

## SNPsea: an algorithm to identify cell types affected by risk loci

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#### Abstract

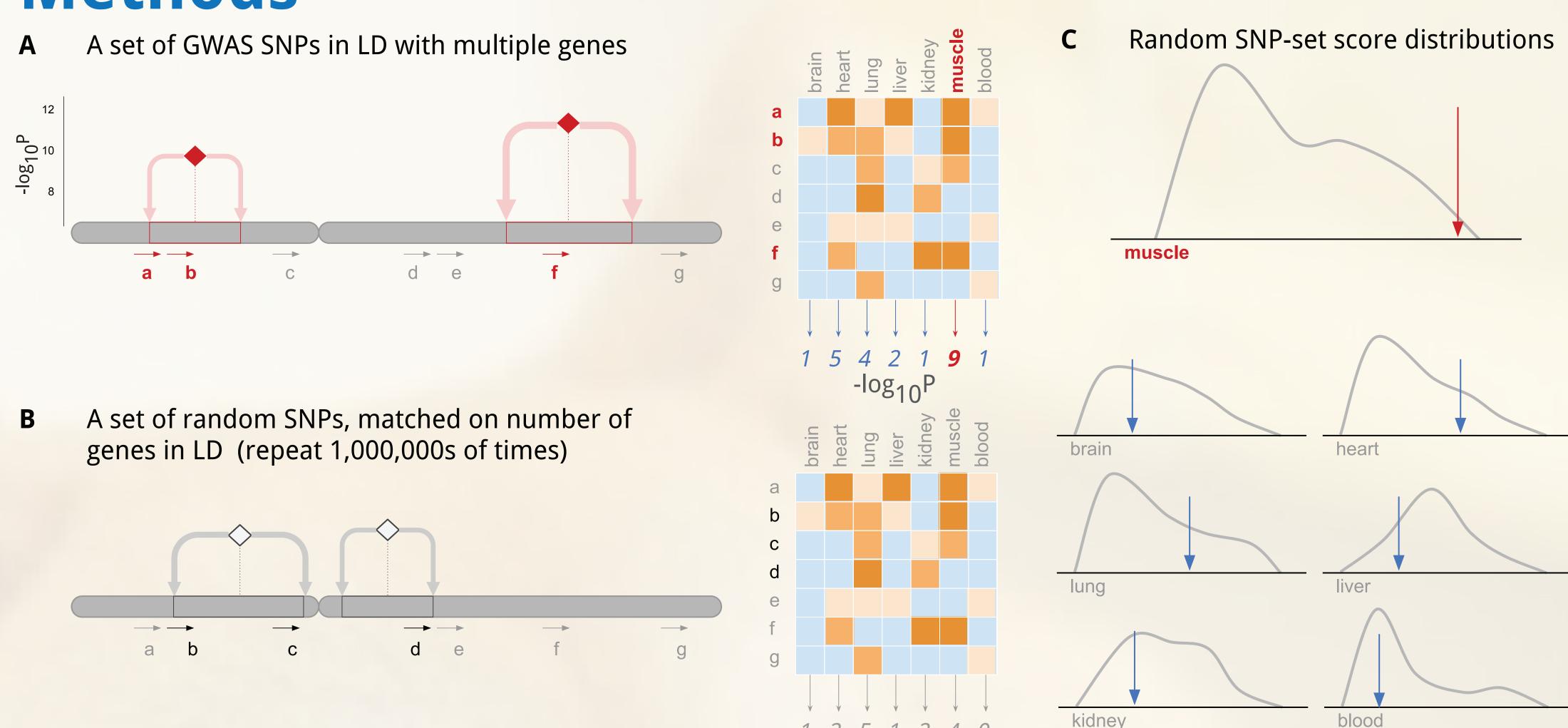
SNPsea is an algorithm to identify cell types and pathways likely to be affected by risk loci. It requires a list of SNP identifiers and a matrix of genes and conditions.

Genome-wide association studies (GWAS) have found many genomic loci associated with risk for different types of disease. SNPsea provides a simple way to determine cell types influenced by genes in these risk loci. Those cell types are likely to be relevant to the etiology of the given disease.

Suppose disease-associated alleles influence a small number of pathogenic cell types. We hypothesize that genes with critical functions in those cell types are likely to be within risk loci for that disease. We assume that a gene's specificity to a cell type is a reasonable indicator of importance to the unique function of that cell type.

htttp://broadinstitute.org/mpg/snpsea

## Methods



#### **Steps:**

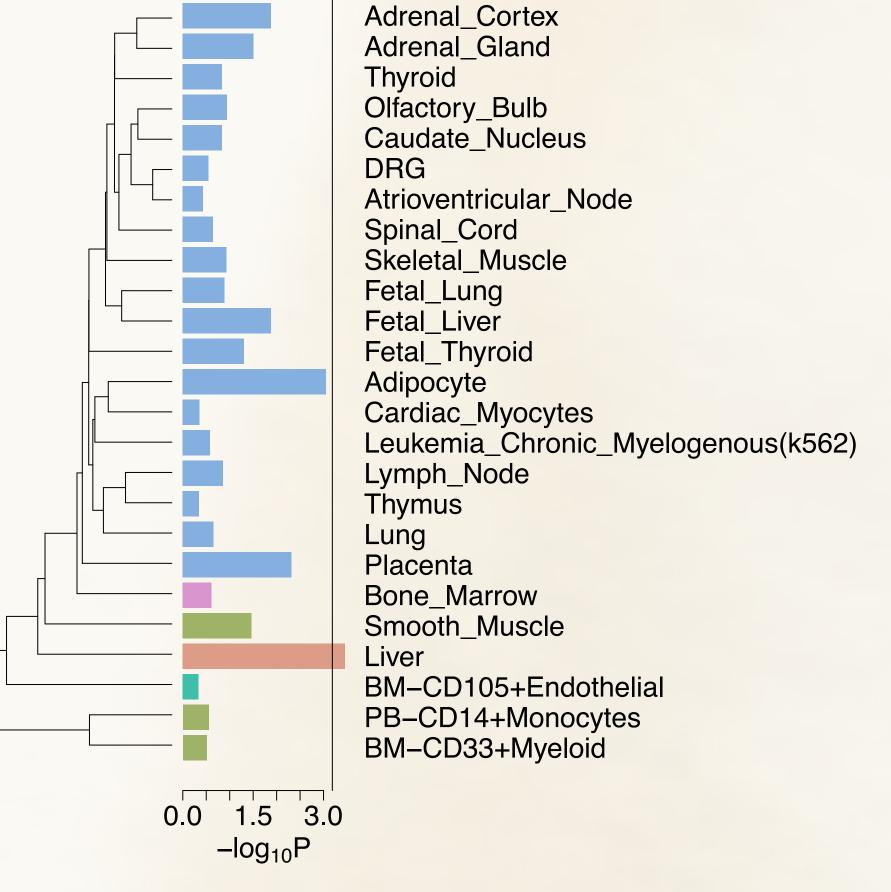
A | Identify genes in linkage disequilibrium (LD) with trait-associated SNPs. Score the gene set for specificity to each cell type.

**B** | Define a null distribution of scores for each cell type by sampling random SNP sets matched on the number of linked genes.

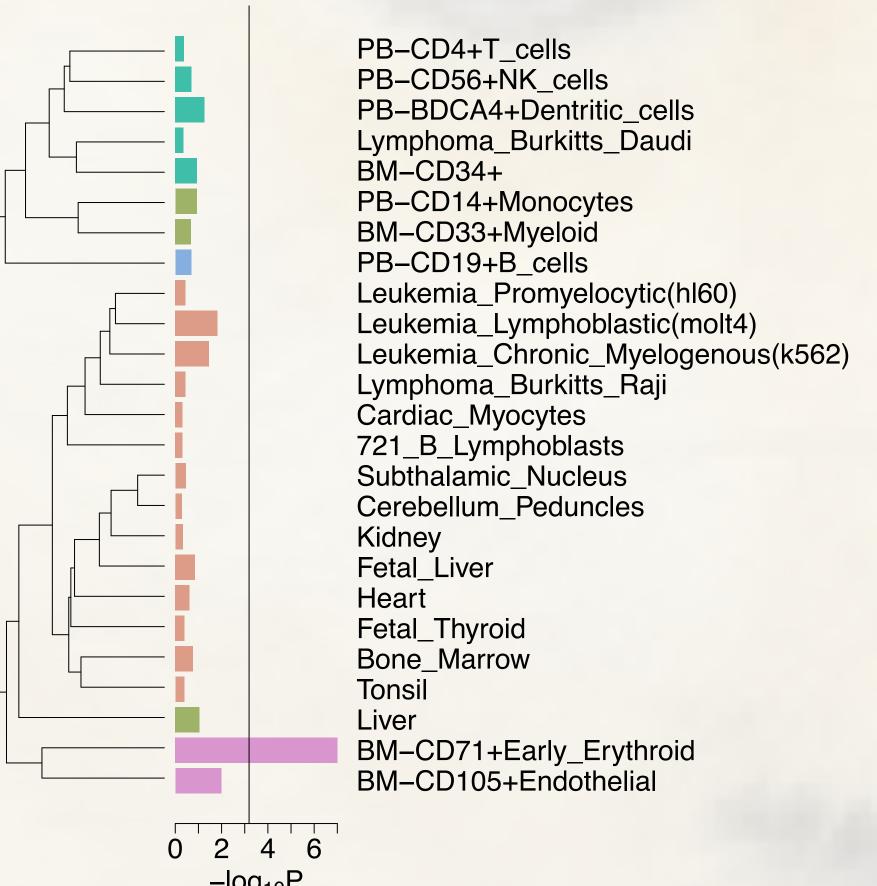
Evaluate significance of the original gene set's specificity by comparison to the null distributions.

#### Results

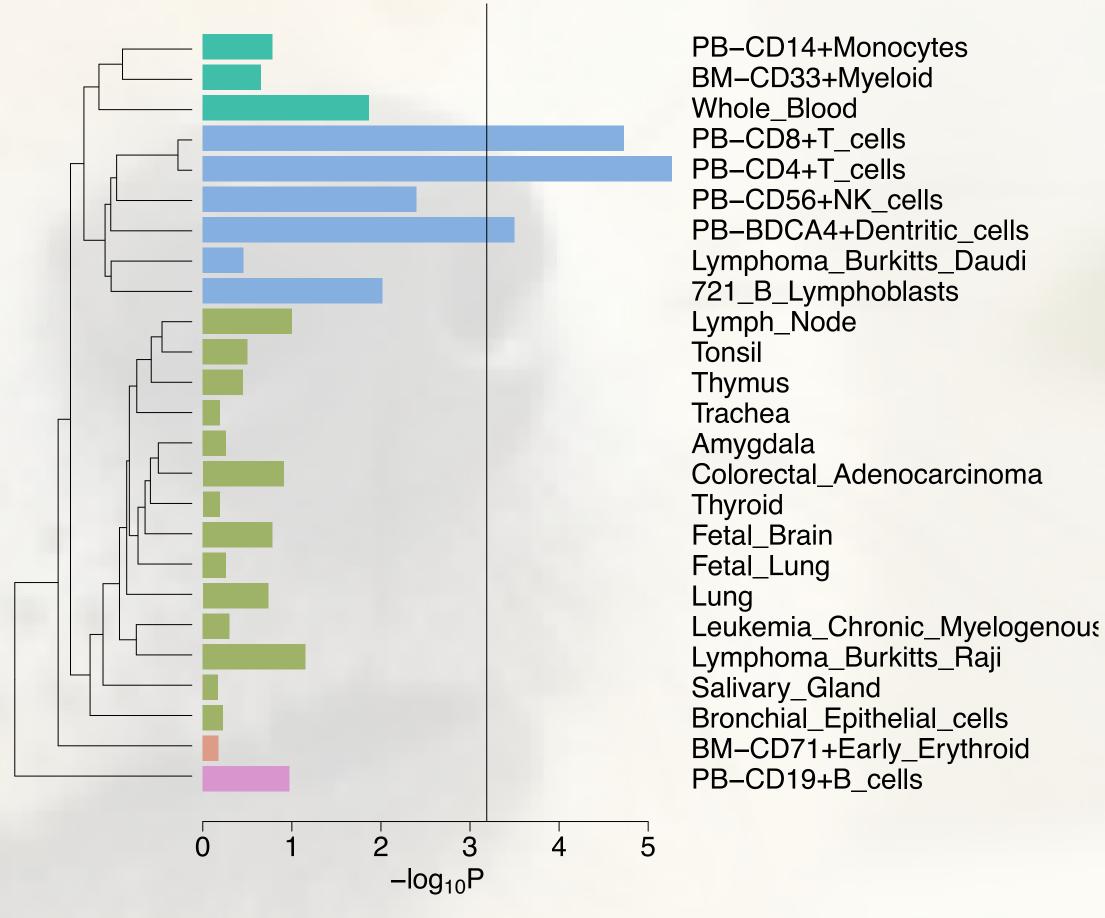
## HDL Cholesterol (46 SNPs)



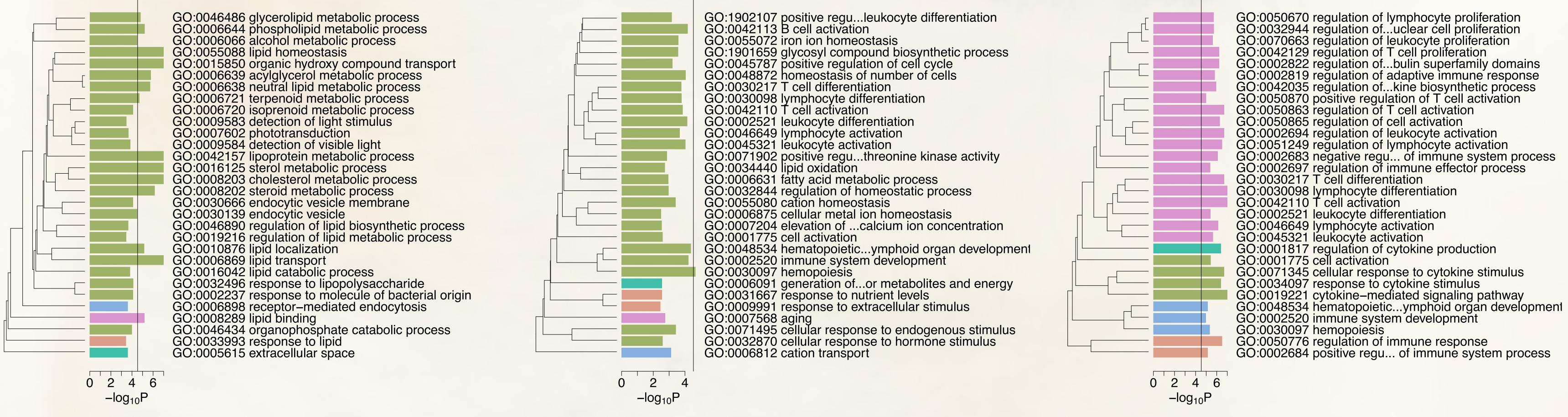
#### Red Blood Cell-Count (45 SNPs)



#### Celiac Disease (35 SNPs)



We tested each SNP set for cell type-specific gene expression relative to 78 other cell types or tissues in the **Gene Atlas** (Su et al. 2004). Top 25 are shown.



We also tested each SNP set for enrichment of gene annotations with 1,751 **Gene Ontology** (GO) terms (Botstein et al. 2000). Top 30 are shown.

## References

Elizabeth J. Rossin, et al. Proteins encoded in genomic regions associated with immune-mediated disease physically interact and suggest underlying biology. PLoS Genet, 7(1):e1001273, January 2011. The 1000 Genomes Project Consortium. An integrated map of genetic variation from 1,092 human genomes. Nature, 491(7422):56–65, November 2012. Simon Myers, et al. A fine-scale map of recombination rates and hotspots across the human genome. Science, 310(5746):321–324, October 2005. Xinli Hu, et al. Integrating autoimmune risk loci with gene-expression data identifies specific pathogenic immune cell subsets. The American Journal of Human Genetics, 89(4):496 506, 2011. Hana Lango Allen, et al. Hundreds of variants clustered in genomic loci and biological pathways affect human height. Nature, 467(7317):832–838, October 2010. Belinda Phipson and Gordon K. Smyth. Permutation p-values should never be zero: calculating exact p-values when permutations are randomly drawn. Statistical Applications in Genetics and Molecular Biology, 9(1), 2010. Pim van der Harst, et al. Seventy-five genetic loci influencing the human red blood cell. Nature, 492(7429):369–375, December 2012.

# Citation

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