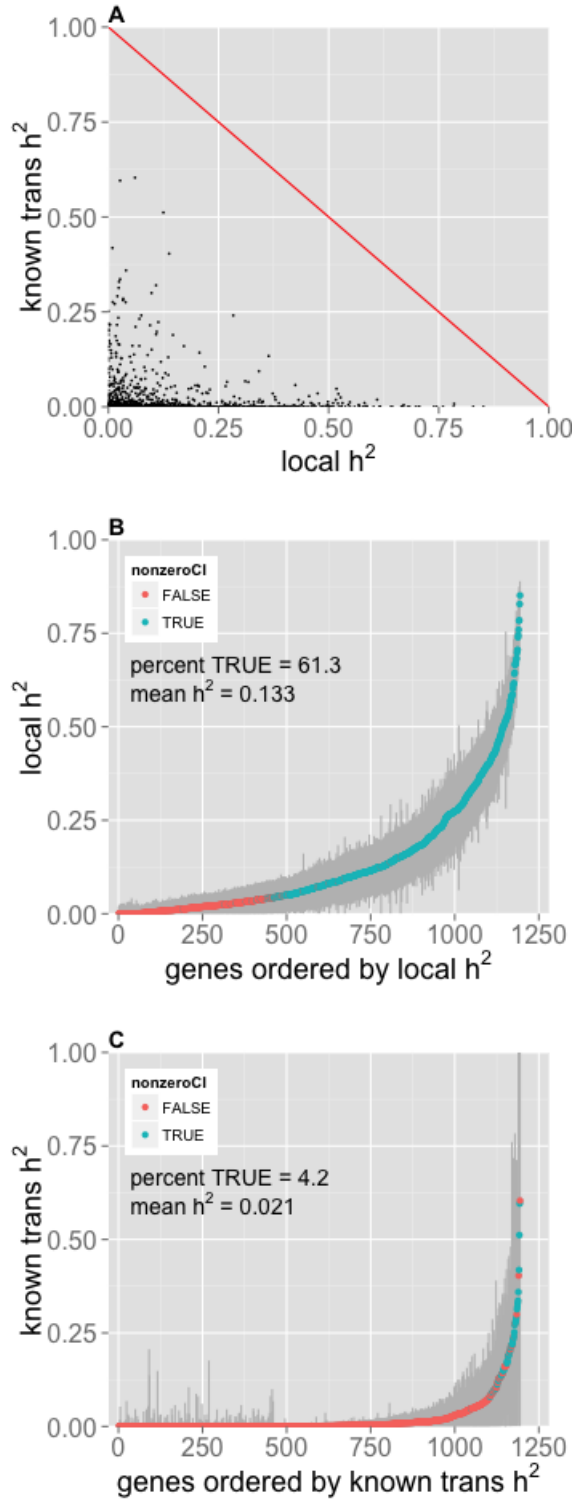


Figure 5: DGN whole blood expression joint heritability ( $h^2$ ) with known trans-eQTLs. Local (SNPs within 1 Mb of each gene) and known trans (SNPs that are trans-eQTLs in the Framingham Heart Study for each gene [FDR < 0.05])  $h^2$  for gene expression were jointly estimated. **(A)** Known trans  $h^2$  compared to local  $h^2$  per gene. **(B)** Local and **(C)** known trans gene expression  $h^2$  estimates ordered by increasing  $h^2$ . The 95% confidence interval (CI) of each  $h^2$  estimate is in gray and genes with a lower bound greater than zero are in blue.



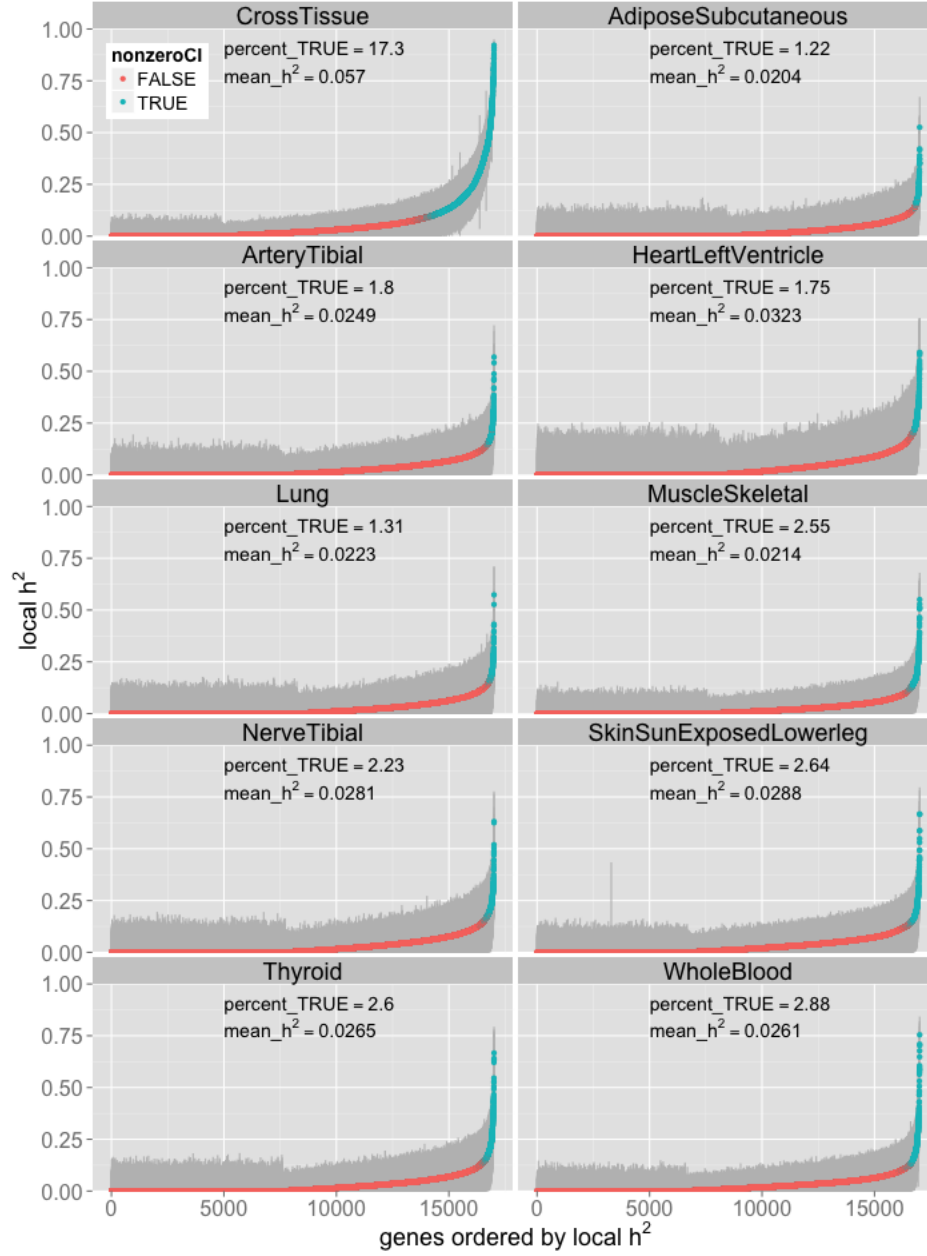


Figure 6: Cross-tissue heritability ( $h^2$ ) compared to tissue-specific  $h^2$ . Cross-tissue local  $h^2$  is estimated using the cross-tissue component (random effects) of the mixed effects model for gene expression and SNPs within 1 Mb of each gene. Tissue-specific local  $h^2$  is estimated using the tissue-specific component (residuals) of the mixed effects model for gene expression for each respective tissue and SNPs within 1 Mb of each gene.

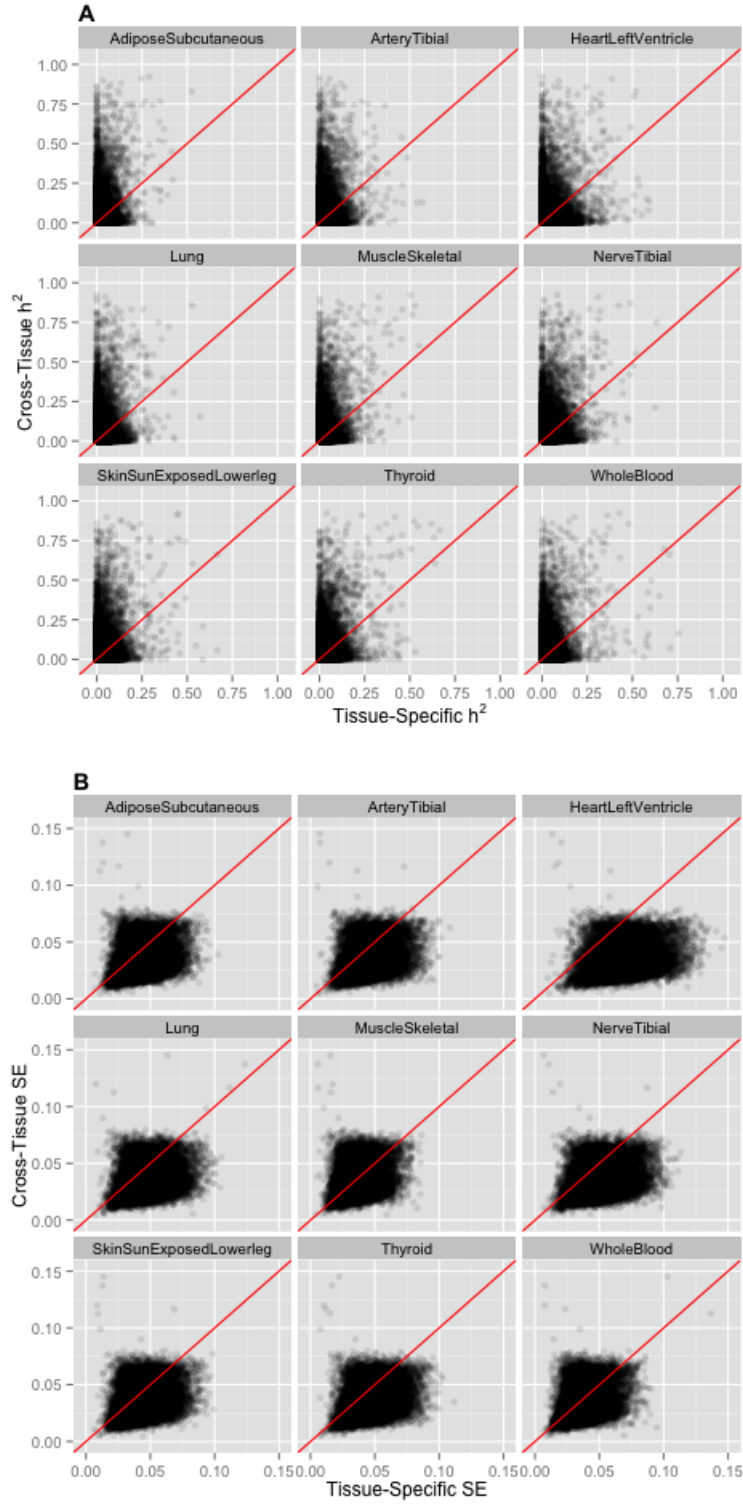


Figure 7: Cross-tissue and tissue-specific comparison of heritability ( $h^2$ , **A**) and standard error (SE, **B**) estimation. Cross-tissue local  $h^2$  is estimated using the cross-tissue component (random effects) of the mixed effects model for gene expression and SNPs within 1 Mb of each gene. Tissue-specific local  $h^2$  is estimated using the tissue-specific component (residuals) of the mixed effects model for gene expression for each respective tissue and SNPs within 1 Mb of each gene.