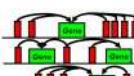


Genome = Genes + Gene Regulation



CIS REGULATION

Type	# in genome	% of genome
genes	20,000	2%
ncRNA	20,000	1%
cis elements	1,000,000	>10%



- Encode causality
- Disease susceptibility
- Driver sequences
- Alter cell state
- Key for evolution

Atomic event – transcription factor binding

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Disease Associated tag SNPs

- Over 15,000 distinct tag SNPs in the GWAS Catalog
- 80-90% far away from (linkage with) gene exons
- Are most gene *cis* regulatory?
- Are they near genes with common functionality?

GWAS Catalog Growth

Category	2008	2011	2016
Disease	~100	~1000	~15,000
Phenotype	~100	~1000	~15,000
Trait	~100	~1000	~15,000

Cis-reg enrichments: GREAT.stanford.edu

½ million job submissions, 700+ references, established defaults

Gene transcription start site
Gene regulatory domain
Function ('abnormal cardiac output')
Cis-reg rich region set

π

$p_{\pi} = 0.33$ of genome annotated with π
 $n = 6$ genomic regions
 $k_{\pi} = 5$ genomic regions hit annotation

$P = \Pr_{\text{binom}}(k_{\pi} \geq 5 | n=6, p_{\pi}=0.33)$

GREAT = Genomic Regions Enrichment of Annotations Tool

[McLean et al, Nature Biotech, 2010]

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Cis-reg enrichment: GREAT.stanford.edu

½ million job submissions, 700+ references, established defaults

Gene transcription start site

Gene regulatory domain

Function ('abnormal cardiac output')

Cis-reg rich region set

Advantages of GREAT:

1. Accounts for both proximal and distal binding sites
2. Variable length gene regulatory domains
3. Multiple hits next to same gene add significance
4. Extensive body of knowledge (16,000 functions)

$p_g = 0.33$ of genome annotated with

$n = 6$ genomic regions

$k_g = 5$ genomic regions hit annotation

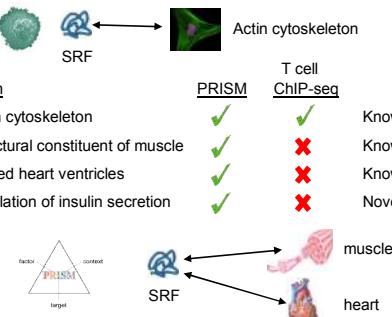
GREAT = Genomic Regions
Enrichment of Annotations Tool

$P = \Pr_{\text{binom}}(k_g \geq 5 | n=6, p_g=0.33)$

[McLean et al, Nature Biotech, 2010]

Unlinked GWAS SNPs → GREAT

PRISM vs. ChIP-seq → GREAT

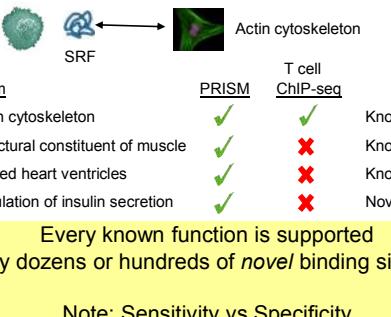


Term	PRISM	T cell ChIP-seq	
actin cytoskeleton	✓	✓	Known
structural constituent of muscle	✓	✗	Known
dilated heart ventricles	✓	✗	Known
regulation of insulin secretion	✓	✗	Novel



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PRISM vs. ChIP-seq → GREAT



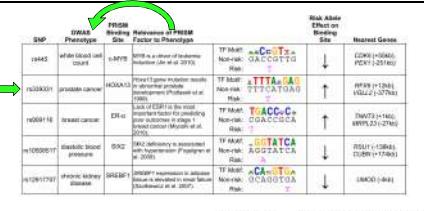
Term	PRISM	T cell ChIP-seq	
actin cytoskeleton	✓	✓	Known
structural constituent of muscle	✓	✗	Known
dilated heart ventricles	✓	✗	Known
regulation of insulin secretion	✓	✗	Novel

Every known function is supported by dozens or hundreds of *novel* binding sites.

Note: Sensitivity vs Specificity

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GWAS SNPs: Predict upstream regulator

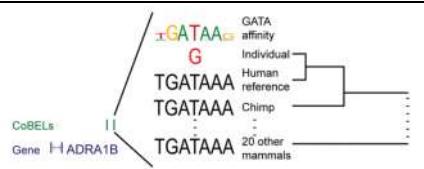


SNP	Phenotype	PRISM Predicted Reference Site	PRISM Factor to Phenotype	Risk Allele Effect on Binding Site	Nearest Gene
rs6452	white blood cell count	Non-risk			
rs1234567	prostate cancer	HOXA13	HOXA13 is a driver of prostate cancer (Jia et al. 2013).	↓	CDXB (+1046); PDXK (+2516)
rs1234567	prostate cancer	HOXA13	HOXA13 is a driver of prostate cancer (Jia et al. 2013).	↑	AVAN (+1248); VGLL2 (+3748)
rs1234567	diabetic ketoacidosis	EF1α	Loss of EF1α is the most common genetic variation in people with type 1 diabetes (Hawley et al. 2012).	↑	DART (+146); WIF1 (+2126)
rs1234567	diabetic ketoacidosis	EF1α	Loss of EF1α is the most common genetic variation in people with type 1 diabetes (Hawley et al. 2012).	↑	ESR1 (+1364); EGR1 (+1746)
rs1234567	epicardial kidney disease	SPNS2	SPNS2 expression is decreased in diabetic kidney disease in mouse models (Makrilia et al. 2007).	↓	LMO4 (+84)
rs1234567	epicardial kidney disease	SPNS2	SPNS2 expression is decreased in diabetic kidney disease in mouse models (Makrilia et al. 2007).	↓	LMO4 (+84)

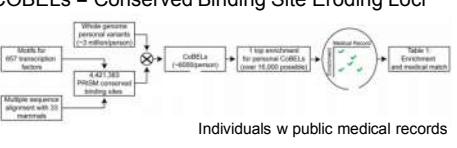
Huang et al. A prostate cancer susceptibility allele at 6q22 increases RFX6 expression by modulating HOXB13 chromatin binding. Nat. Genet. 2014 Feb;46(2):126–135.

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Personal Deleterious Binding Sites

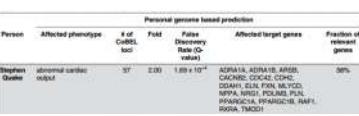


COBELs = Conserved Binding Site Eroding Loci



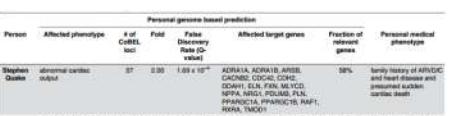
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COBELs → GREAT



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COBELs → GREAT



Autosomal right ventricular dysplasia cardiomyopathy is a rare disease whose penetrance is estimated to affect approximately 1 in 5,000 individuals. [1] This disease is frequently familial and typically involves autosomal dominant transmission with low penetrance and variable expressivity. [2]

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COBELs → GREAT

Person	Personal genome-based prediction					Personal medical phenotype
	Affected phenotype	# of C4EFL	Fold	P-value (Discovery or meta-analysis p-value)	Affected target genes	
Stephen	abdominal cardiac celiac	57	2.85	1.09 x 10 ⁻¹⁰	ADRA1A, ADRA1B, ARRB1, CACNB1, CCDC106, CACNA1D, SMC3, PIPKIN, PLXNC1, PRKDC, PRKDC, PRKDC, PRKDC, RAP1GAP	58%
George	“Arrhythmogenic right ventricular dysplasia/cardiomyopathy is an inherited cardiomyopathy estimated to affect approximately 1 in 5,000 individuals worldwide. The disease is usually fatal and typically presents before age 30 years.” propranolol: prazosin: verapamil: nervous system: TPRFSA	23	2.39	1.18 x 10 ⁻¹⁰	EDH2, HES1, HES3, HOXA1, HOXA10, HOXA11, PLXNC1, TPRFSA	85% hypertension
Charmain	“a non-syndromic involvement of the autonomic nervous system is markedly increased” TPRFSA	23	2.39	1.18 x 10 ⁻¹⁰	TPRFSA	85% hypertension

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COBELs → GREAT

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COBELs → GREAT

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COBELs → GREAT

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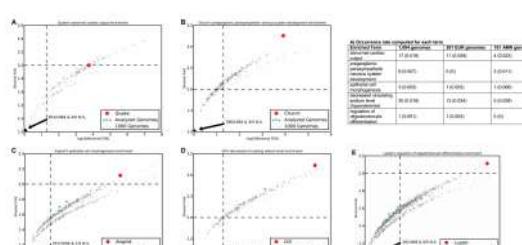
Randomize COBELs

- Replace every CoBEL with a random binding site prediction for the same transcription factor of same affinity and similar cross-species conservation.
 - Using 10,000 random control sets, the likelihood of obtaining the functions reported in Table 1 as top prediction due to bias in the distribution of binding sites in the genome is low (Quake P = 3×10^{-4} , Church P = 5.7×10^{-3} , Angrist P = 4.8×10^{-3} , Gill P = 1×10^{-4} , Lupski P = 1.9×10^{-3} , and combined P = 1.6×10^{-15}).
 - Significance remains high when we relax the requirement to recover each exact same term with matching any one of a broader group of 12–60 related functions as a top prediction (Quake P = 1.1×10^{-3} , Church P = 1.3×10^{-2} , Angrist P = 7.7×10^{-3} , Gill P = 7.4×10^{-3} , Lupski P = 6.5×10^{-3} , and combined P = 5.2×10^{-12}).

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KGP as Controls



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Randomize Medical Histories

- Define an **association matrix linking enrichment and medical history**, with the phenotypes observed in the five individuals as rows, and top enriched terms in all as columns. A cell in the matrix would be marked "true" only where the enriched term (of any individual) is thought to be related to the etiology of the phenotype (of any individual).
 - One instance of this matrix was filled by a **medical doctor** based on their medical knowledge and training and another instance was independently filled using a **literature survey**. The objective was to compute the chance of associating a set of five individuals with random medical histories with the observed enrichments using one of the two association matrices as the "gold" association.
 - We generated 1,000 sets of five individuals with random medical histories composed of similar disease profiles and assessed the likelihood of being able to associate them with enrichments. Successfully linking five random individuals with enrichments was highly significant using the association matrix generated by the **medical doctor** ($P = 3.0 \times 10^{-3}$) and by the matrix generated by **literature survey** ($P = 3.0 \times 10^{-2}$) suggesting our links between enrichment and medical histories are not just a function of the listed histories.

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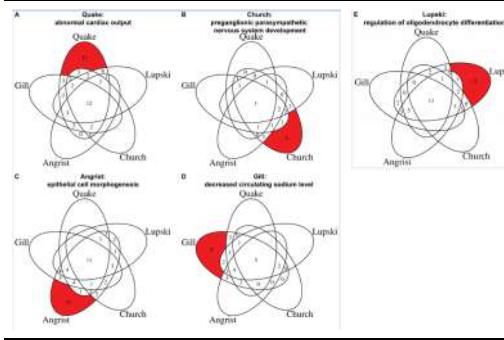
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COBELs ≠ GWAS SNPs or HGMD

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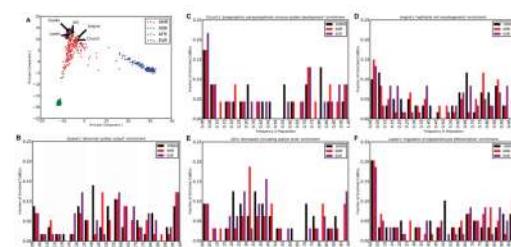
Most Predictive COBELs are Private



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Contributions from Common & Rare



Subset to 1% freq in KGP → lose all enrichments

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Summary

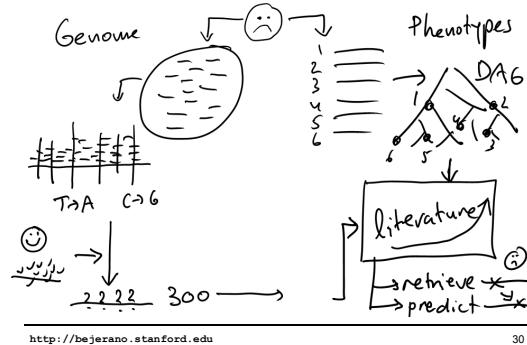
- We define likely deleterious events as personal variants that erode the affinity of human conserved binding sites.
 - When the set of all such events is probed for lying next to gene sets of particular function or phenotype, we repeatedly get a solid match between top genomic prediction and self reported medical summary.
 - Top genomic predictions are eroded at both gene and gene set level.
 - The variants we highlight appear to be part of the mutational load pre-disposing individual lineages to different diseases.



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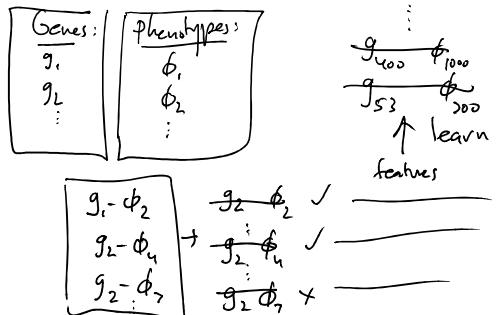
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Other Lab Interests: 1) Solve Patient



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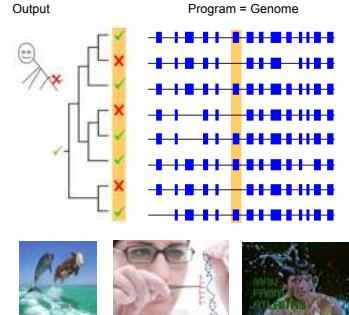
2) Automate Patient Solving



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3) Discover Mammalian Adaptations



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Kudos

COBELs: (PLoS Comp Bio, 2016)

Harendra Guturu, Sandeep Chinchali, Shoa Clarke

PRISM: (Genome Research, 2013)

Aaron Wenger, Shoa Clarke, Harendra Guturu, Jenny Chen, Bruce Schaar, Cory McLean

GREAT: (Nature Biotechnology, 2010)

Cory McLean, Dave Bristor, Michael Hiller, Shoa Clarke, Bruce Schaar, Craig Lowe, Aaron Wenger

Bejerano Lab past & present

The Organizers

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