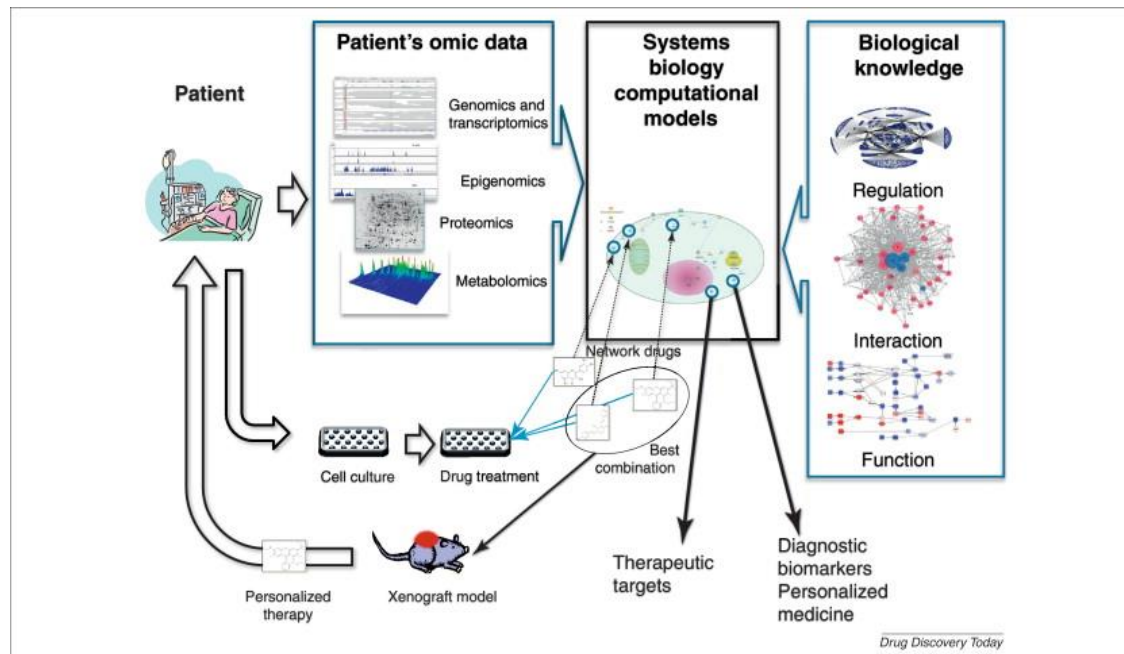


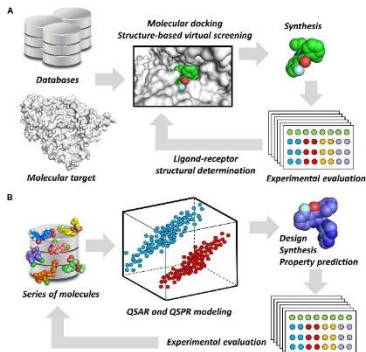
APPLICATION IN MEDICINE

- GENE FUNCTION AND REGULATION
- THERAPEUTIC APPROACHES

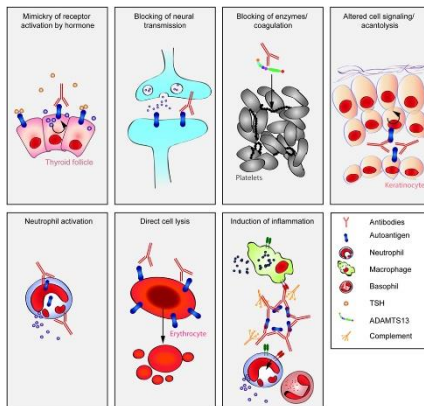


THERAPEUTIC APPROACHES

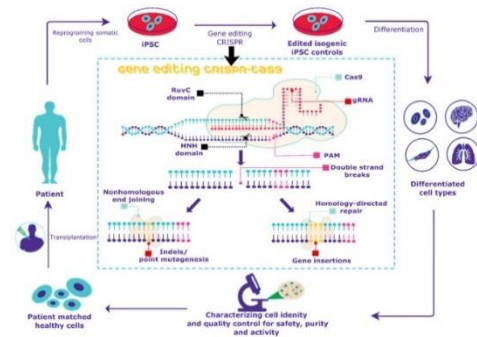
Chemical Drugs



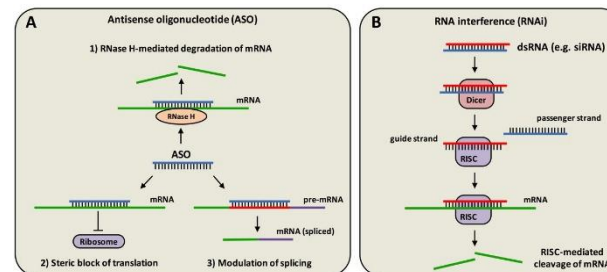
Antibodies



CRISPR-CAS SYSTEM



Oligonucleotides antisense



In this lesson

- What is the main focus of the course
- Definition of Functional Genomics
- How Functional Genomics is the basis for understanding diseases
- Application of Functional Genomics and Integration Data
- Role of single nucleotide variants in genomic regulatory regions: an example of functional genomics application

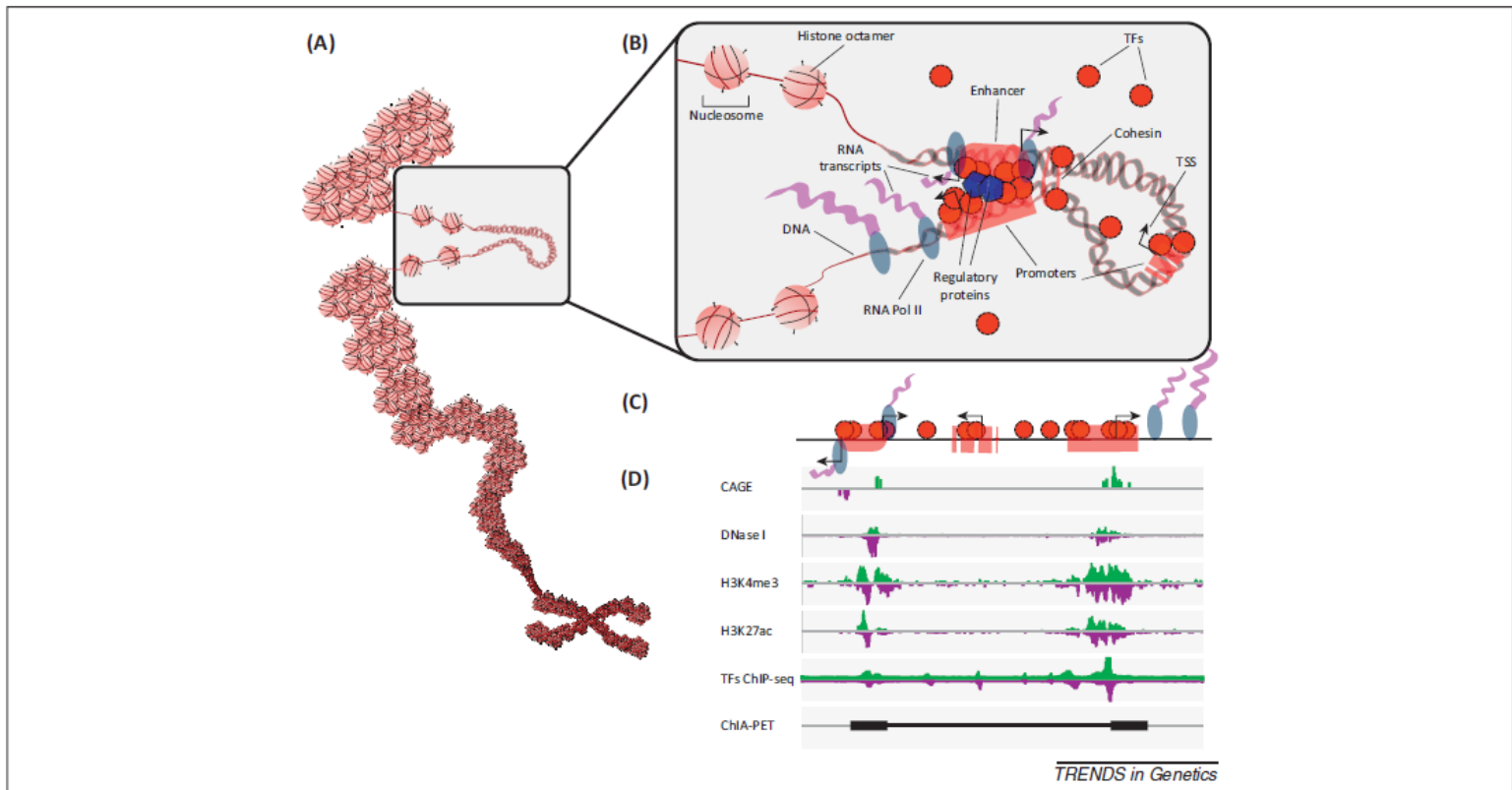


The main focus of this course is
functional genomics
APPLICATIONS ON MEDICINE



Functional genomics

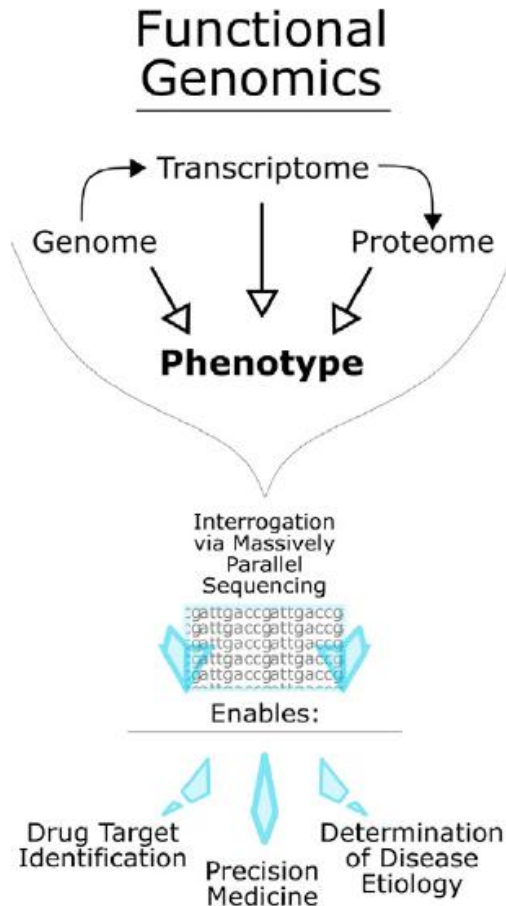
Functional genomics uses genomic data to study gene expression, regulation and biological functions on a global scale (genome-wide or system-wide), focusing on gene transcription, epigenetic modifications, chromatin remodelling enzymes, transcription factors association involving high-throughput methods.



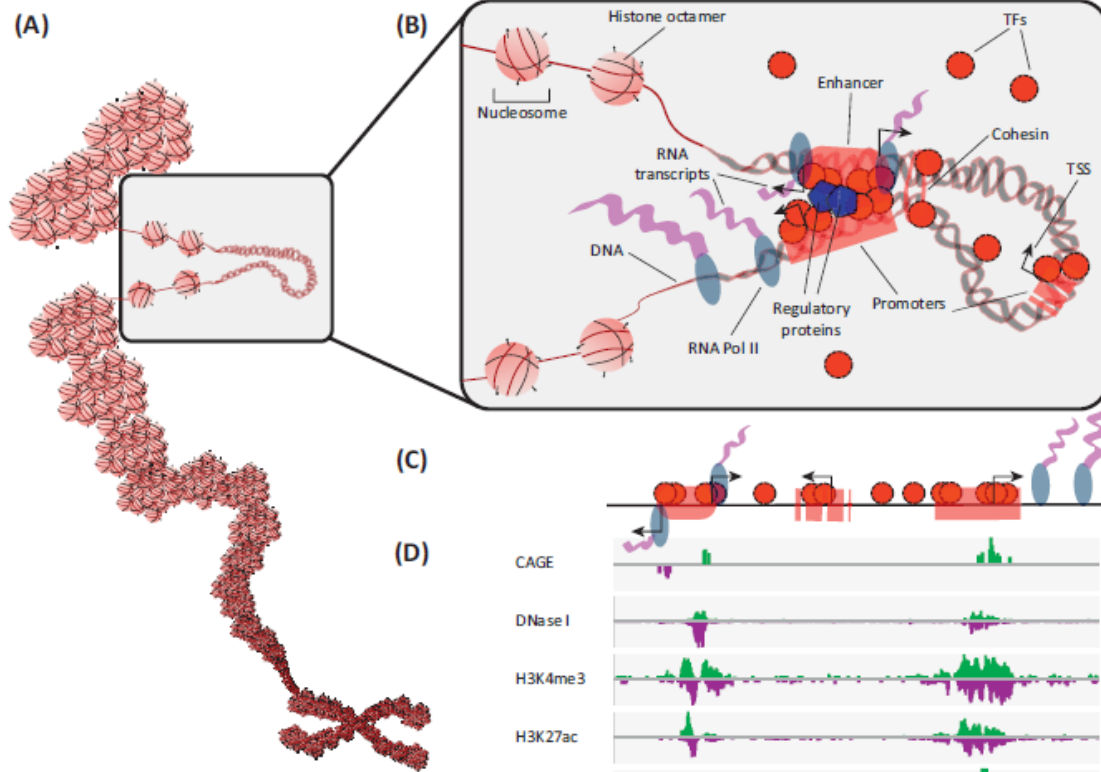
GENE REGULATION

How we can understand gene regulation
Using genome-wide sequencing data

- **FUNCTIONAL GENOMICS**



GENOMIC REGULATORY REGIONS



← GENE FEATURES

← GENE REGULATION

TRENDS in Genetics



GENE REGULATION

How we can understand gene regulation
Using genome-wide sequencing data

- **FUNCTIONAL GENOMICS**
- **INTEGRATION DATA APPROACH**

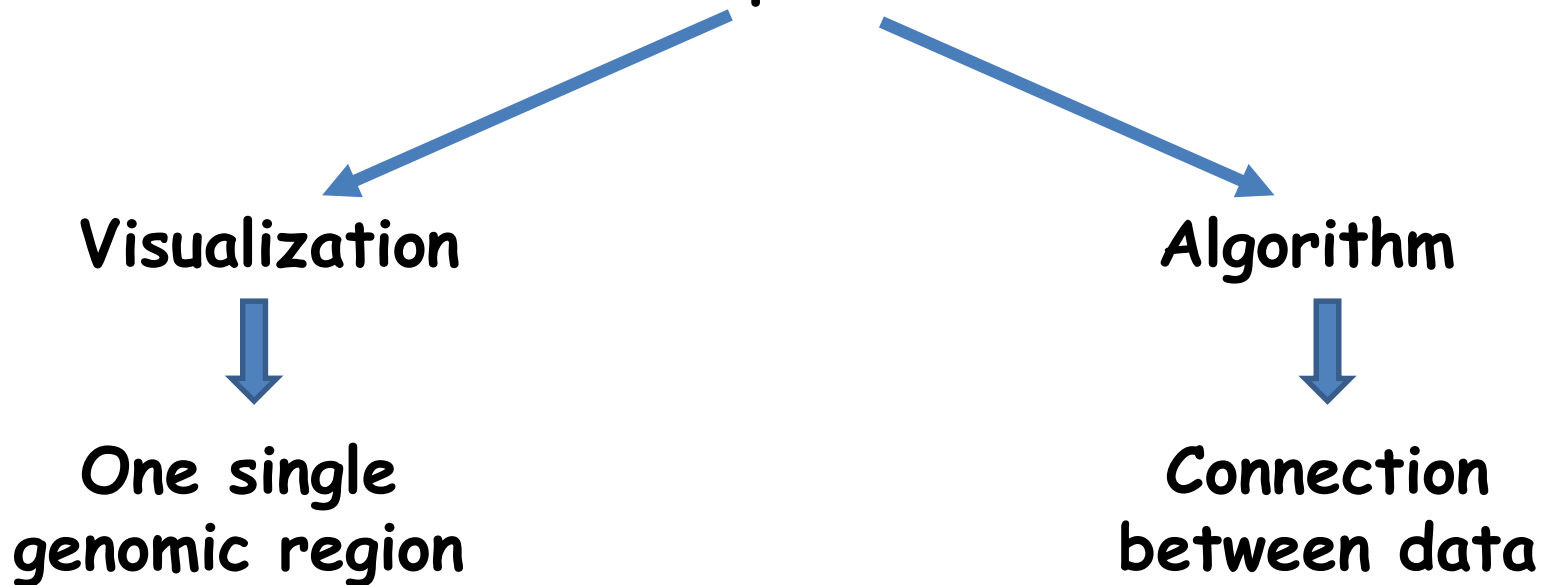


GENE REGULATION

How we can understand gene regulation
Using genome-wide sequencing data

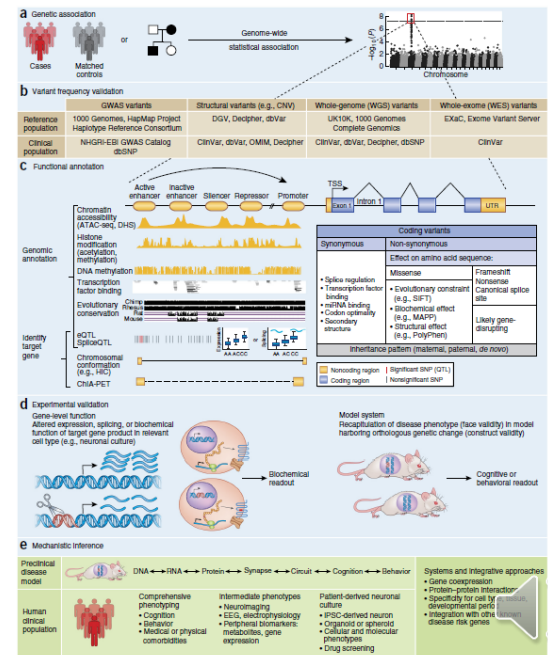
- **INTEGRATION DATA APPROACH**

Is based on the comparison of different data



Framework for interpretation of individual disease-associated variants

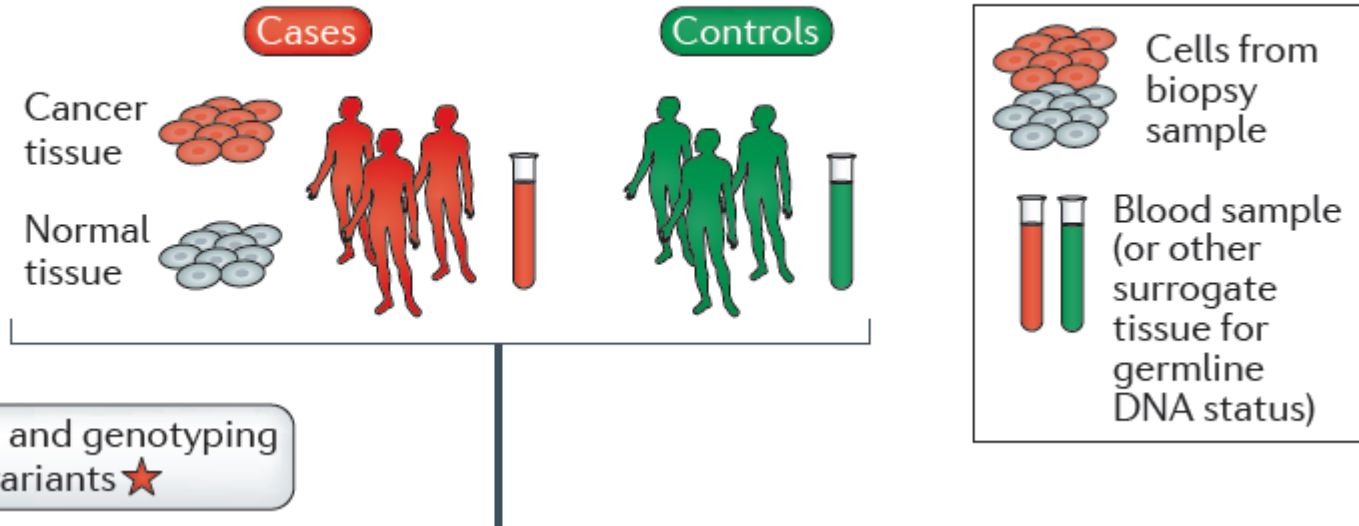
- Single nucleotide polymorphisms (SNPs) is the nucleotide variations associated with disease
- Genome-wide association studies (GWAS) have successfully identified thousands of common genetic variants associated with complex diseases (<http://www.ebi.ac.uk/gwas/>)
- Functional annotation: to define genomic regulatory regions by genome-wide integration data
- Experimental validation
- Disease Animal models
- Correlation between molecular mechanisms and disease symptoms
- Drug Discovery



SNPs with an impact in tumorigenesis

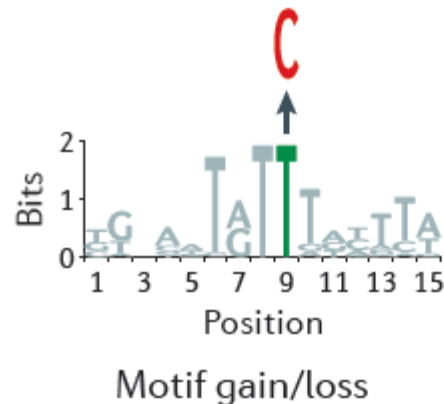
Steps for studying the role of SNP

1



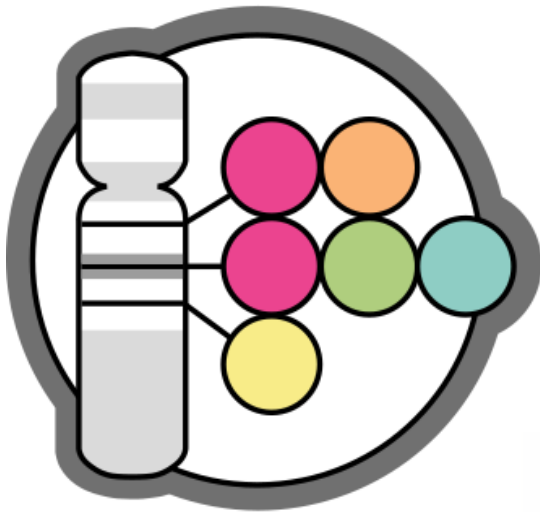
2

Computationally based functional prioritization and interpretation



Genome-wide association studies (GWAS) have capitalized on the millions of common single nucleotide polymorphisms (SNPs) to identify those SNPs that are genome-wide significantly associated with a disease or trait.

GWAS CATALOG



RETRIEVAL FUNCTIONS

`get_studies()`

`get_associations()`

`get_variants()`

`get_traits()`

S4 CLASSES

S studies

A associations

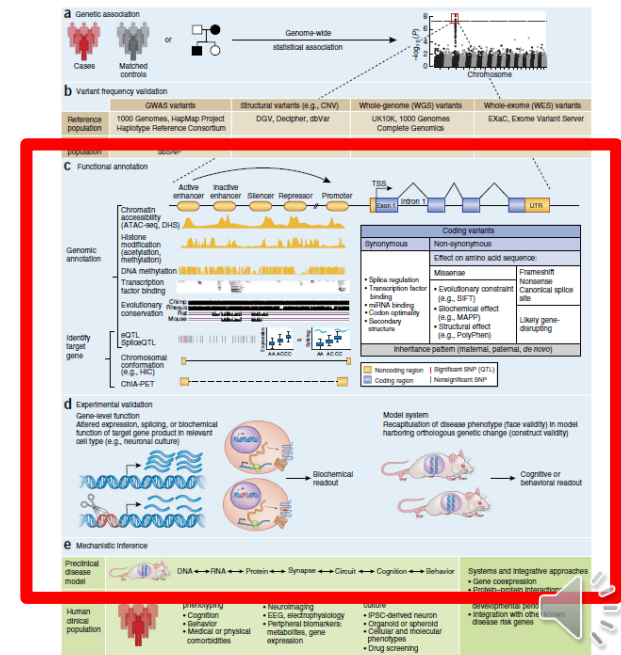
V variants

T traits



Framework for interpretation of individual disease-associated variants

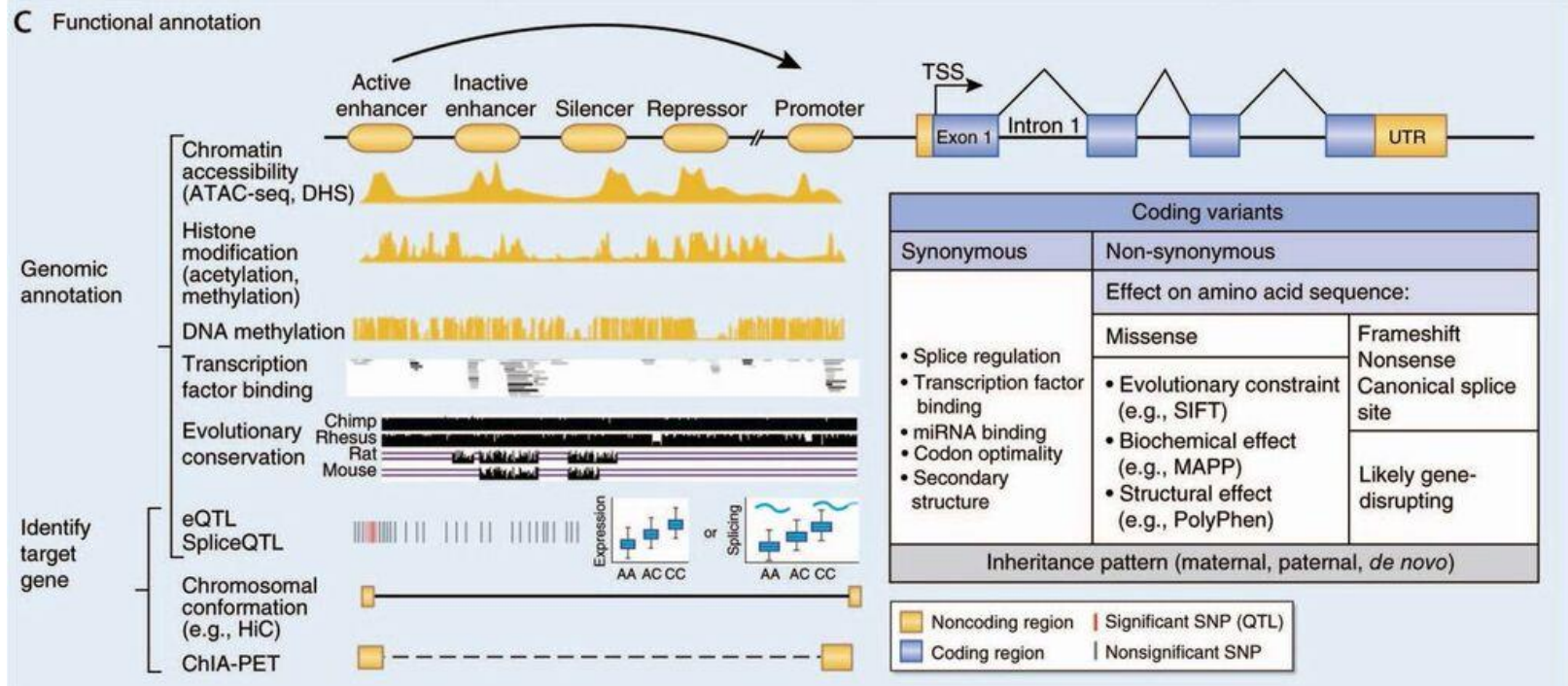
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- Correlation between molecular mechanisms and disease symptoms
- Drug Discovery



Functional Genomics

Functional genomics is a branch that integrates molecular biology and cell biology studies, and deals with the whole structure, function and regulation of a gene in contrast to the gene-by-gene approach of classical molecular biology technique.

From: [Encyclopedia of Bioinformatics and Computational Biology, 2019](#)



GENE REGULATION

How we can understand gene regulation Using genome-wide sequencing data

● INTEGRATION DATA APPROACH

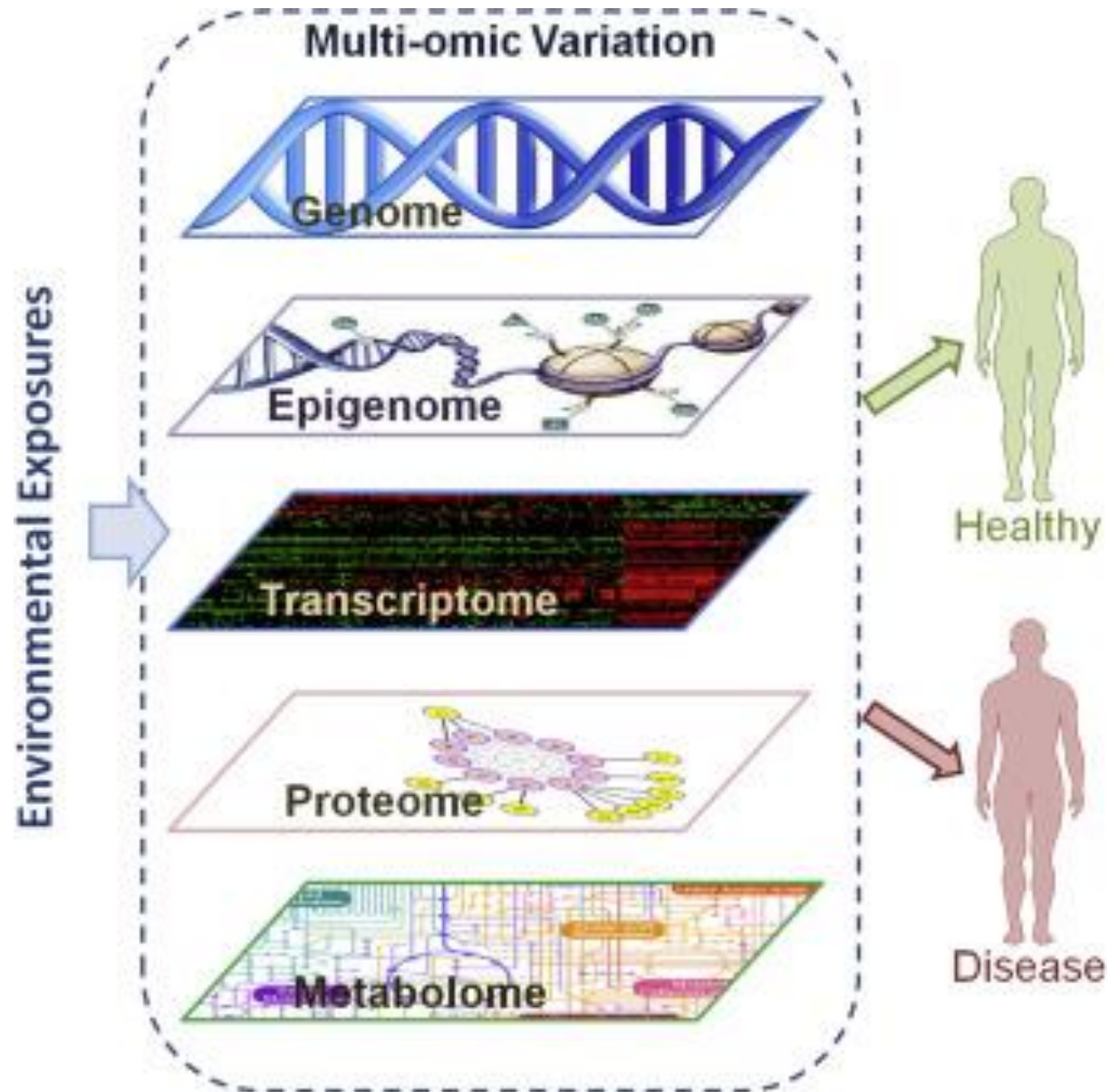
Integrative omics for health and disease

Konrad J. Karczewski^{1,2} and Michael P. Snyder³

Abstract | Advances in omics technologies — such as genomics, transcriptomics, proteomics and metabolomics — have begun to enable personalized medicine at an extraordinarily detailed molecular level. Individually, these technologies have contributed medical advances that have begun to enter clinical practice. However, each technology individually cannot capture the entire biological complexity of most human diseases. Integration of multiple technologies has emerged as an approach to provide a more comprehensive view of biology and disease. In this Review, we discuss the potential for combining diverse types of data and the utility of this approach in human health and disease. We provide examples of data integration to understand, diagnose and inform treatment of diseases, including rare and common diseases as well as cancer and transplant biology. Finally, we discuss technical and other challenges to clinical implementation of integrative omics.

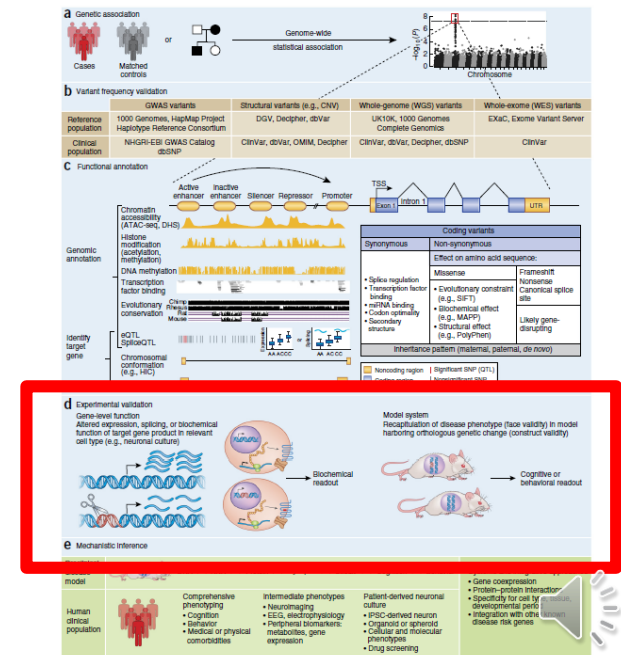


Integrative Omics for health and disease

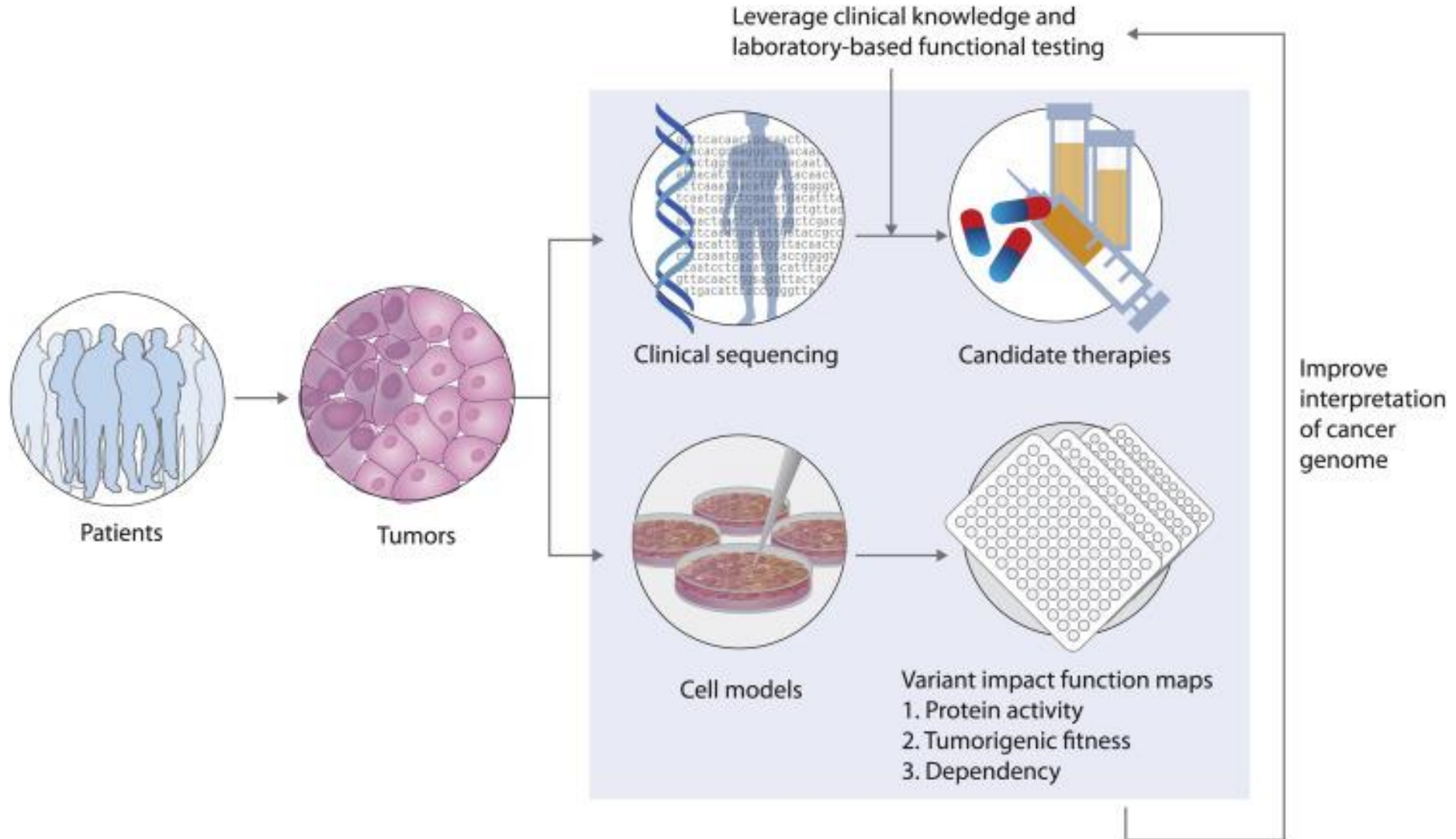


Framework for interpretation of individual disease-associated variants

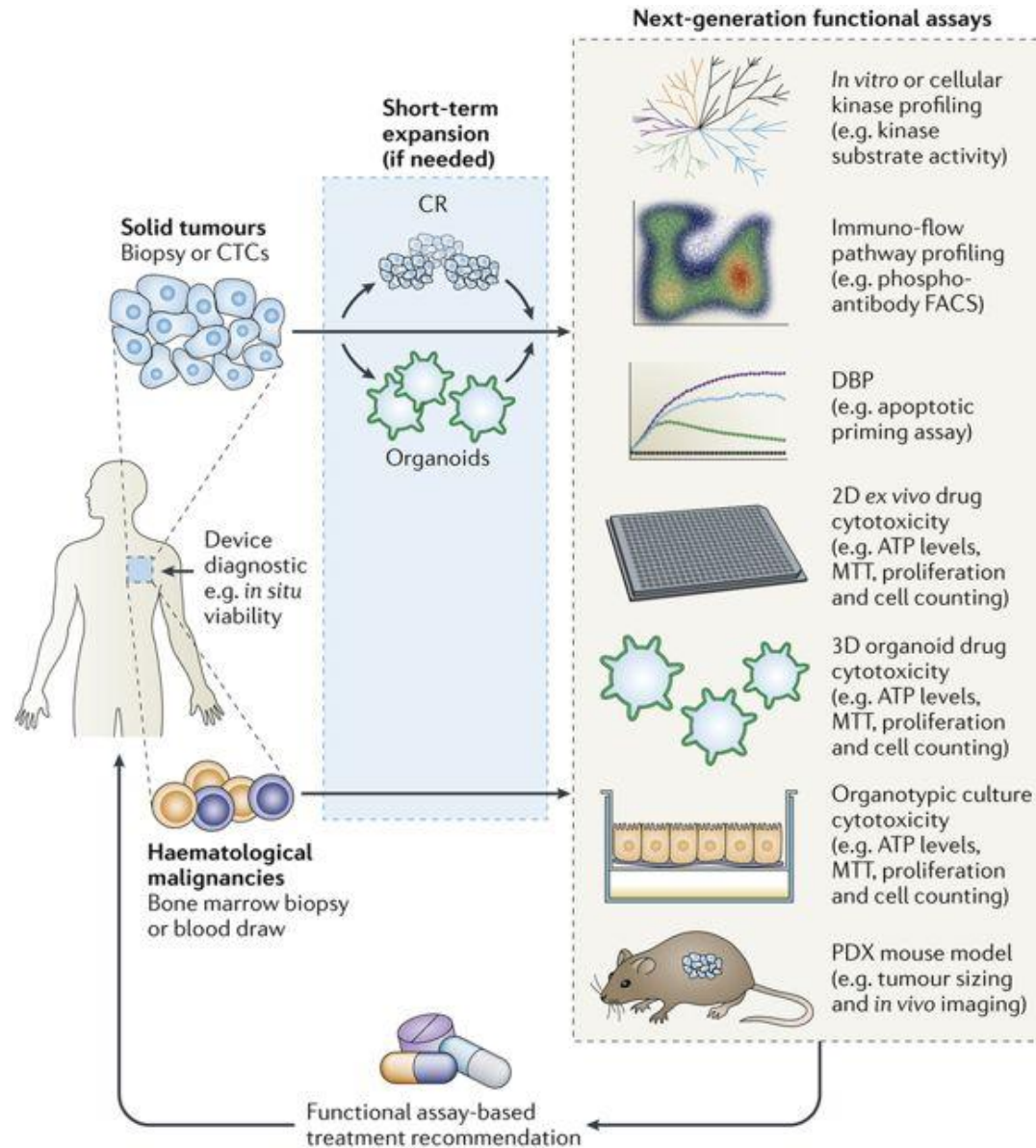
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- **Disease Animal models**
- Correlation between molecular mechanisms and disease symptoms
- Drug Discovery



Functional genomics uses genome-wide data with functional tests

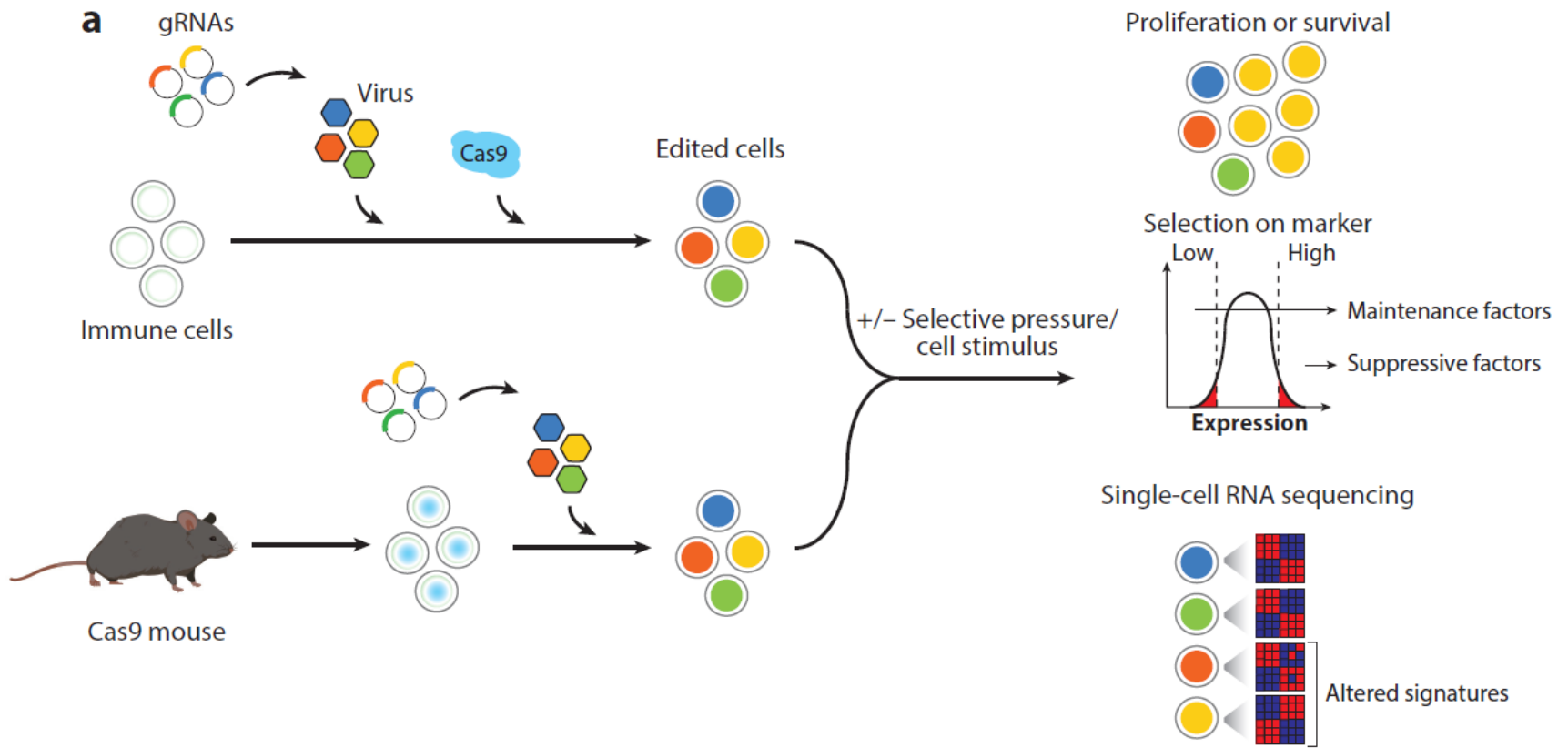
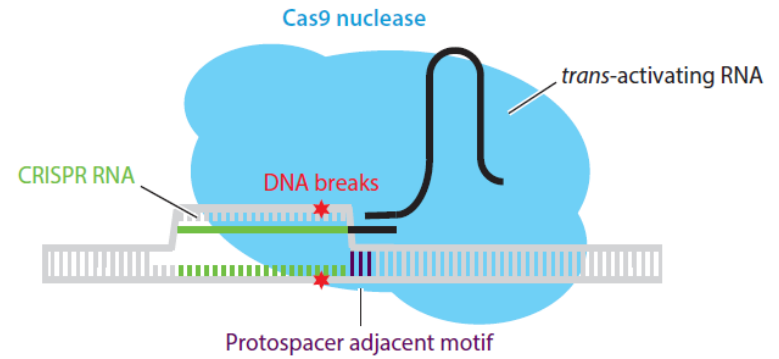


Functional genomics uses genome-wide data with functional tests

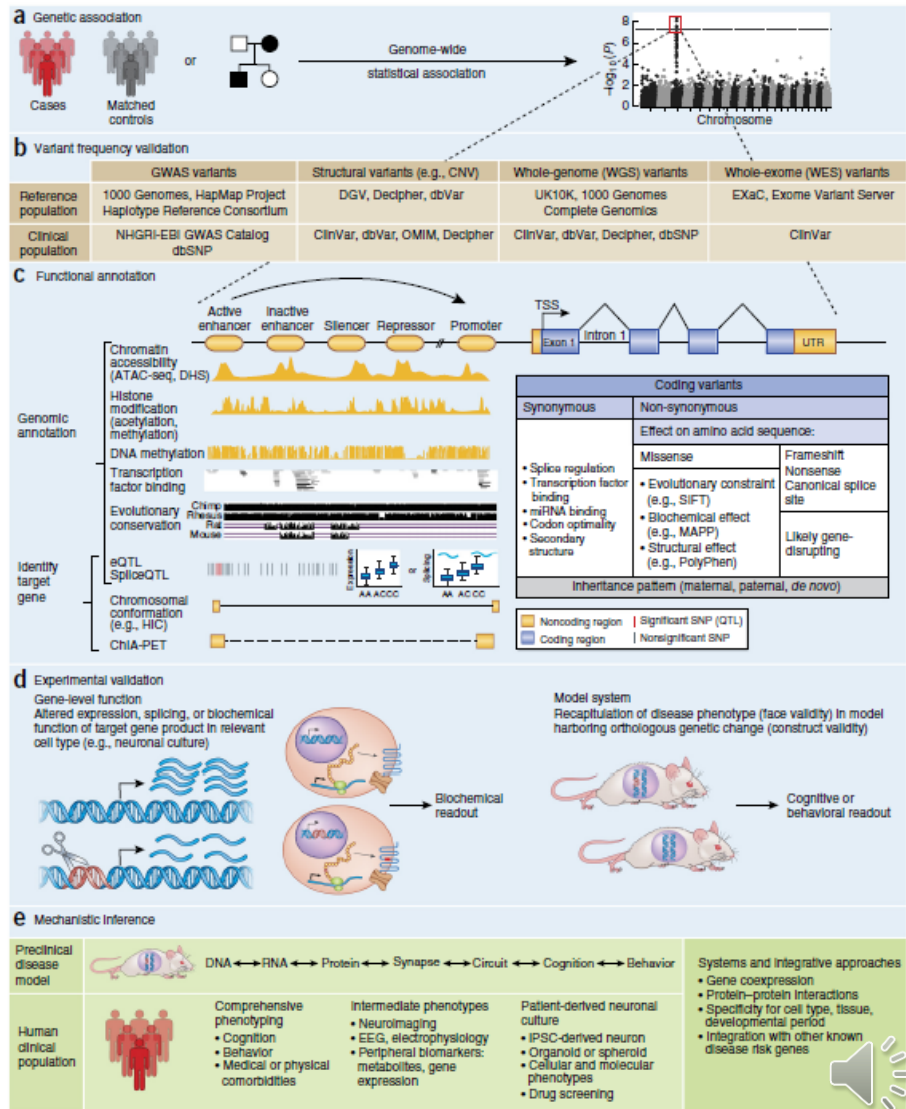


CRISPR-Based Tools in Immunity

Dimitre R. Simeonov^{1,2,3} and Alexander Marson^{2,3,4,5,6,7}



Which are the steps to understand the SNPs meaning?



Experimental validation ●

Disease Animal models ●

Correlation between molecular mechanisms and disease symptoms ●

Drug Discovery ●

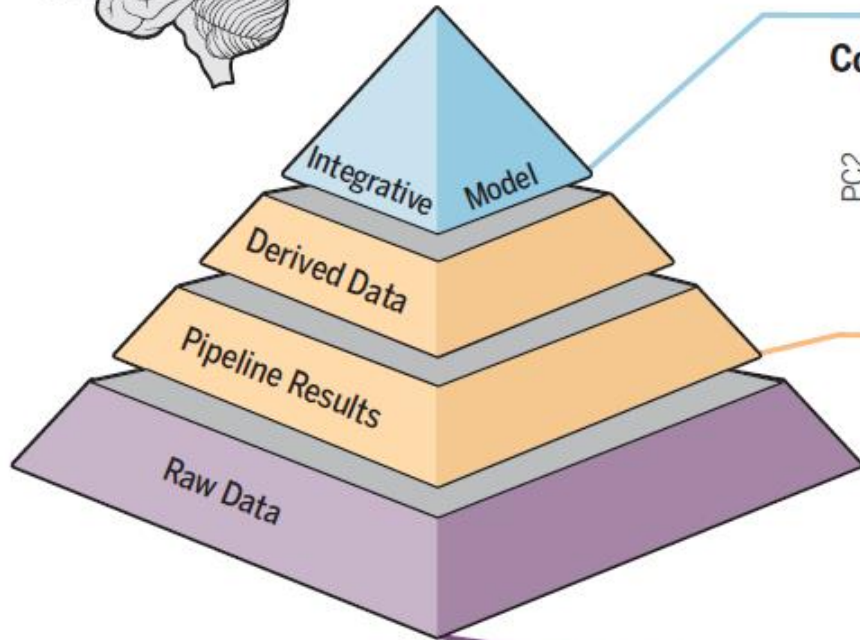
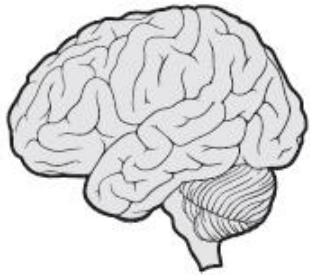


Comprehensive functional genomic resource and integrative model for the human brain

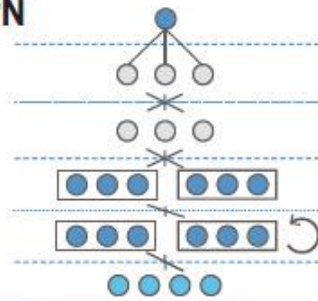
Daifeng Wang*, Shuang Liu*, Jonathan Warrell*, Hyejung Won*, Xu Shi*, Fabio C. P. Navarro*, Declan Clarke*, Mengting Gu*, Prashant Emani*, Yucheng T. Yang, Min Xu, Michael J. Gandal, Shaoke Lou, Jing Zhang, Jonathan J. Park, Chengfei Yan, Suhm Kyong Rhie, Kasidet Manakongtreecheep, Holly Zhou, Aparna Nathan, Mette Peters, Eugenio Mattei, Dominic Fitzgerald, Tonya Brunetti, Jill Moore, Yan Jiang, Kiran Girdhar, Gabriel E. Hoffman, Selim Kalayci, Zeynep H. Gümüŝ, Gregory E. Crawford, PsychENCODE Consortium†, Panos Roussos, Schahram Akbarian, Andrew E. Jaffe, Kevin P. White, Zhiping Weng, Nenad Sestan, Daniel H. Geschwind‡, James A. Knowles‡, Mark B. Gerstein‡



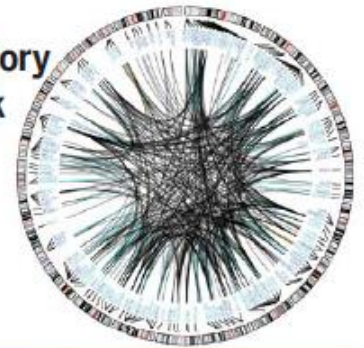
Functional genomic resource and integrative model for the human brain



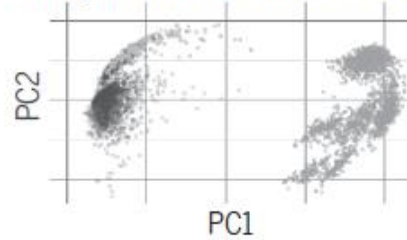
DSPN



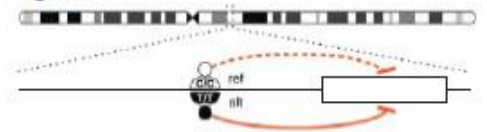
Regulatory network



Comparison to other tissues



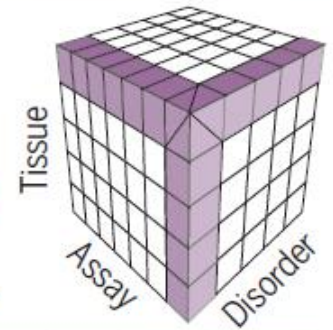
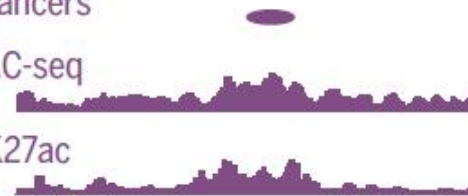
QTLs



Enhancers

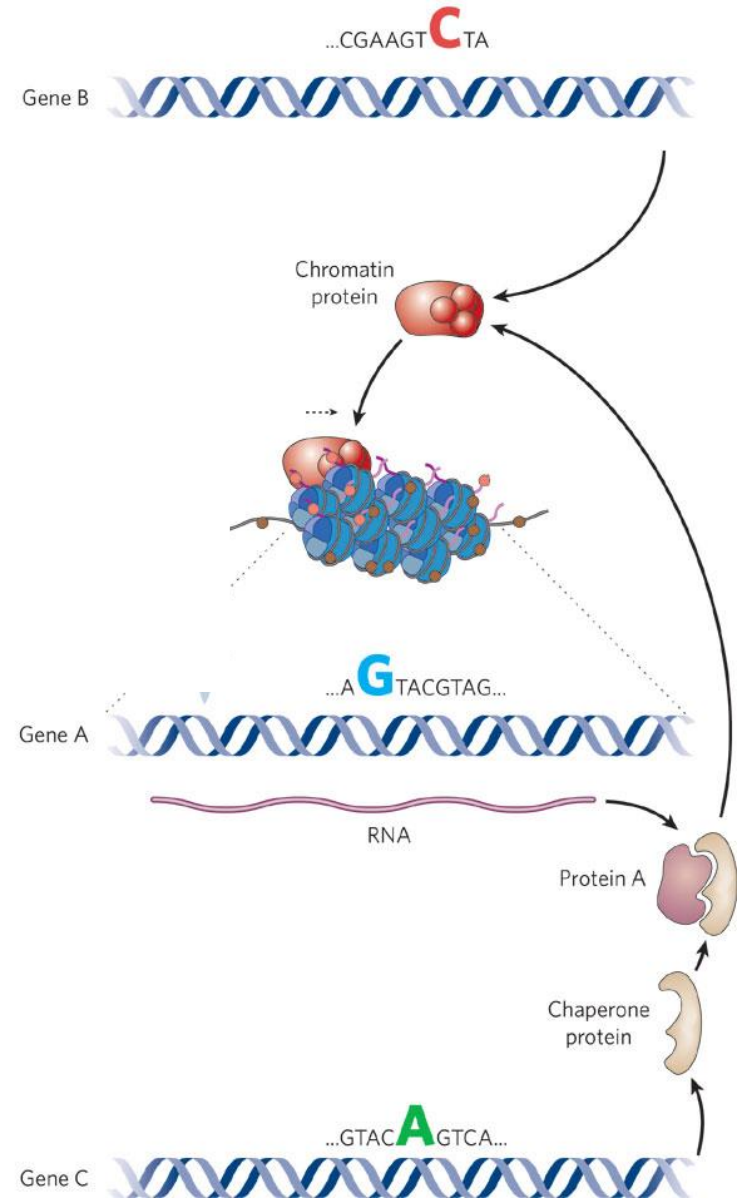
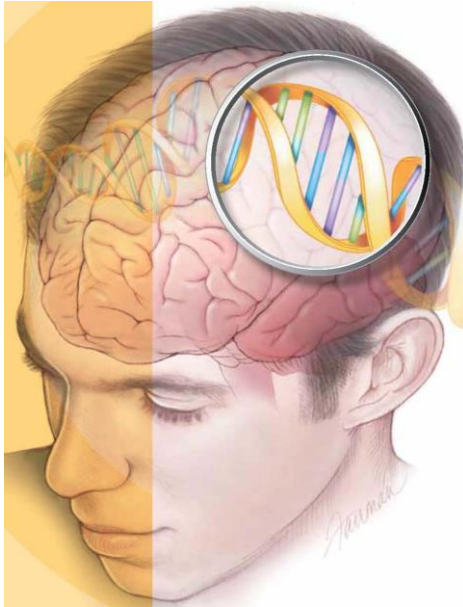
ATAC-seq

H3K27ac



INTRODUCTION: Strong genetic associations have been found for a number of psychiatric disorders. However, understanding the underlying molecular mechanisms remains challenging.

RATIONALE: To address this challenge, the PsychENCODE Consortium has developed a comprehensive online resource and integrative models for the functional genomics of the human brain.



RESULTS: The base of the pyramidal resource is the datasets generated by PsychENCODE, including bulk transcriptome, chromatin, genotype, and Hi-C datasets and single-cell transcriptomic data from ~32,000 cells for major brain regions. We have merged these with data from Genotype-Tissue Expression (GTEx), ENCODE, Roadmap Epigenomics, and single-cell analyses. Via uniform processing, we created a harmonized resource, allowing us to survey functional genomics data on the brain over a sample size of 1866 individuals.

1. New genome-wide data

2. Comparison of New genome-wide data with data derived from several databases



From this uniformly processed dataset, we created derived data products. These include lists of brain-expressed genes, coexpression modules, and single-cell expression profiles for many brain cell types; ~79,000 brain-active enhancers with associated Hi-C loops and topologically associating domains; and ~2.5 million expression quantitative-trait loci (QTLs) comprising ~238,000 linkage-disequilibrium-independent single-nucleotide polymorphisms and of other types of QTLs associated with splice isoforms, cell fractions, and chromatin activity. By using these, we found that >88% of the cross-population variation in brain gene expression can be accounted for by cell fraction changes. Furthermore, a number of disorders and aging

ON OUR WEBSITE

Read the full article at <http://dx.doi.org/10.1126/science.aat8464>

that the brain has distinct expression and epigenetic patterns, including a greater extent of noncoding transcription than other tissues.

are associated with changes in cell-type proportions. The derived data also enable comparison between the brain and other tissues. In particular, by using spectral analyses, we found

3. Interpretation of data

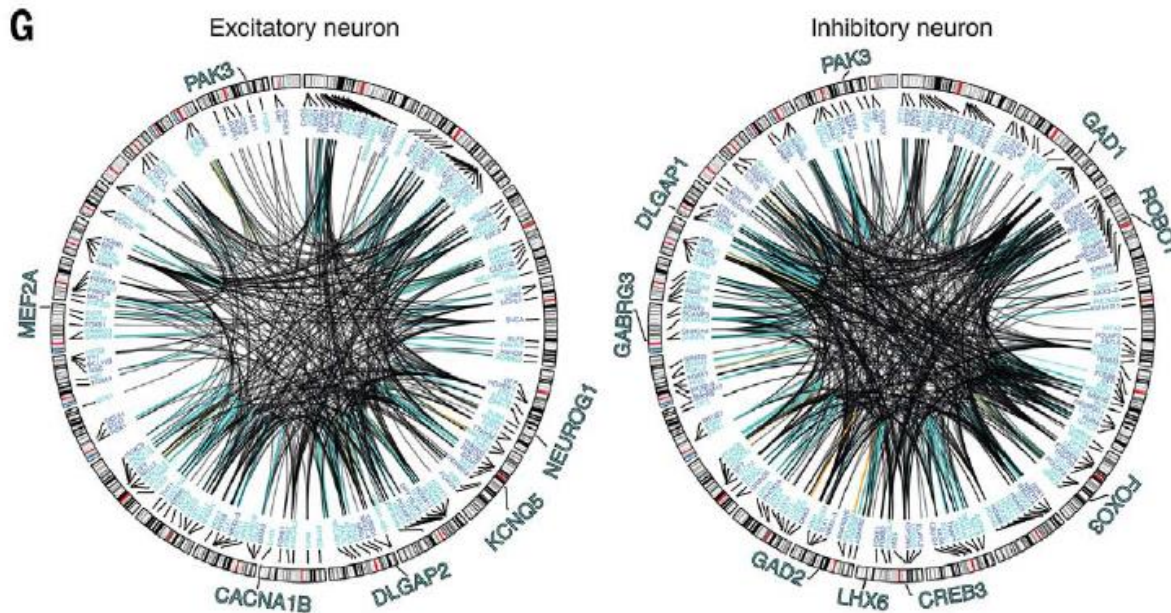
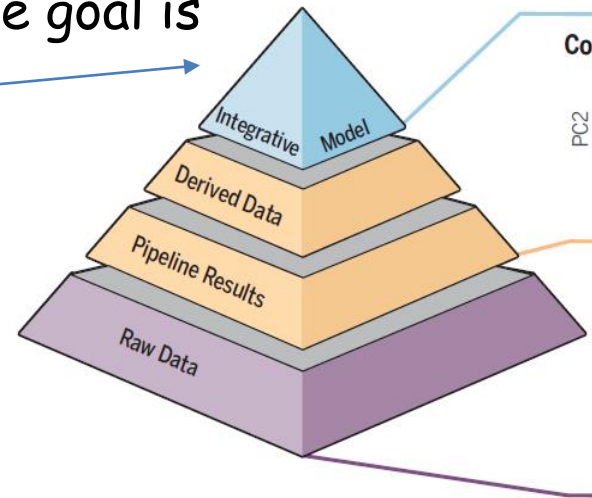
4. Gene expression profile is specific for a group of cells as shown by Single Cell-RNA-Seq



5. Identification of SNPs with specific functions

The top level of the resource consists of integrative networks for regulation and machine-learning models for disease prediction. The networks include a full gene regulatory network (GRN) for the brain, linking transcription factors, enhancers, and target genes from merging of the QTLs, generalized element-activity correlations, and Hi-C data. By using this network, we link disease genes to genome-wide association study (GWAS) variants for psychiatric disorders. For schizophrenia, we linked 321 genes to the 142 reported GWAS loci. We

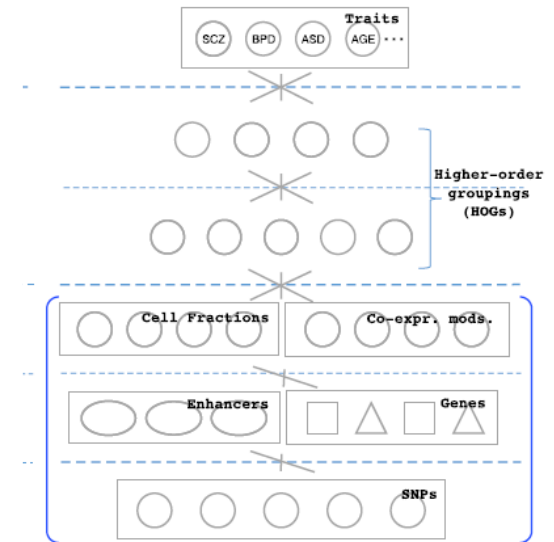
The goal is



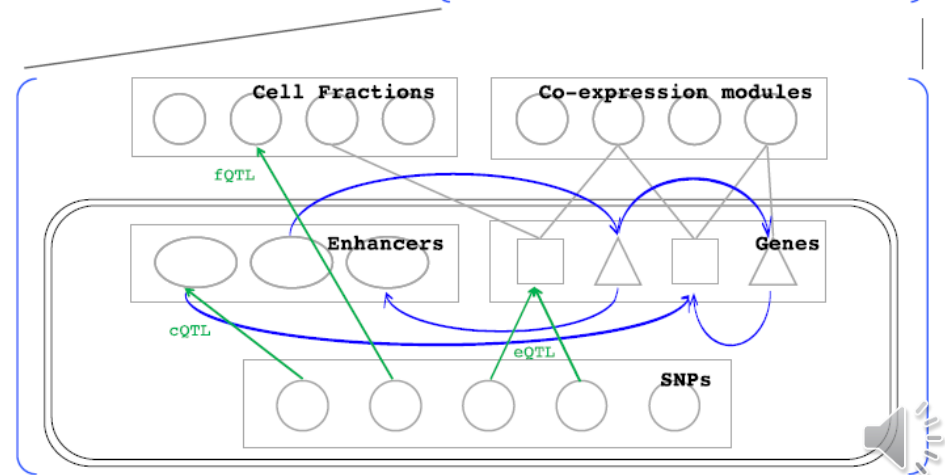
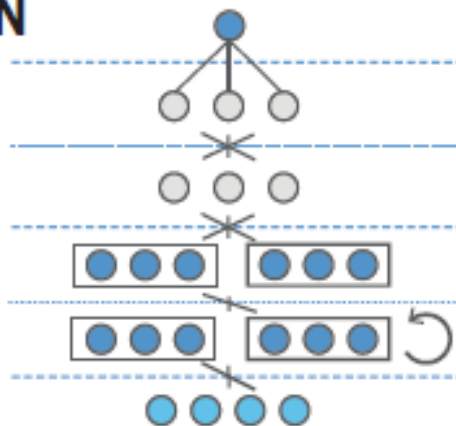
How integrative analysis is done?

321 genes to the 142 reported GWAS loci. We then embedded the regulatory network into a deep-learning model to predict psychiatric phenotypes from genotype and expression. Our model gives a ~6-fold improvement in prediction over additive polygenic risk scores. Moreover, it achieves a ~3-fold improvement over additive models, even when the gene expression data are imputed, highlighting the value of having just a small amount of transcriptome data for disease prediction. Lastly, it highlights key genes and pathways associated with disorder prediction, including immunological, synaptic, and metabolic pathways, recapitulating de novo results from more targeted analyses.

- Gene regulatory network
- **Deep-learning model**



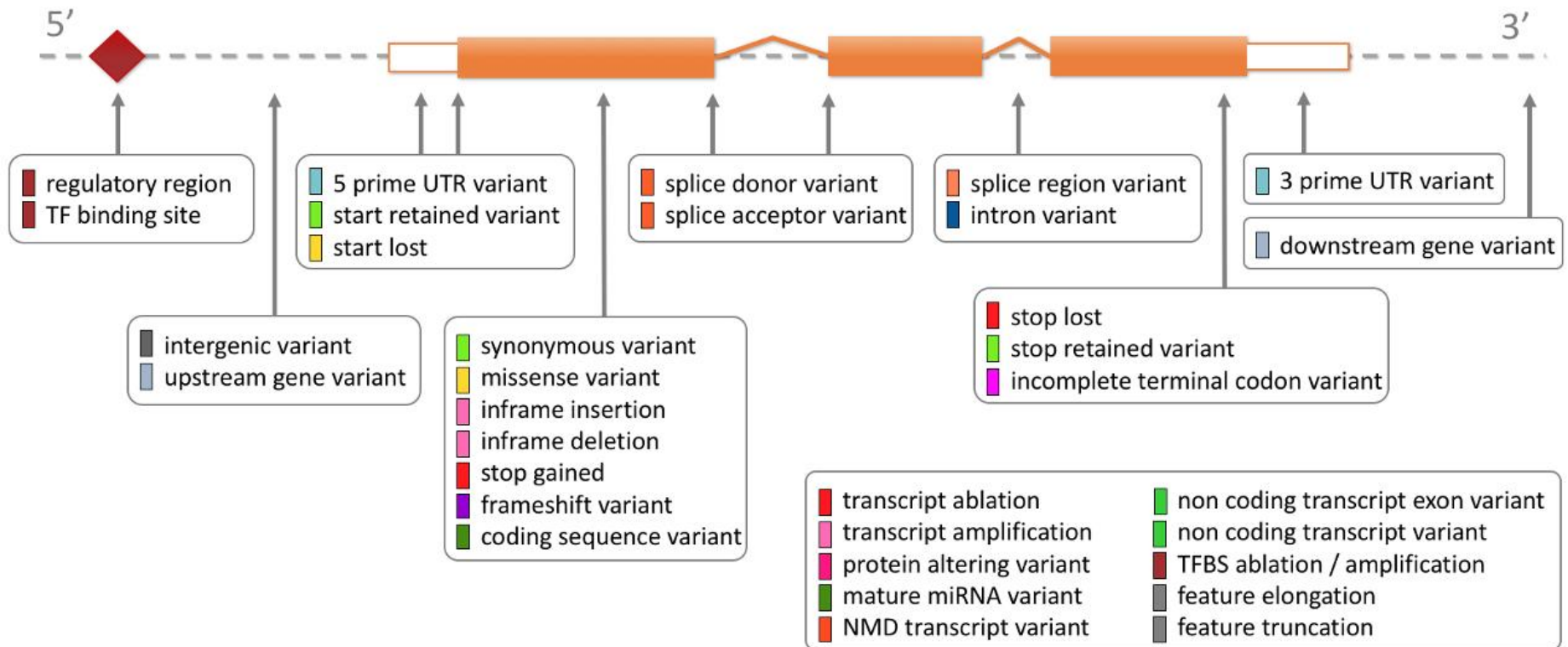
DSPN



Task 2- What is the impact of single nucleotide variants



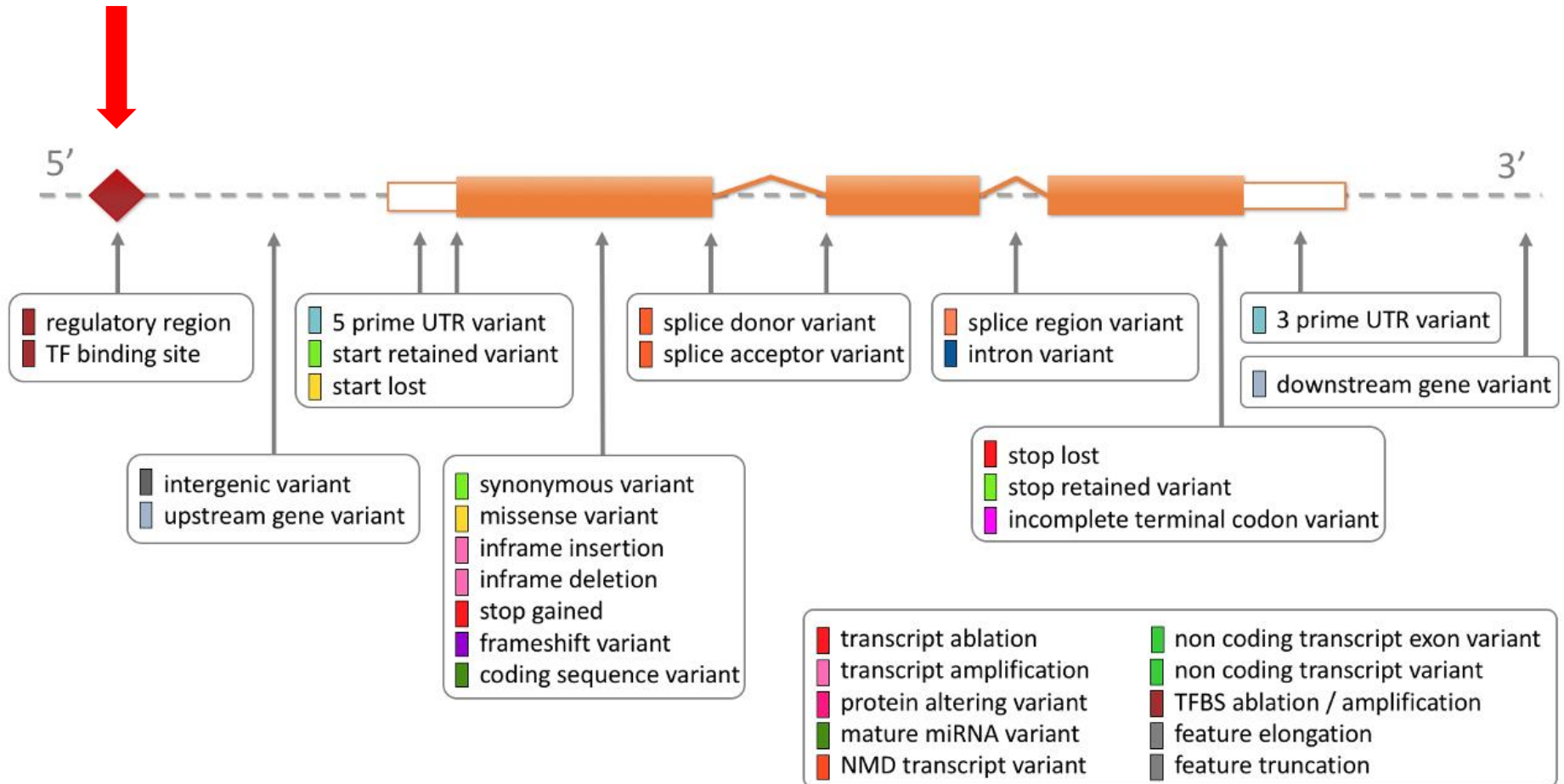
CLASSIFICATION OF SINGLE NUCLEOTIDE VARIANTS



https://m.ensembl.org/info/genome/variation/prediction/predicted_data.html

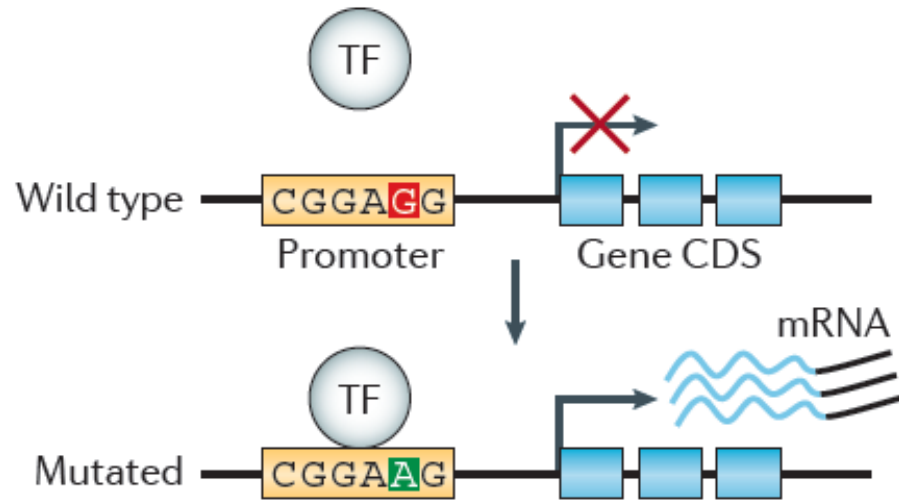


Role of single nucleotide variants in the genomic regulatory regions

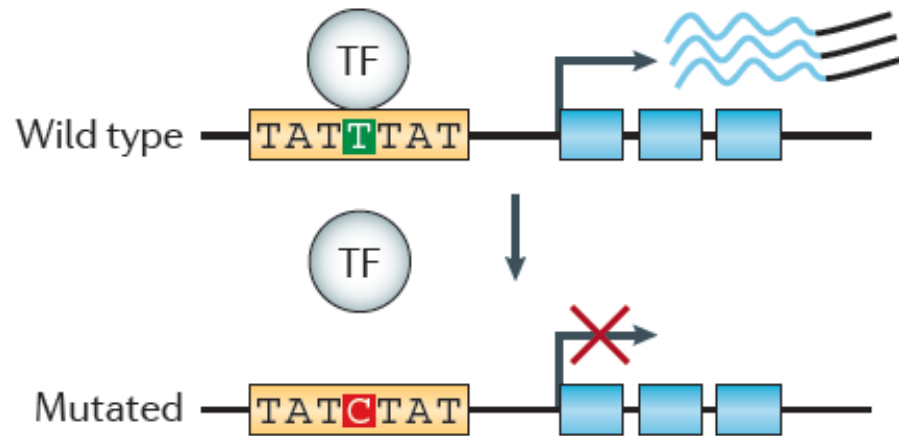
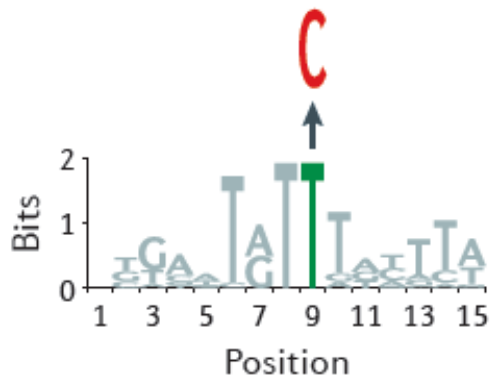


SNPs types functions at trascription factor binding sites:

Ba Gain of motif

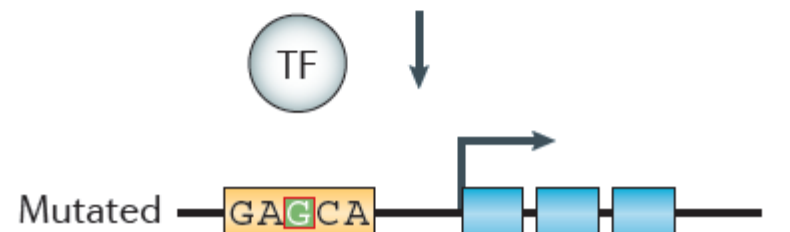
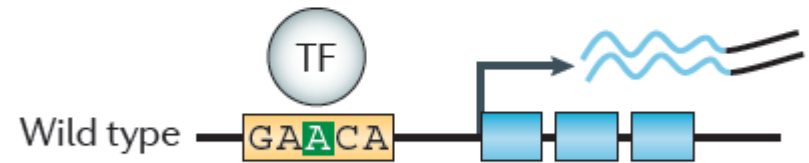
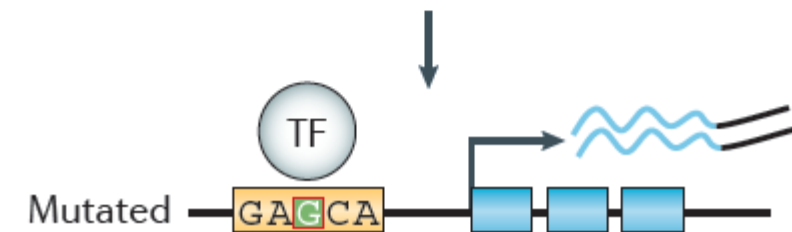
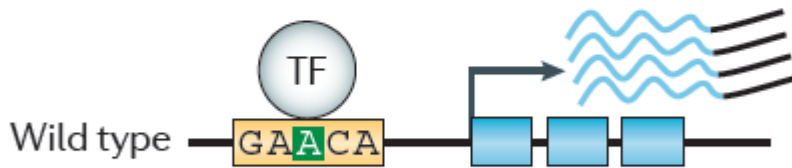
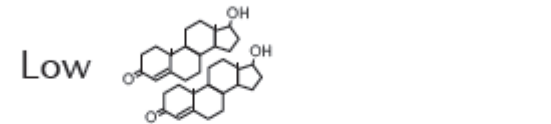
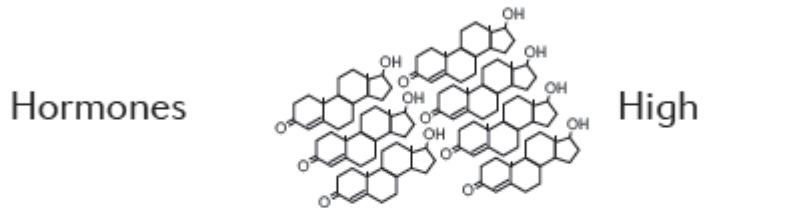
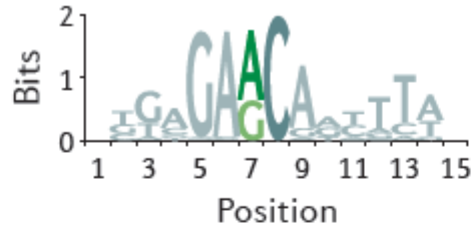


Bb Loss of motif



SNPs types functions on hormone response:

Bc Altered binding effects in hormonal cancers



SNPs in the genomic regulatory regions DEFINITION

Q: Is a variant regulatory?

Sequence environment

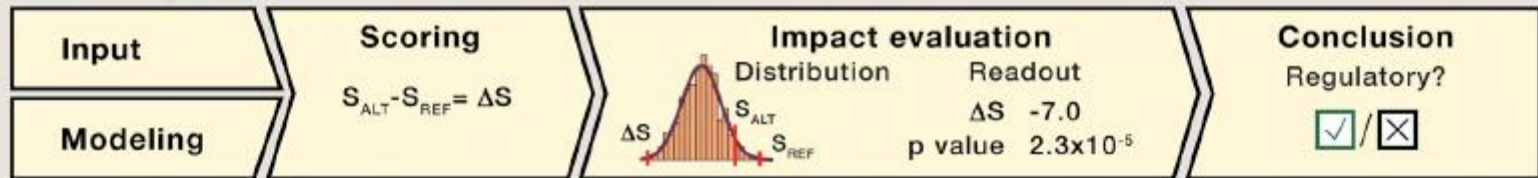
Reference

...TGATTAGGGCTGGTTGGGGCAGGCAGTGACGGGTCGGTTAGGCTAGCAGTCGTAAGAGGAGTGGTGCAGATCGCCAGTGCCAGTCCCTCTTGCTACAGTCGCAGCCATGGGTGAGGTAGTCC...

Variant

...TGATTAGGGCTGGTTGGGGGAGGCAGTGACGGGTCGGTTAGGCTAGCAGTCGTAAGAGGTAGTGGTGCAGATCGCCAGTGCCAGTCCCTCTTGCTACAGTCGCAGCCATGGGTGAGGTAGTCC...

Alternate



Does the variant affect a TF binding site?



Does the variant affect a TF binding site?

Motif modeling

Regular expression

Consensus
AGAGGAAGTG

IUPAC
DVRGGAAVTN

Position-weight matrix

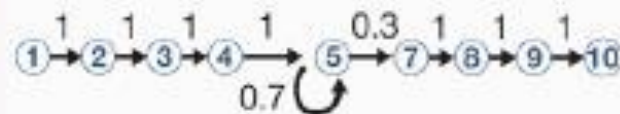


Hidden Markov Model

Linear

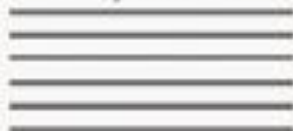


Including spacer



De novo motif discovery

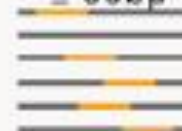
Background sequences



Putative regulatory regions
(ChIP-seq, DHS-seq)
~ 200bp



Random library
(PBM, B1H, HT-SELEX)
≤ 30bp

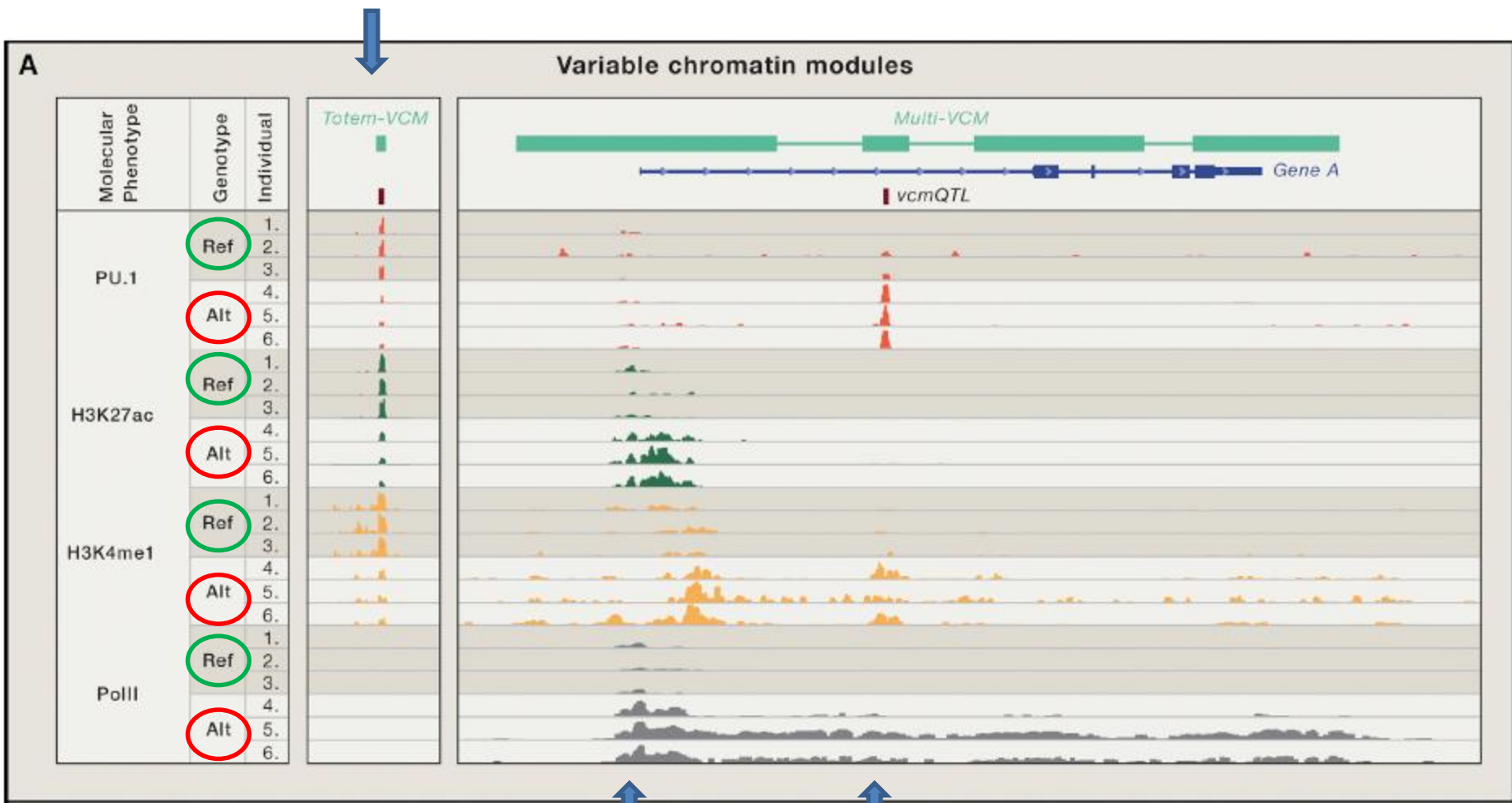


motifs

Sequence over-representation

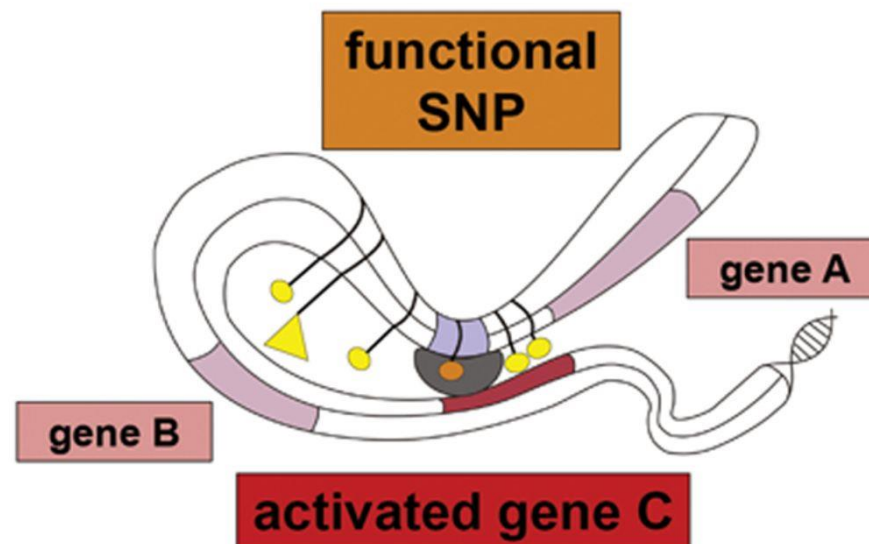
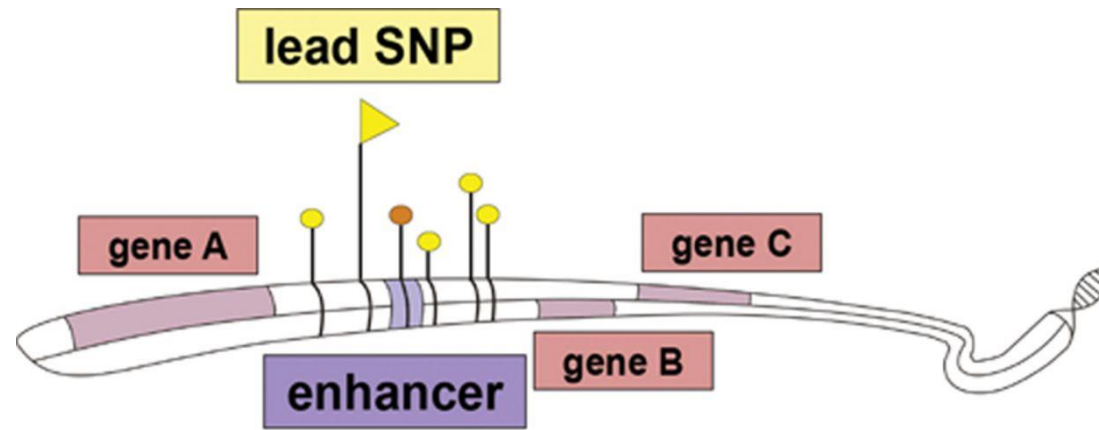


SNPs in the genomic regions may alter a binding site of a specific TFs, such as PU.1 and chromatin states change in the same region

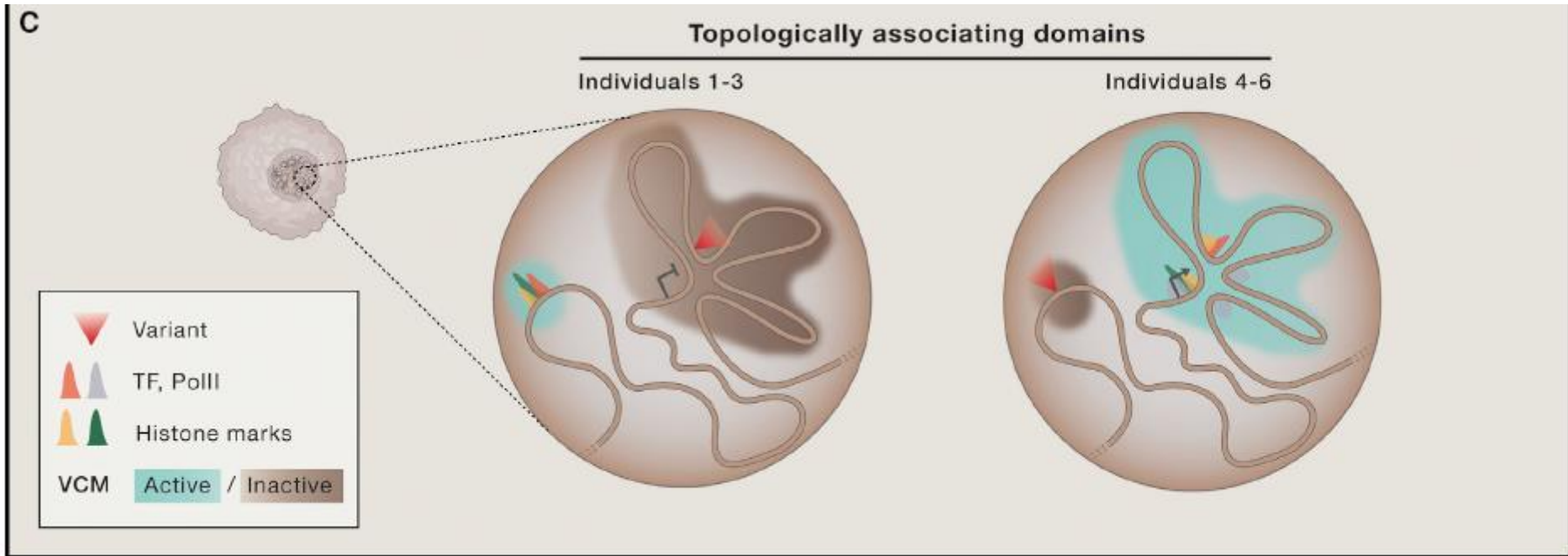


SNPs in the genomic regions may alter a binding site of a specific TFs, such as PU.1 while chromatin states change a whidespread region

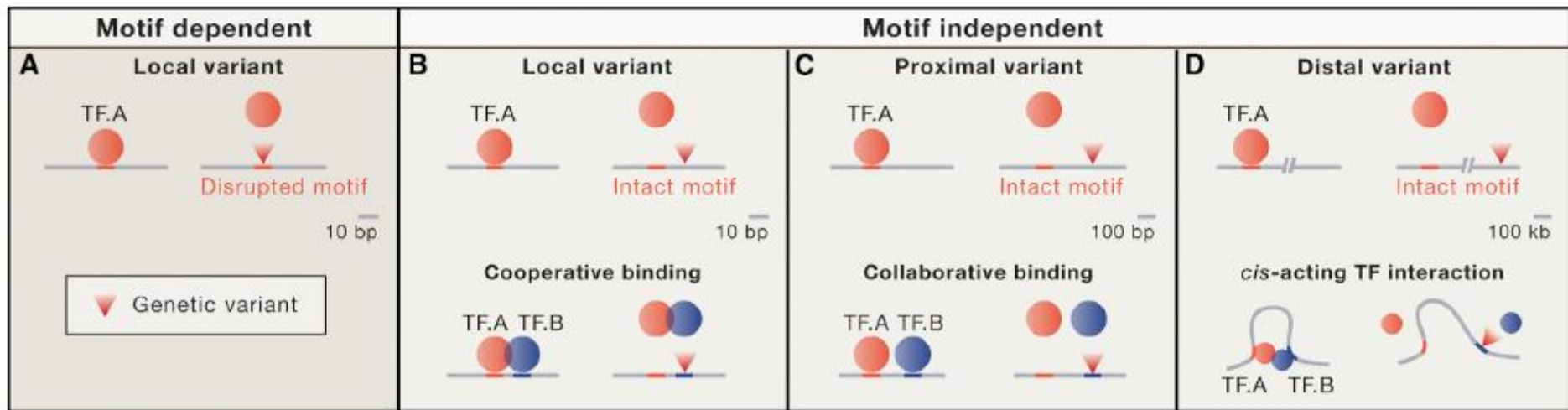
Single nucleotide variants in genomic regulatory regions



SNPs may change long range interactions



SNPs mechanisms for alteration of regulatory transcription factors complexes



SNPs roles in the genomic regulatory regions are:

- Changing the transcription factor binding site
- Disrupting long range interaction
- Inhibition of interaction between transcription factors
- Changing epigenetic modifications



Role of non-coding sequence variants in cancer

Ekta Khurana¹⁻⁴, Yao Fu⁵, Dimple Chakravarty^{2,6}, Francesca Demichelis^{2,3,7}, Mark A. Rubin^{1,2,6} and Mark Gerstein⁸⁻¹⁰

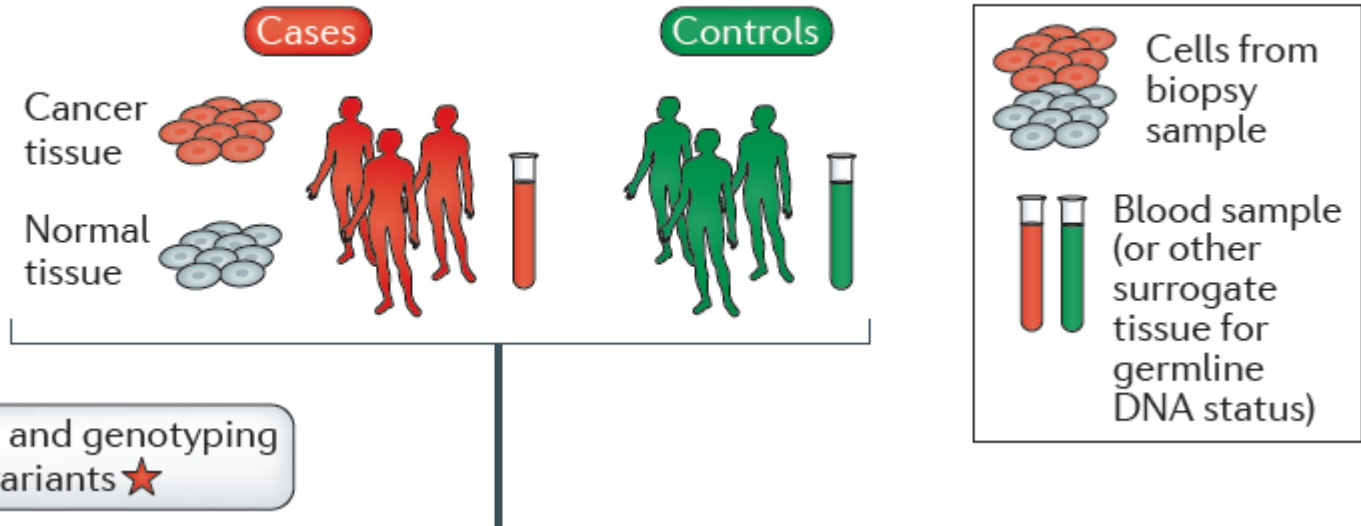
Abstract | Patients with cancer carry somatic sequence variants in their tumour in addition to the germline variants in their inherited genome. Although variants in protein-coding regions have received the most attention, numerous studies have noted the importance of non-coding variants in cancer. Moreover, the overwhelming majority of variants, both somatic and germline, occur in non-coding portions of the genome. We review the current understanding of non-coding variants in cancer, including the great diversity of the mutation types — from single nucleotide variants to large genomic rearrangements — and the wide range of mechanisms by which they affect gene expression to promote tumorigenesis, such as disrupting transcription factor-binding sites or functions of non-coding RNAs. We highlight specific case studies of somatic and germline variants, and discuss how non-coding variants can be interpreted on a large-scale through computational and experimental methods.



SNPs with an impact in tumorigenesis

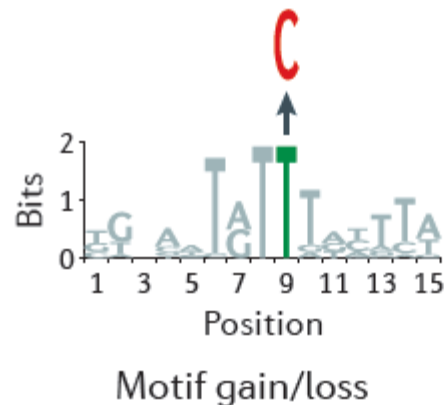
Steps for studying the role of SNP

1



2

Computationally based functional prioritization and interpretation



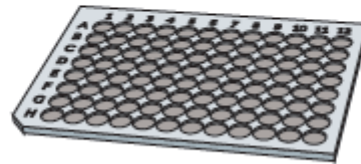
SNPs may have an impact in tumorigenesis

2

FUNCTIONAL ANNOTATION OF SNPs

Experimental validation of functional effects

3



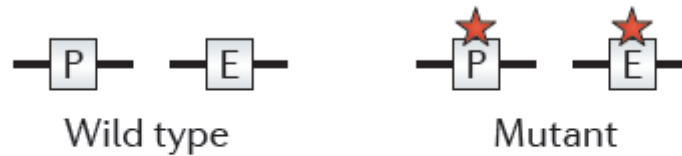
(CRISPR-Cas9,
reporter assays etc.)



SNPs EXPERIMENTAL VALIDATIONS

a Synthesize mutated sequence

- Site-directed mutagenesis
- CRISPR-Cas system
- Oligonucleotide synthesis



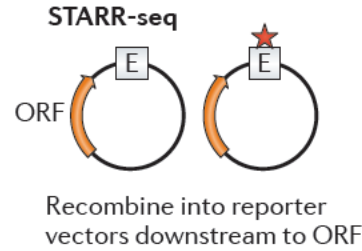
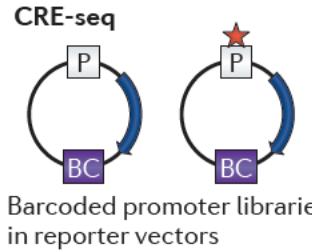
I



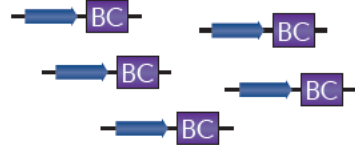
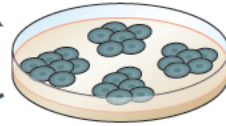
MOLECULAR FUNCTIONAL EFFECTS

b Test molecular functional effects on target gene

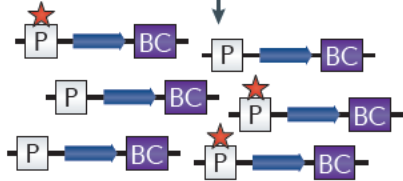
Combined analysis and validation using high-throughput sequencing



Cell lines or model systems



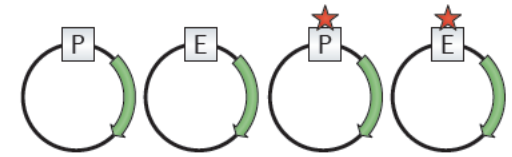
Inference of regulatory element from the transcribed barcode



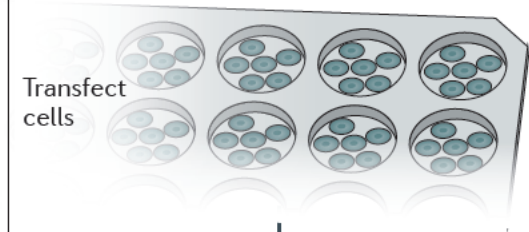
Alignment of reads to the reference genome

High-throughput RNA sequencing to quantify transcription driven by each *cis*-regulatory element

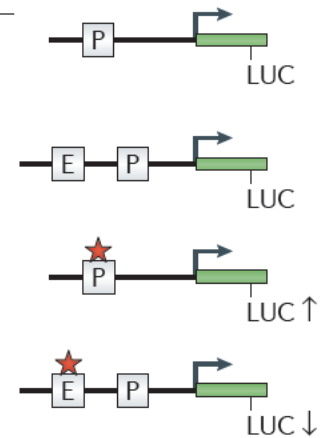
LUC reporter activity



Transfect cells



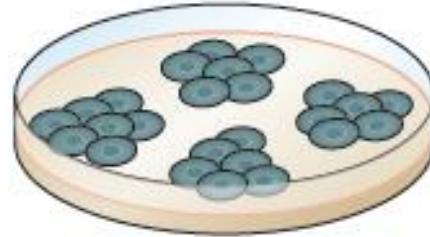
Assay LUC activity



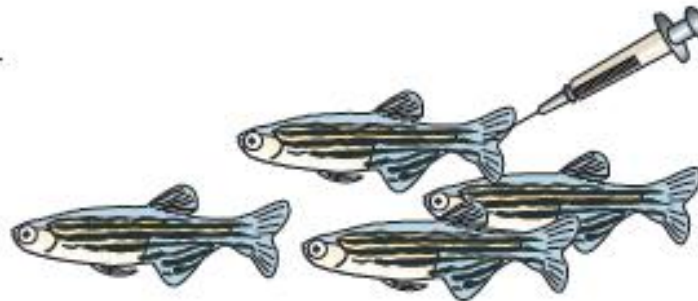
BIOLOGICAL FUNCTION TESTS

c Test effects on oncogenesis

- Proliferation
- Invasion
- Migration



Cell lines



Zebrafish

Tumorigenesis

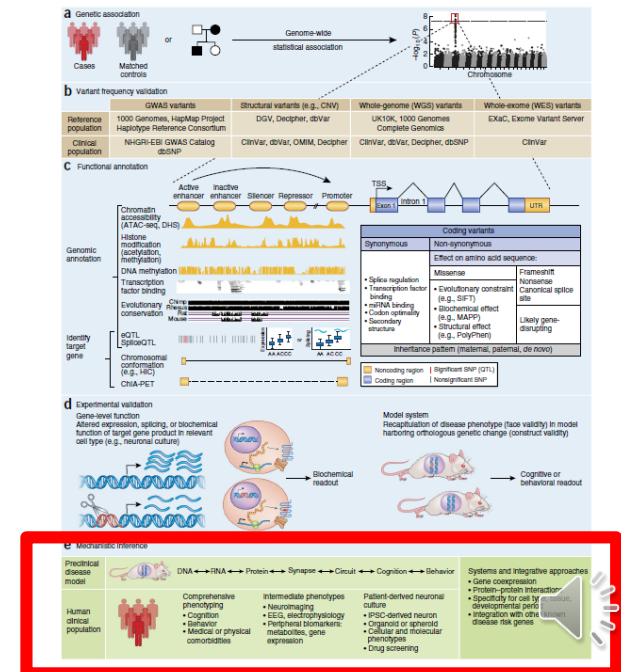


Mouse


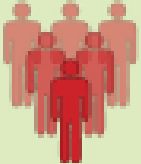


Framework for interpretation of individual disease-associated variants

- Single nucleotide polymorphisms (SNPs) is the nucleotide variations associated with disease
- Genome-wide association studies (GWAS) have successfully identified thousands of common genetic variants associated with complex diseases (<http://www.ebi.ac.uk/gwas/>)
- Functional annotation: to define genomic regulatory regions by genome-wide integration data
- Experimental validation
- Disease Animal models
- **Correlation between molecular mechanisms and disease symptoms**
- **Drug Discovery**



Correlation of SNP/functions with several clinical analysis

e Mechanistic Inference				
Preclinical disease model		DNA ↔ RNA ↔ Protein ↔ Synapse ↔ Circuit ↔ Cognition ↔ Behavior		Systems and Integrative approaches <ul style="list-style-type: none"> • Gene coexpression • Protein-protein Interactions • Specificity for cell type, tissue, developmental period • Integration with other known disease risk genes
Human clinical population		Comprehensive phenotyping <ul style="list-style-type: none"> • Cognition • Behavior • Medical or physical comorbidities 	Intermediate phenotypes <ul style="list-style-type: none"> • Neuroimaging • EEG, electrophysiology • Peripheral biomarkers: metabolites, gene expression 	

How can we use these knowledge?

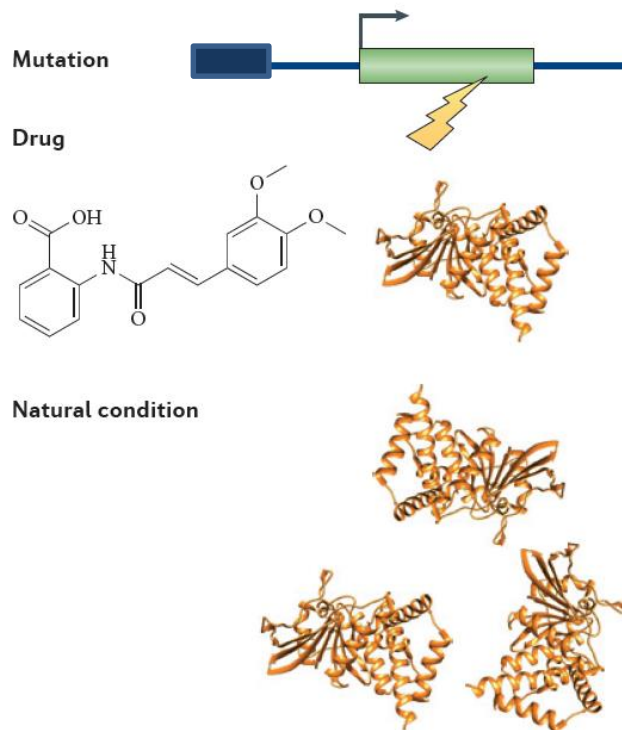


EXAMPLE

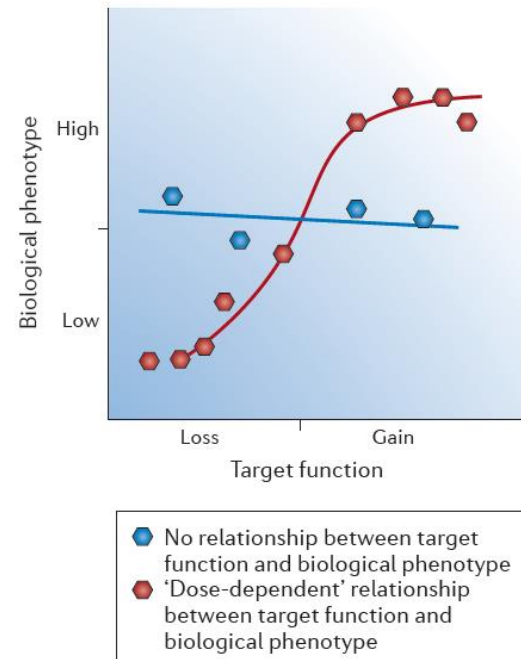
**Gene expression alteration in disease
May be used as BIOMARKERS
(molecules acting as sensor
of disease)**

**Gene expression alteration in disease
May be used as DRUG TARGET
(drug discovery to stop disease and
restore health)**

a Target modulation



b Function-phenotype



c Clinical outcome

Symptoms



Healthy



In Summary:

- **Functional genomics is a field of molecular biology based on genome-wide sequencing data.**
- **Genome-wide sequencing data describe genomic regulatory regions that control gene expression**
- **SNPs-involved in gene expression dysregulation can be linked to the disease**
- **Understanding molecular mechanisms of disease outcome opens the way to discovery drug and identify biomarkers**

