GenomeRunner web server: epigenomic similarity and differences define the functional impact of SNP sets

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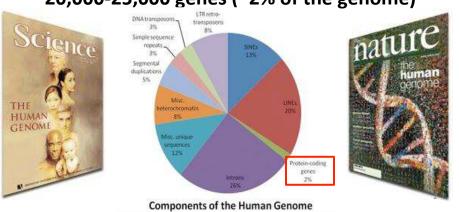
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03/05/2016

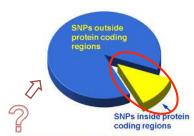
The human genome in a nutshell

- Human genome is big ~3.2 billion base pairs (~6ft long)
- 20,000-25,000 genes (~2% of the genome)



Genome research is gene-centric

- Microarrays
- Exome sequencing
- SNPs single nucleotide polymorphisms and other genomic variants (CNVs, InDels, SVs) are located everywhere



Only 12% of SNPs are located in, or occur in tight linkage disequilibrium with, protein-coding regions

Potential etiologic and functional implications of genome-wide association loci for human diseases and traits.

Hindorff LA, Sethupathy P, Junkins HA, Ramos EM, Mehta JP, Collins FS, Manolio TA.

Proc Natl Acad Sci U S A. 2009 Jun 9;106(23):9362-7. Epub 2009 May 27.

Genome gets annotated with regulatory information

ENCODE project – annotating regulatory elements

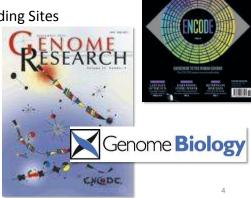
- DNasel hypersensitive sites

- Histone modification marks

- Transcription Factor Binding Sites

DNA methylation

 Epigenomic (regulatory) data genomic regions annotated as carrying functional and/or regulatory potential



nature

Even more epigenomic data

 Roadmap Epigenomics — uniformly processed annotations of regulatory elements across normal tissues.











Gap between data generation & data understanding



The Road to the \$1,000 Genome

"There is a growing gap between the generation of massively parallel sequencing output and the ability to process and analyze the resulting data," says Canadian cancer research John

being solved. Thus, the idea of a '\$1000 genome, \$1000000 genome interpretation' was expressed by the president of the American College of Medical Genetics, Bruce Korf. In this review, we

GenomeRunner – a global positioning system within the genome

 Finds significantly enriched co-localizations between 'omics' data and genome-wide epigenomic data



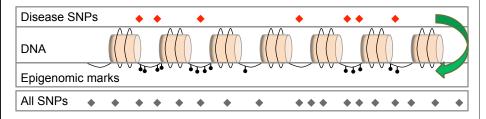
Epigenomic enrichment analysis

Gene enrichment and epigenome enrichment

- Gene set enrichment analysis
 - Genes
 - enriched in
 - gene ontology (functional) annotations
- Epigenome enrichment analysis
 - SNPs
 - enriched in
 - regulatory (epigenomic) annotations

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Statistics of epigenomic enrichments



- 6 out of 7 disease-associated SNPs overlap with epigenomic marks
- How likely this to be observed by chance?
 (Chi-square test/Binomial test/Permutation test)

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EpiEuplorer	http://epiesplaner.mpi-inf.mag-		7	197	2	Through registration		1	100		Visiother tools	18	450	- 19	10
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Genomic Association Tester (GAT)	https://github.com/Andreashing				- 12		721		. 67	3.	1	120	7/7	72	
PodBet	http://www.podbat.org		100	+.	- 81			Limited				380	\$10	100	+
EpiRegNet GREAT	http://wangiab.bku.hk/Ep/RegN http://bejerano.stanford.edu/gr			- 3				Limited		100		- 31	7.0		- 3
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GenomeRunner	http://www.integrativegenomic		. +:								4	4		100	4.

Functional impact of SNPs

- Hypothesis: SNPs in epigenomic regions may disrupt regulation
- Epigenomic regions enriched in SNPs (SNP burden) are disrupted more
- Epigenomic signature = types of epigenomic marks most enriched in SNPs



Functional impact of SNPs HDL cholesterol Enriched regulatory associations 3.0e-17 2.0e-15 1.4e-13 9.6e-12 26.6e-10 24.5e-08 43.1e-06 2.1e-04 1.5e-02 1.0e+00

Epigenomic enrichments are sorted from most to least significant

E118-H3K4me1_bPk-processed = <cell/tissue ID>-<factor ID>-<source info>

Functional impact of SNPs HDL cholesterol

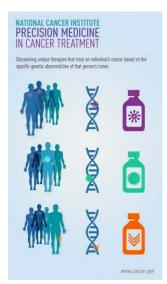


HDL cholesterol-associated SNPs are enriched in liver-specific activating H3K4me1 and H3K27ac histone modification marks

E118-H3K4me1_bPk-processed = <cell/tissue ID>-<factor ID>-<source info>

Epigenomic similarity analysis

Epigenomic similarity among SNP sets



- Similar SNP sets may have similar functional impact = be enriched in similar types of epigenomic regions
- Comparing functional impact of SNP sets may help understanding epigenomic mechanisms of complex diseases

Epigenomic similarity among diseaseassociated SNP sets

Name
Sone mineral density
Creative protein
All other
All other
Creative protein
All other
All other
All other
All further all other
All further all other
Real blood cell traits
Alse other
Real sell sell sell other
Real sell sell sell other
Alse other

Disease SNPs₂

Disease SNPs_N

1.17E-08

6.45E-04

- We analyzed SNP sets associated with 39 complex diseases and traits
 - 21 immunologic diseases
 - 4 neurologic diseases/traits
 - 7 metabolic diseases/traits
 - 7 other traits

Genetic and epigenetic fine mapping of causal autoimmune disease variants

Kyle Kai-How Fark¹⁻², Alexander Marson¹⁺, Jiang Zhu^{1-A-5}, Markis Kielnewierfeld¹⁺, William J. Hossley², Smantha Bekl², Osum Shoresh², Holly William², Rassell J. H. Kyan²⁻, Alexander A. Shishikin⁴⁻, Metal Haran², Maries J. Carrasco-Alfonno².

.22E-01

4.07E-05

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Defining epigenomic enrichment profiles of disease-associated SNP sets Types of epigenomic features Enriched in DNAse hypersensitive sites Disease SNPs, 3.00E-07 5.21E-02 7.37E-05 2.01E-04

• **Epigenomic enrichment profile** – SNP set-specific vector of epigenomic enrichment p-values

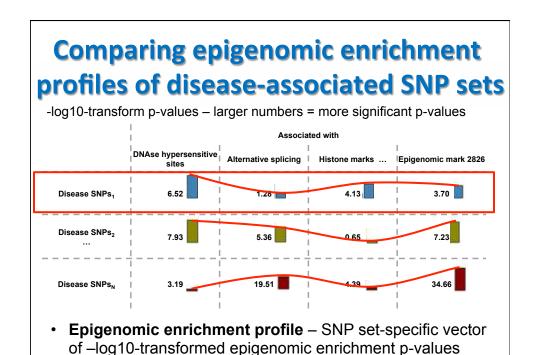
-log₁₀(p-value)

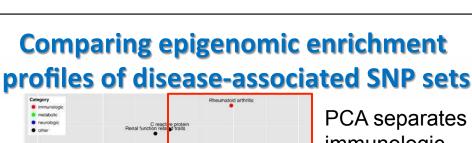
3.07E-20

18

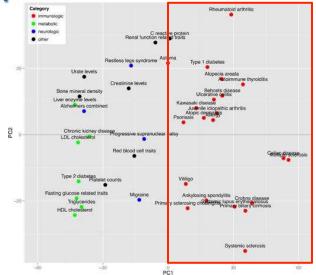
5.93E-08

2.19E-35

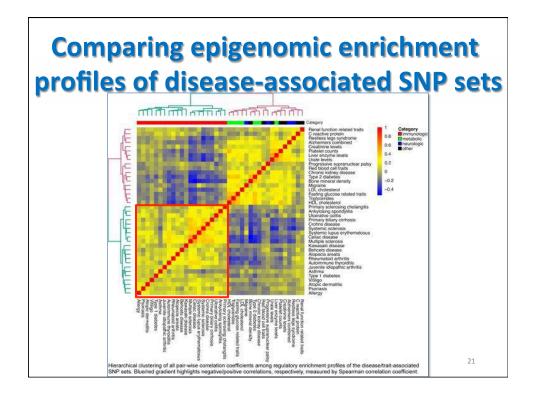


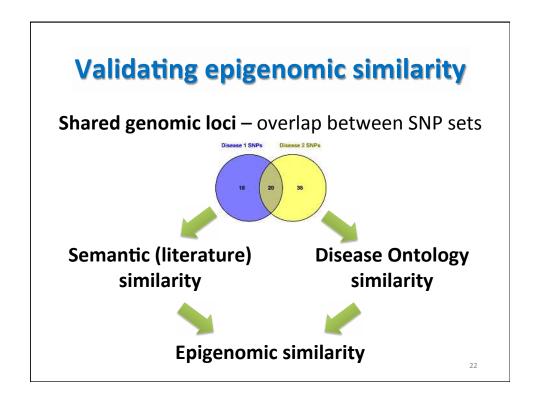


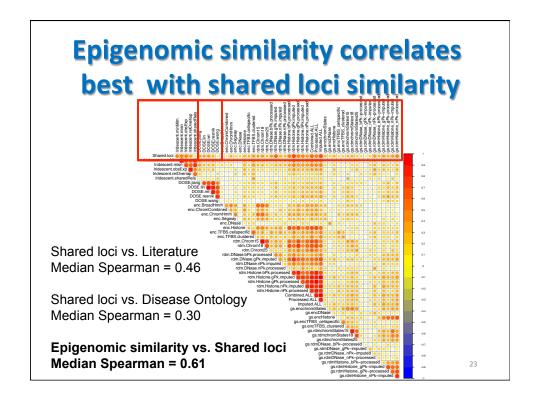
· Compare them using PCA, Spearman correlation



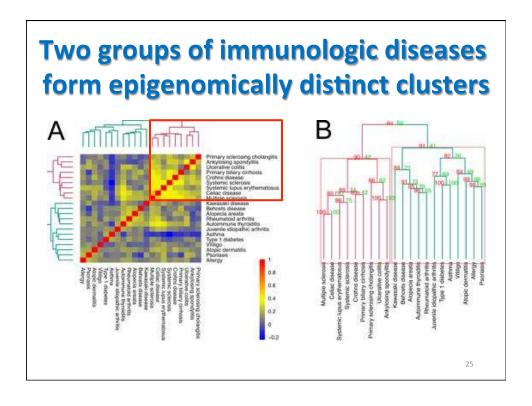
PCA separates immunologic disease-associated SNP sets as the most epigenomically distinct from others

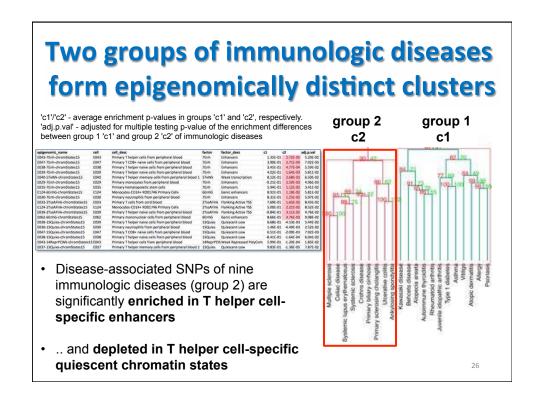






Differential epigenomic analysis

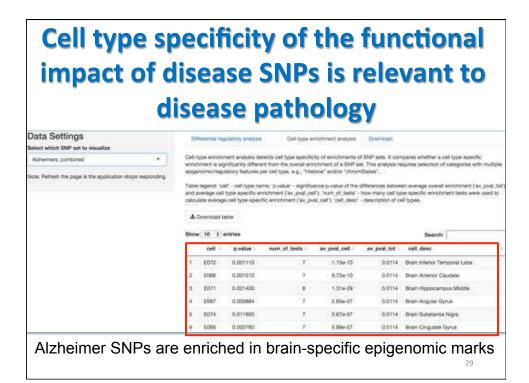




Cell type-specific epigenomic enrichment analysis

Cell type-specific epigenomic enrichment analysis defines the origin of the functional impact

- Epigenomic elements are cell- and tissue type specific
- Cell type-specific epigenomic enrichment analysis identifies cell types where epigenomic enrichments are the most significant



Conclusions

- GenomeRunner defines potential functional impact of SNP sets via **epigenomic enrichment analysis**
- Epigenomic similarity analysis identifies regulatory similarity and differences among SNP sets
- Cell type-specific enrichment analysis prioritizes cell/tissue type specificity of the epigenomic enrichments
- Epigenomic enrichment analyses can be applied to any genomic signature, from disease-associated SNP sets to patient-specific genotypes

Thank you

Genome analysis

GenomeRunner web server: Regulatory similarity and differences define the functional impact of SNP sets

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https://github.com/mdozmorov/presentations