Identify Disease-Causal Genes from GWAS Loci by 3D Genome Structure, Regulatory Landscapes & Deep Learning

Yi-Hsiang Hsu, MD, ScD

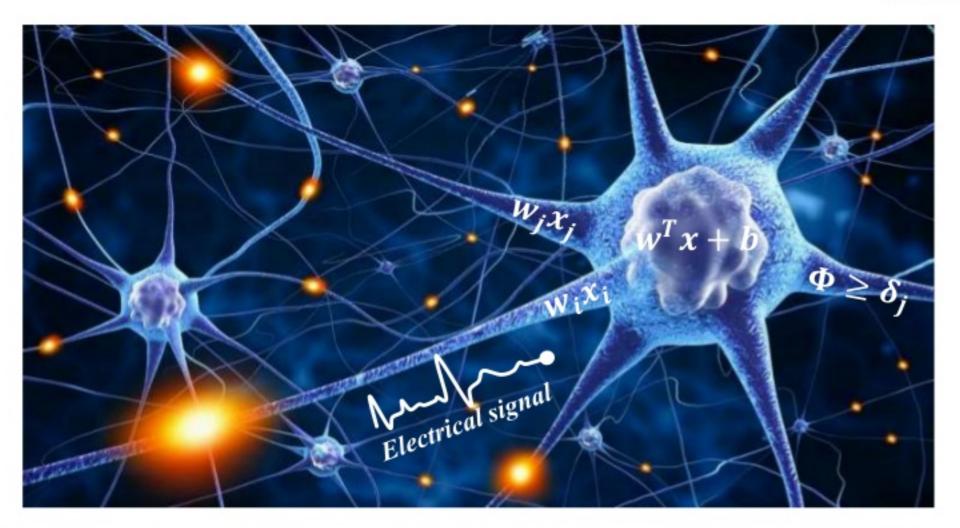


YongSheng Huang, Ph.D



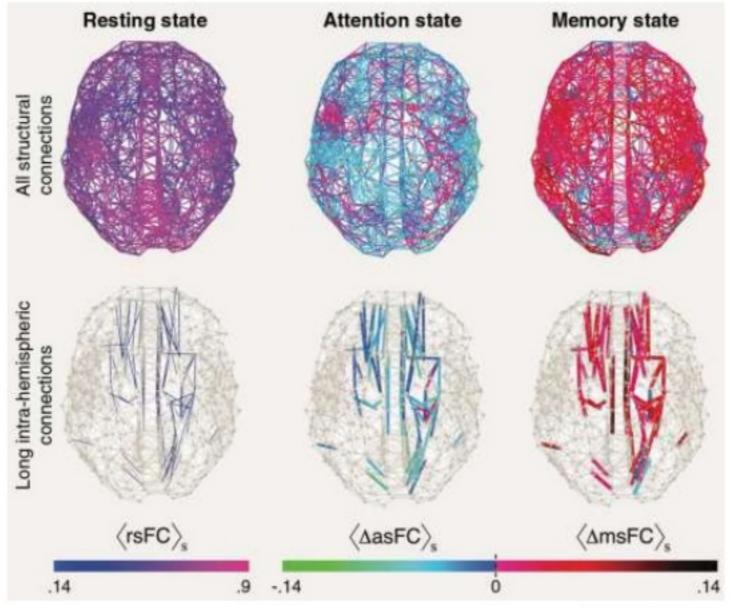
Deep Learning: The Inspiration

"the deepest concepts in mathematics are those which link one world of ideas with another"
---- Freeman Dyson





Deep Learning: The Natural Form





Deep-Learning: The Renaissance

The New York Times

Scientists See Promise in Deep-Learning Programs

by JOHN MARKOFF Nov. 23, 2012

In the 1960s, believed that a workable artificial intelligence system was just 10 years away. In the 1980s, a wave of commercial start-ups collapsed, leading to what some people called the "A.I. winter."

But recent achievements have impressed In October, for example, a team of graduate students studying with the University of Toronto computer scientist Geoffrey E. Hinton won the top prize in a contest sponsored by Merck to design software to help find molecules that might lead to new drugs.



Deep Learning: Impact on Medicine

On par performance as 21 board-certified pathologists

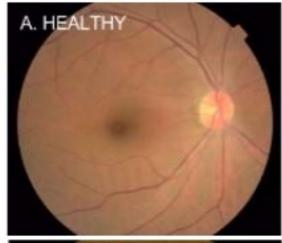
Epidermal lesions

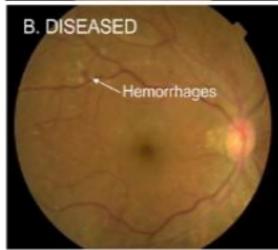




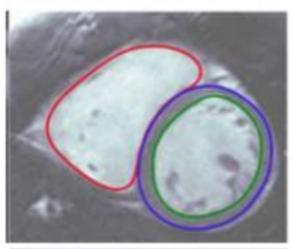
Nature 2017 Feb

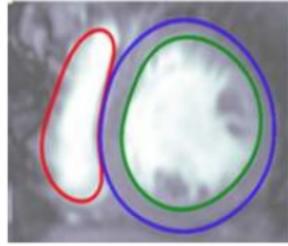
>90% specificity and sensitivity as board-certified ophthalmologists





Artery's Cardio DL wins FDA approval for clinical diagnosis (10-sec vs. 1hr)







Deep Learning: The New Disruption

Can we leverage DL to identify genetic variants that are disease causal, so that we can treat diseases at its root level per individual patient?



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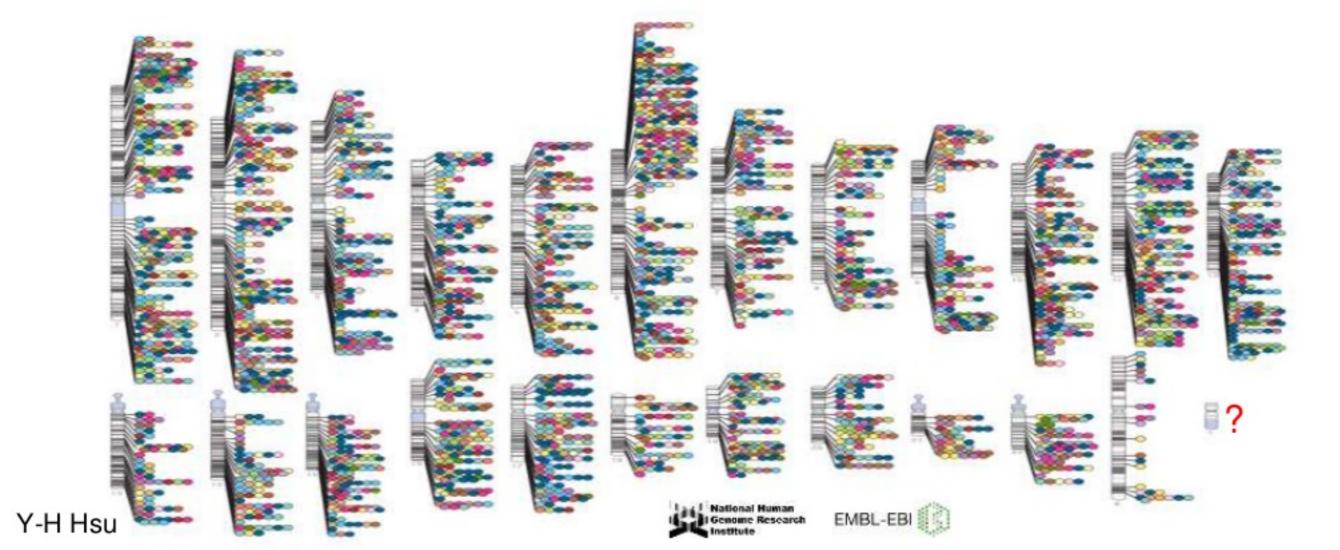




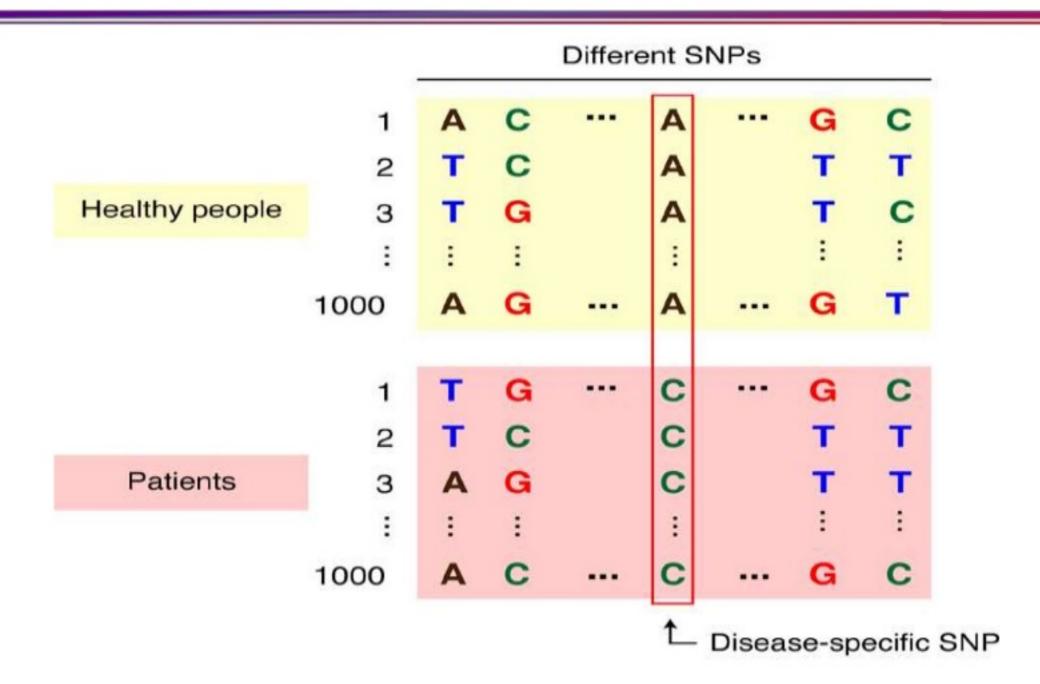


Genome-Wide Association Studies (GWAS) Catalog

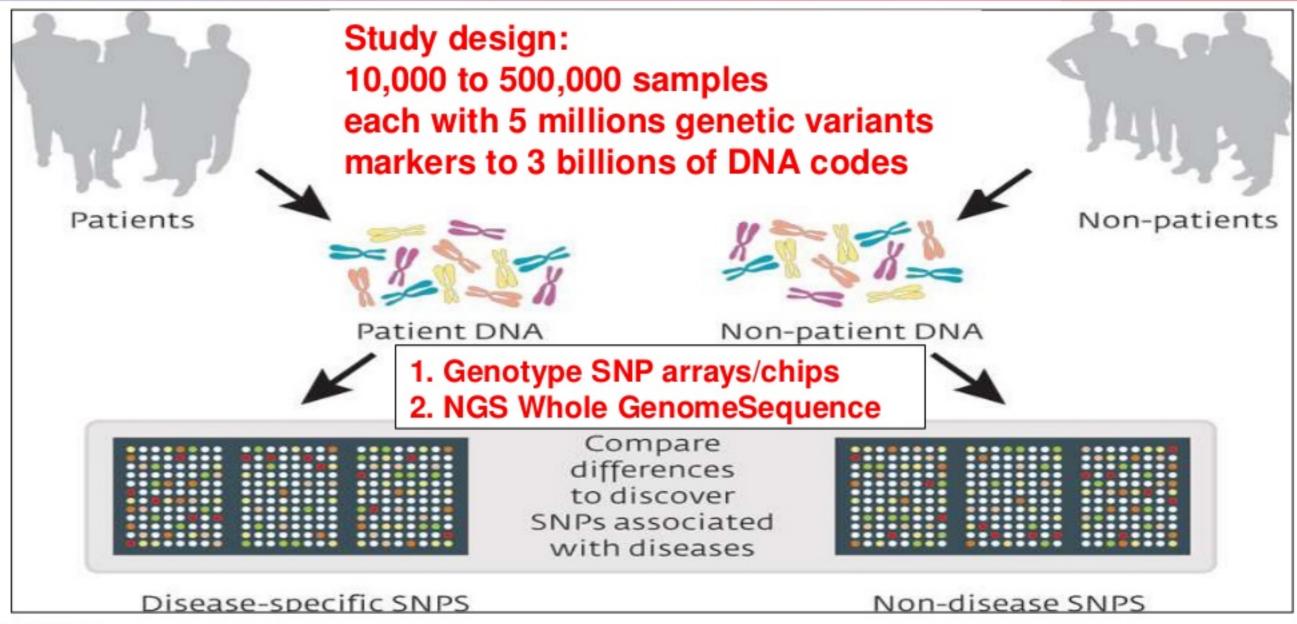
Identified ~13,000 genetic variants (single nucleotide mutations/ polymorphisms) to be associated with ~2,000 diseases/phenotypes



Genome-Wide Association Scans

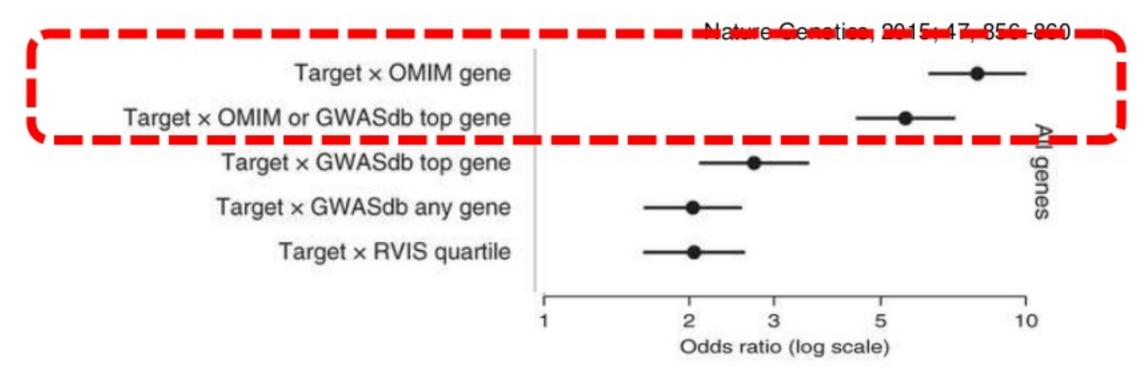


GWAS (Whole Genome Association) Scans



% Successfully Approved Drugs & Human Genetics

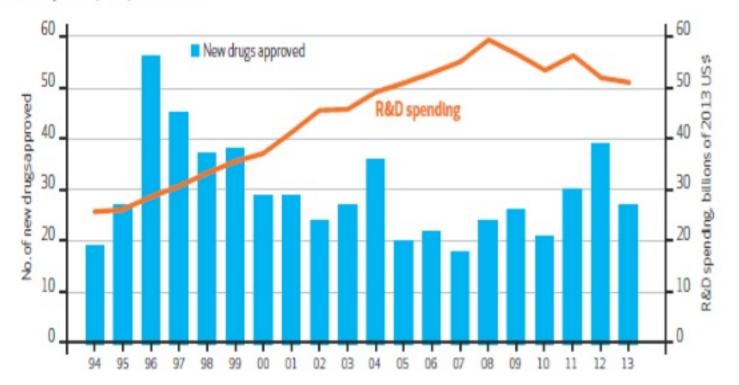
The impact on medical care from GWAS could potentially be substantial



- FDA approved drugs with human genetic information are 5~10X more likely to be successful
- Failure targets at each drug development stage (pre-clinical, phase I, II, III) are more likely to be those targets without genetic validation

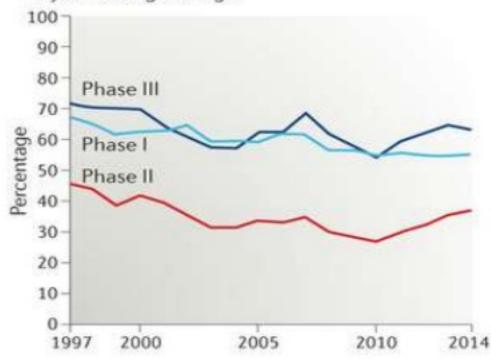
R&D Spending on New Drugs # Drug Approvals

Annual New Drug Approvals By The Food And Drug Administration (FDA) And Industry Spending On Research And Development (R&D), 1994–2013



a Success rates by phase

Percentage likelihood of moving to next phase, 3-year rolling average*



- New a better drug development pipeline
- Utilizing human genetic information/validation is the key

Genome-Wide Association Studies (GWAS) Catalog

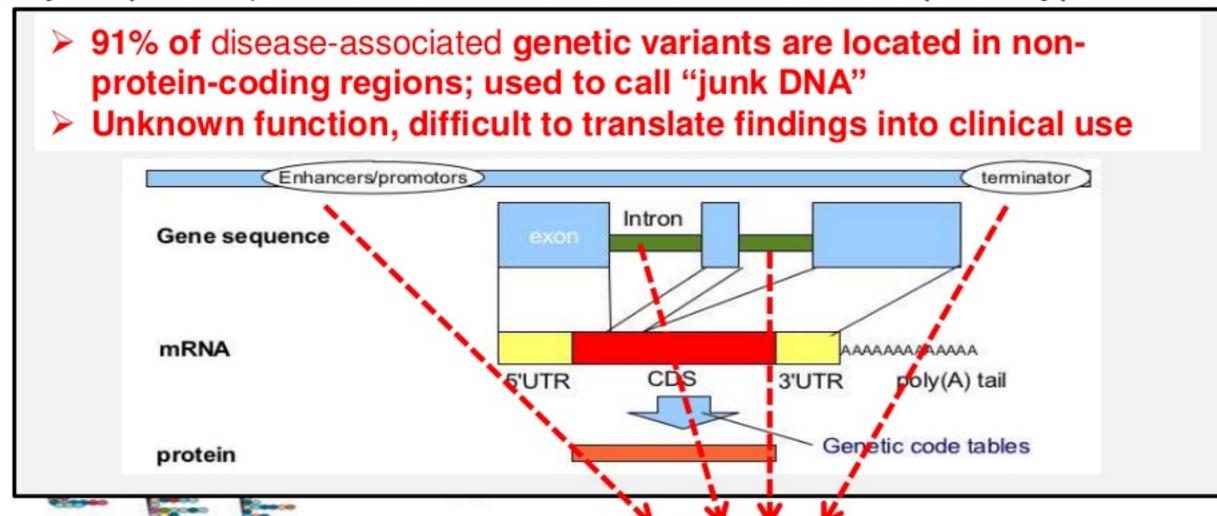
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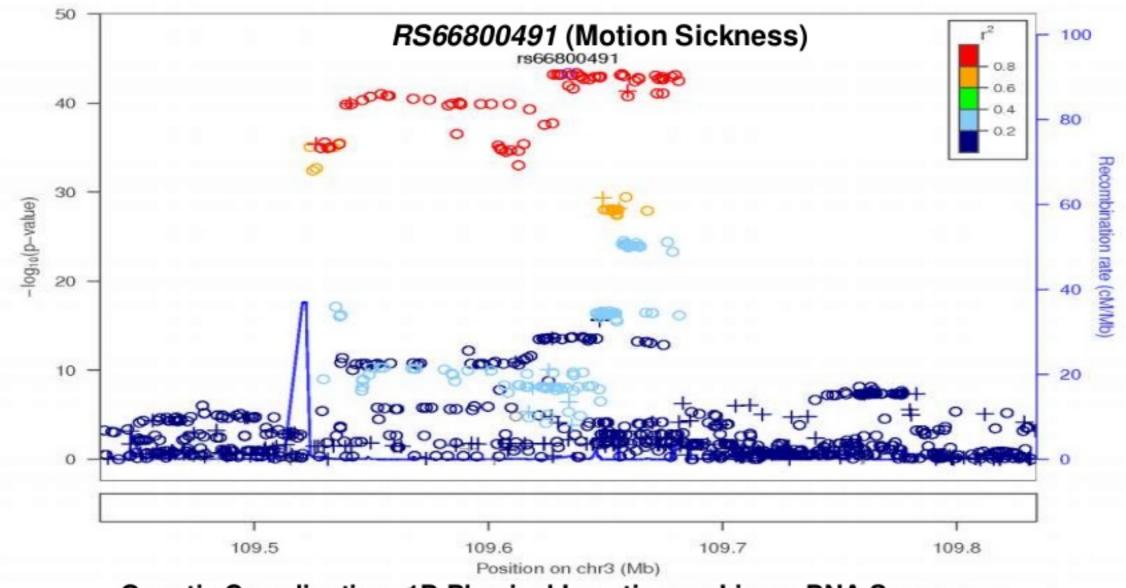
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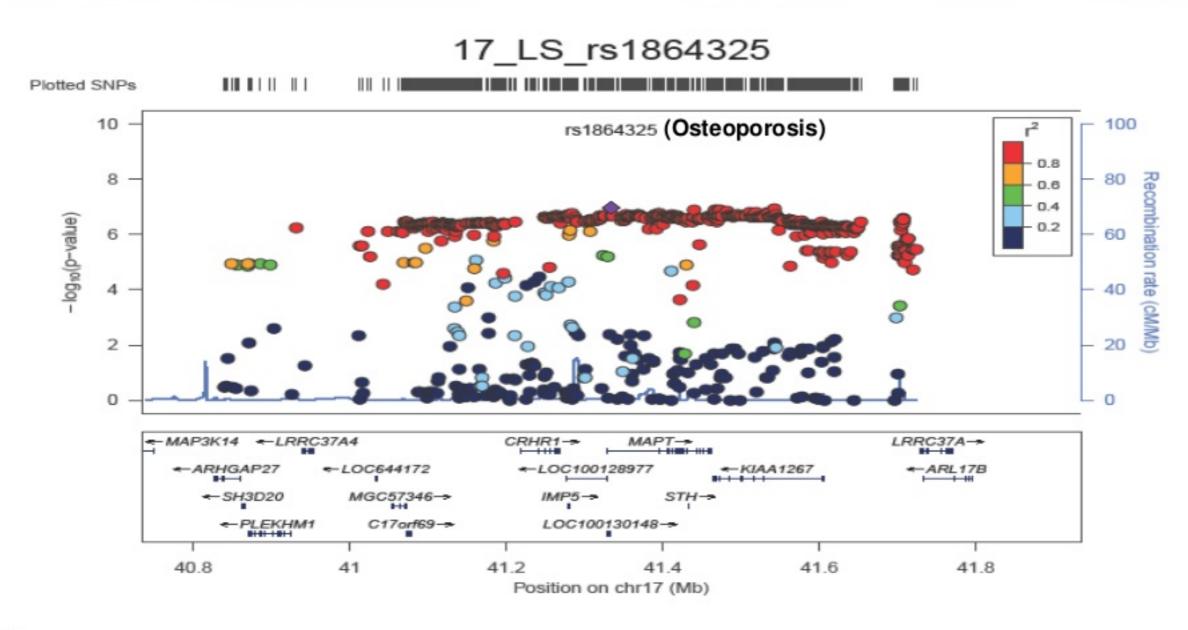


Associated Variants Located in Gene Desert

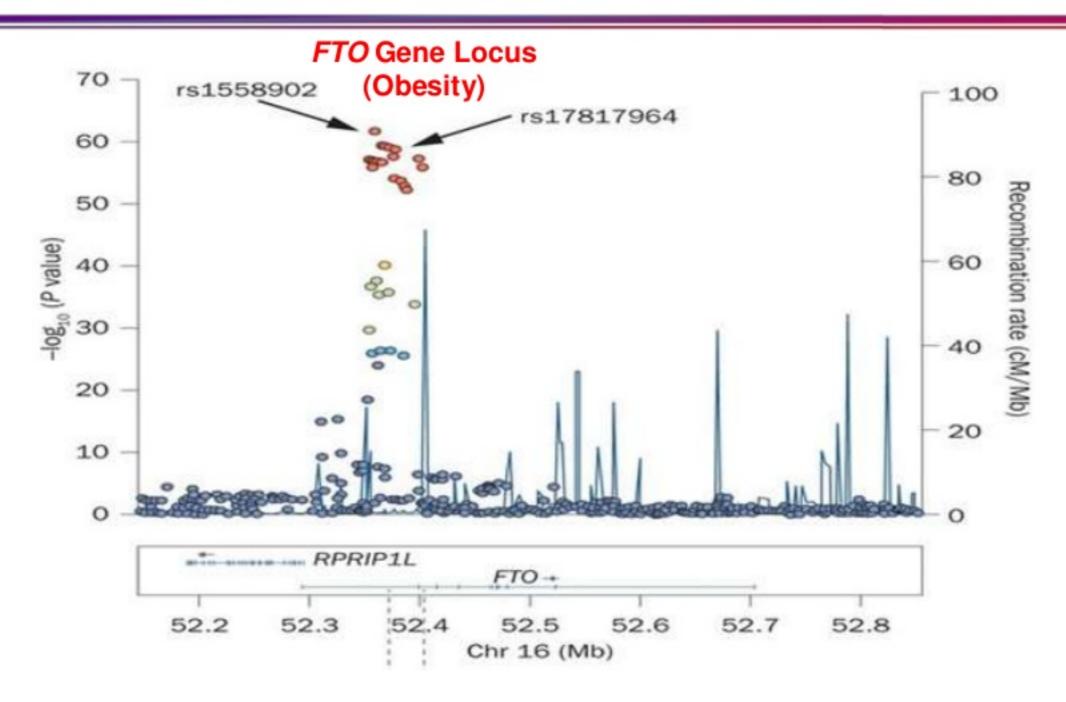


Genetic Coordination: 1D Physical Location on Linear DNA Sequences

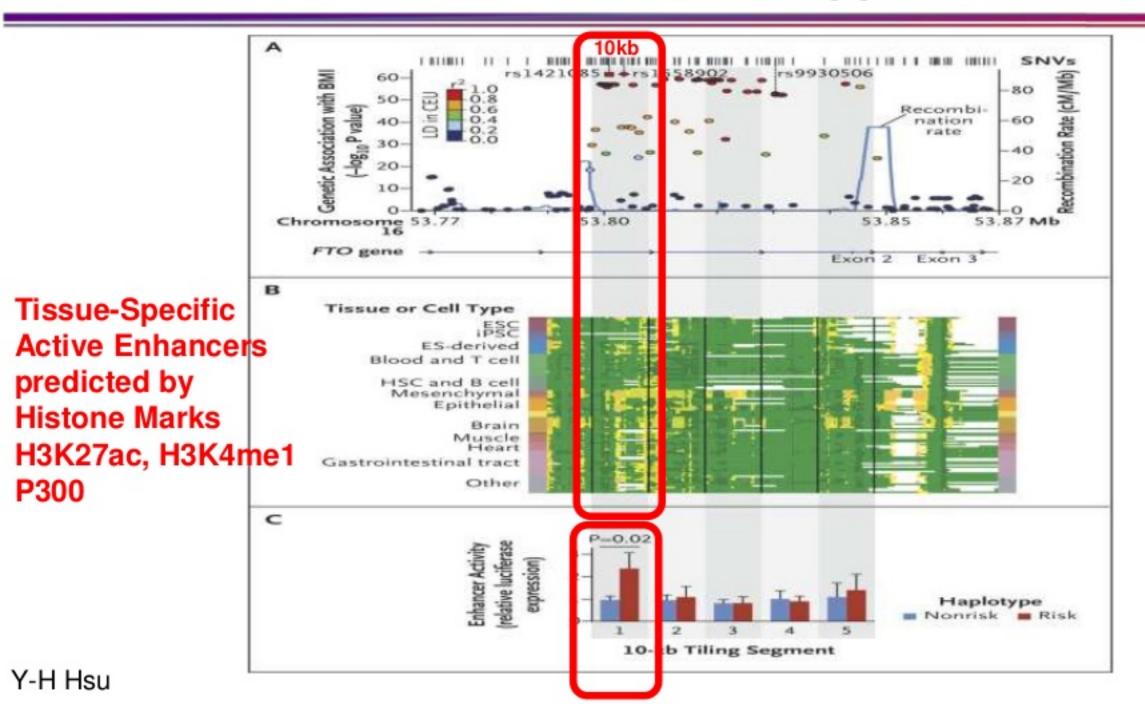
Too Many Genes: Which Gene(s)?



Associated Variants Located in Introns: Looks Promising?

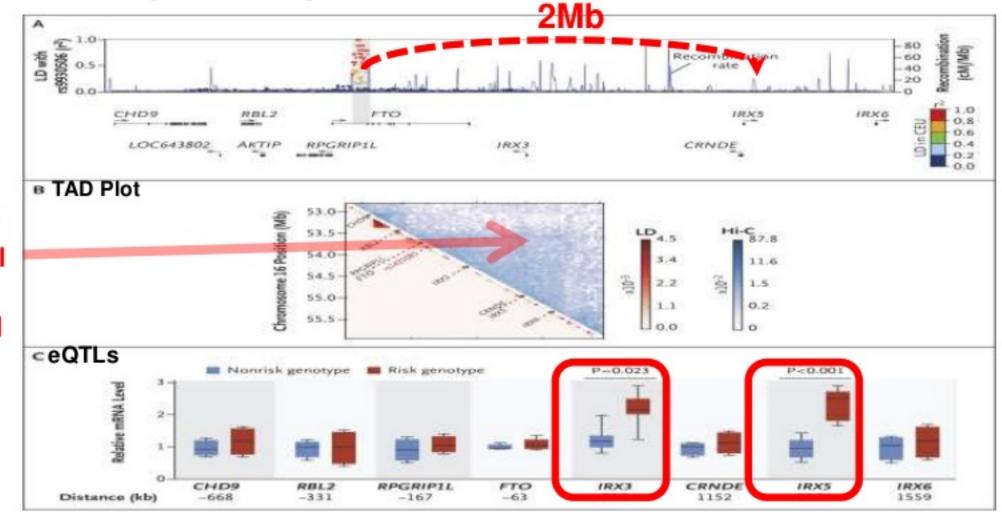


Functional Genomics Approaches



3D Genome Interaction Structure with IRX5 Gene

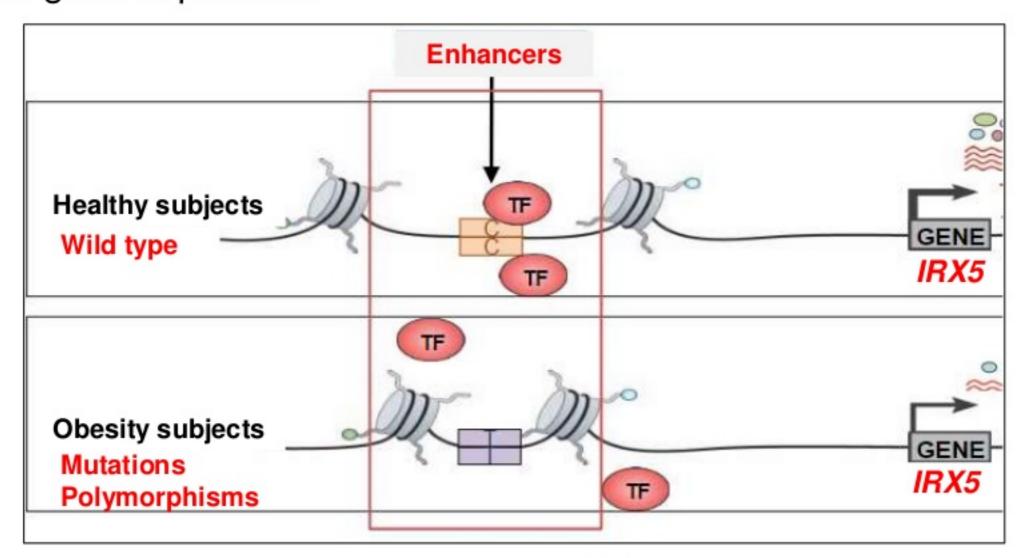
- ➤ Tissue-Specific Chromatin Confirmation Capture (3C Tech)
- eQTLs (associations between variants and gene expression)
- Allele-specific expression



Intensity of 3D Physical Interaction by Hi-C seq

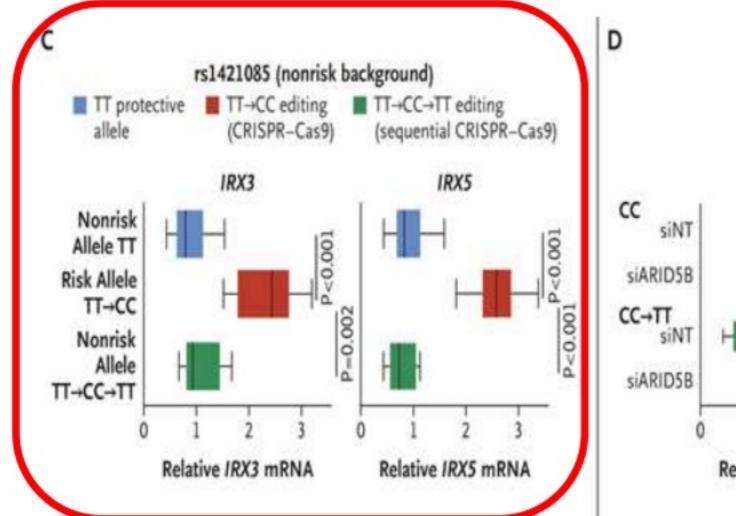
FTO Genetic Variants and IRX5 Gene Regulation

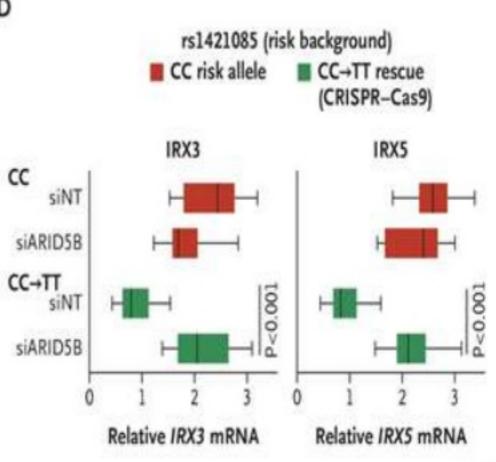
Obesity associated genetic variants disrupt TF binding and then reduce IRX5 gene expression



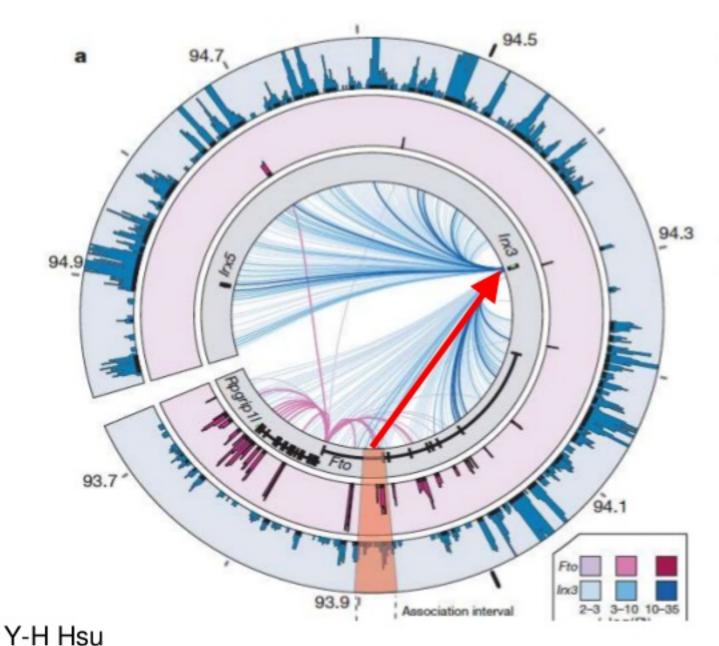
Gene-Editing: Functional Validation

- Gene Editing by CRISPR/Cas9 in Human adipocytes from subjects carried "risk allele" and subjects carried "protective allele"
- The Risk Allele C: Gain-of-function





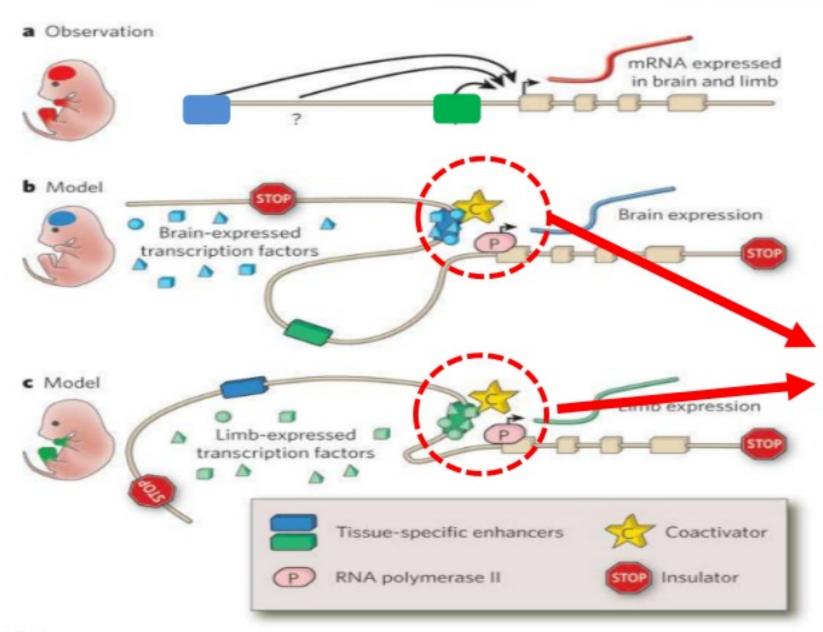
FTO Variants Link to Irx3 Gene in Brain



- The obesity associated variants physically interacts with promoter of *Irx3* gene, but not *Fto*, not *Irx5* in mouse brain by 4C-seq
- > 4C-seq: Regional Chromatin Confirmation Capture (3C Tech)

Nature, 2014

Gene Regulatory Models



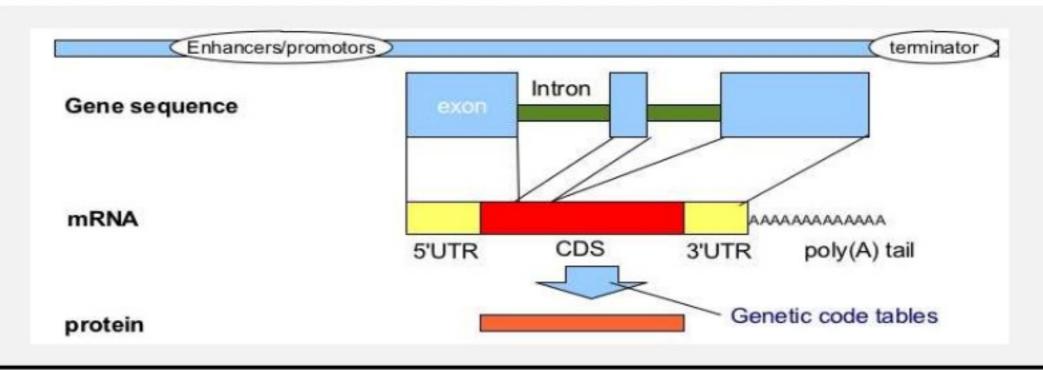
Gene regulatory elements in physical proximity (3D space) with the gene promoters via looping mechanisms

Tissue (Cell)-Specific DNA Loops: Enhancer-Promoter Interactions

Genome-Widely Identify/Predict Targeted Genes?

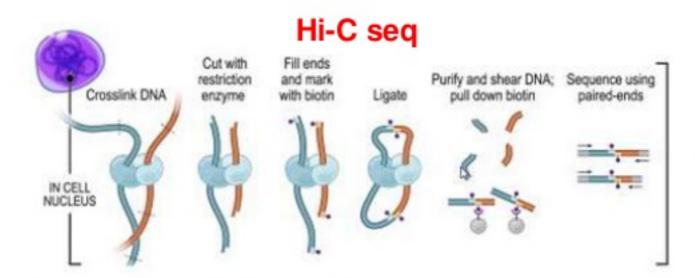
- Identified ~13,000 genetic variants (single nucleotide mutations/ polymorphisms) to be associated with ~2,000 diseases/phenotypes
 - 91% of disease-associated genetic variants are located in nonprotein-coding regions; used to call "junk DNA"
 - > Unknown function, difficult to translate findings into clinical use
 - May involve in tissue/cell type-specific gene regulation

Y-H Hslu

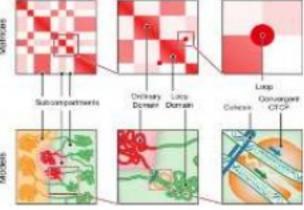


Chromosome Conformation Capture To Identify DNA Loops

3C, 4C, 5C, HiC, capture-HiC, etc to estimate 3D interaction among genome



Contact Map

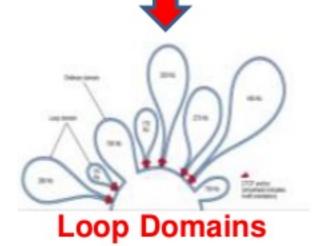


Science. 2009.; 326(5950): 289–293 Nat Rev Genet. 2010;11(6):439-46. Cell. 2014;159(7):1665-80 Nature Genetics 2016; 48, 488–496

Enhancer-Promoter

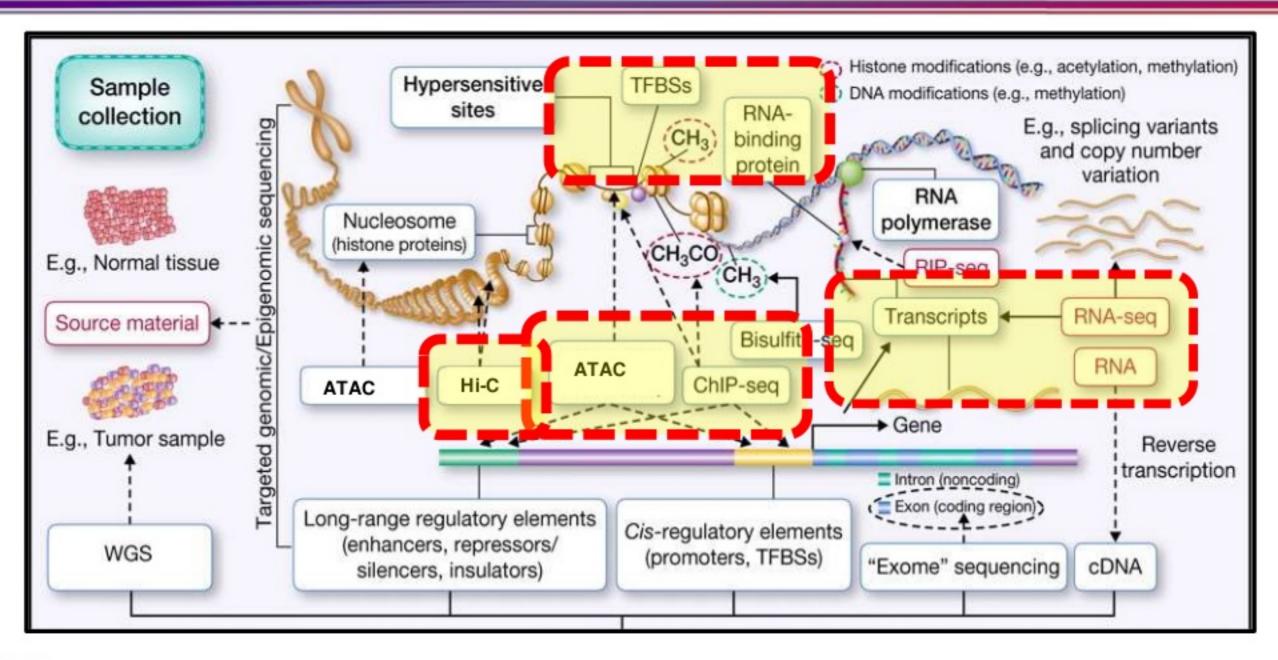
Enhancer-Enhancer Promoter-Promoter Physical Interactions False-Pos (seq error, miss-matched cutting,...)





Prediction

Building Tissue-Specific Gene Regulatory Circuits



Building Gene Regulatory Circuits On Human Heart

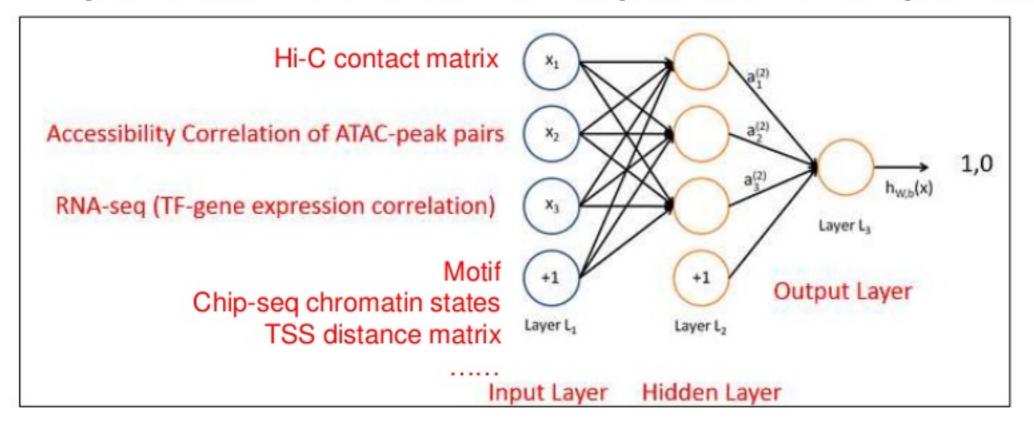
- Omics experiments on normal human primary cardiac fibroblasts and myocytes from atrium and ventricle; HMSC; skeletal muscle cells
- > Publicly available (low resolution): Left ventricle, right ventricle and aorta tissues

Experiments	Functions	Notes
ATAC-seq	Active cis-regulatory region	Active TF binding
Hi-C	Chromatin confirmation capture	1.5 to 2 kb resoultion (DpnII, 2 Billions Reads, 2Tb)
ChIP-seq	H3K4me3: Aactive promoter	
	H3K27ac: Active enhancer/promoter	
	CTCF: Insulator	
	Cohesin: Insulator-RAD21	Predicted chromatin states
	Cohesin: Insulator-SMC3	by HMM
	H3K27me3: Polycomb repressed/bivalent promoter/enhancer	
	H3K9me3: Heterochromatin	
	H3K36me3: Transcribed region	
RNA-seq	mRNA (active and)	Isoforms; coexpression with TF
	microRNA and small RNA	Enhancer RNA

Y-H Hsu

Model Gene Regulation with Deep Neural Network (DNN)

DNN implemented in the TensorFlow to predict enhancer-promoter gene pairs



- > Training sets (VISTA: enhancer elements are in 100kb of genes):
 - 1,564 Enhancer-promoter gene pairs (the positive set) functionally validated to have regulatory relationships in mouse models
 - 1,207 EP pairs without regulatory relationships (the negative set)

Acknowledgements





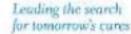












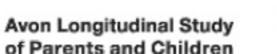
































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