## Genome variation and function 2

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# **Learning Outcomes**

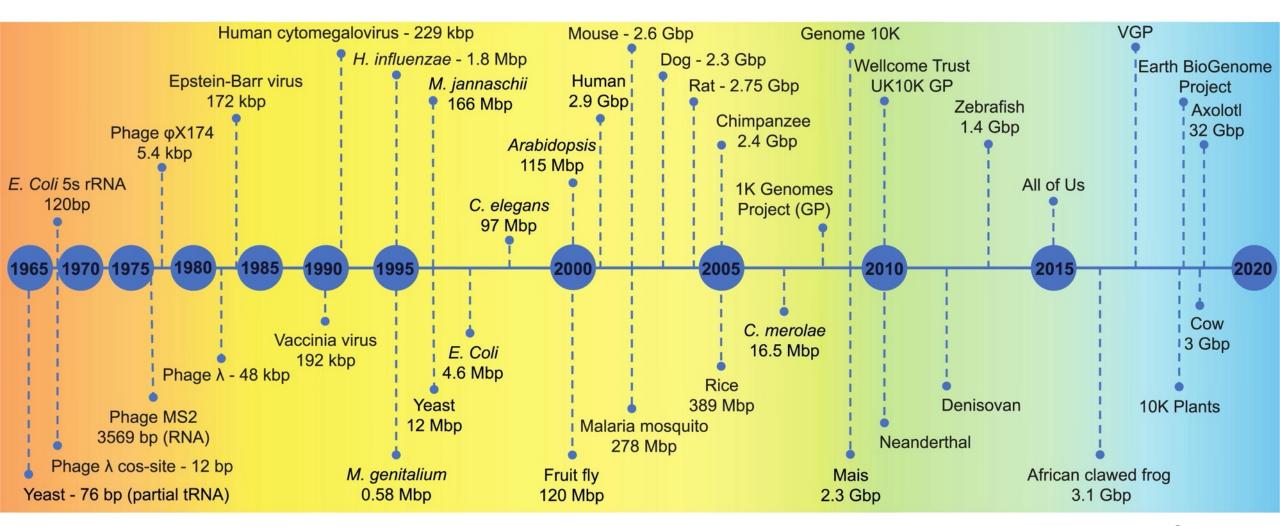
#### **Prologue -** Current topics in genomics

- Human genome sequence
- Nobel prize 2022

#### We will learn about:

- Methods and significance of functional genomics
- Challenges of human genomics

# Brief Review: The timeline of genomics



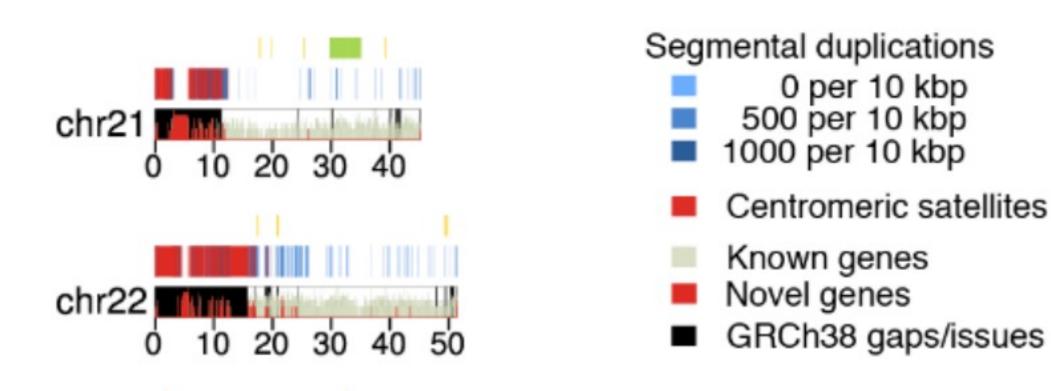
# The complete human genome (2022)

https://www.biorxiv.org/content/10.1101/2021.05.26.445798v1.full

- Adds nearly 200 million base pairs to the 2013 version of the human genome sequence.
- Lasers to scan long stretches of DNA isolated from cells
  - up to 20,000 base pairs at a time. (PacBio)
- Insight into traits/disease/evolution with further studies



# Adds nearly 200 million base pairs to the 2013 version of the human genome sequence.



# The latest technology

https://www.pacb.com/blog/the-evolution-of-dna-sequencing-tools/

#### **First Generation**

#### Second Generation (Next Generation Sequencing)

#### Third Generation





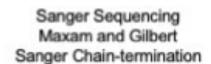








454, Solexa, Ion Torrent Illumina



- Infer nucleotide identity using dNTPs then visualize with electrophoresis
- 500-1000 bp fragments

- High throughput from the parallelization of
- ~50-500 bp fragments

sequencing reactions



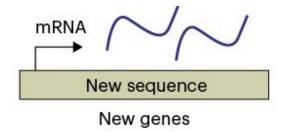
PacBio Oxford Nanopore

- Sequence native DNA in real time with single-molecule resolution
- Tens of kb fragments, on average

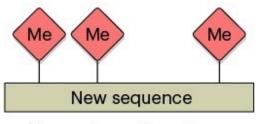
Short-read sequencing

Long-read sequencing

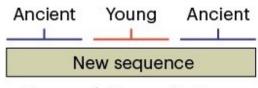
## Evolution/Function/Disease...



 New genes, epigenetic patterns and evolutionary features to be identified in previously unseen regions



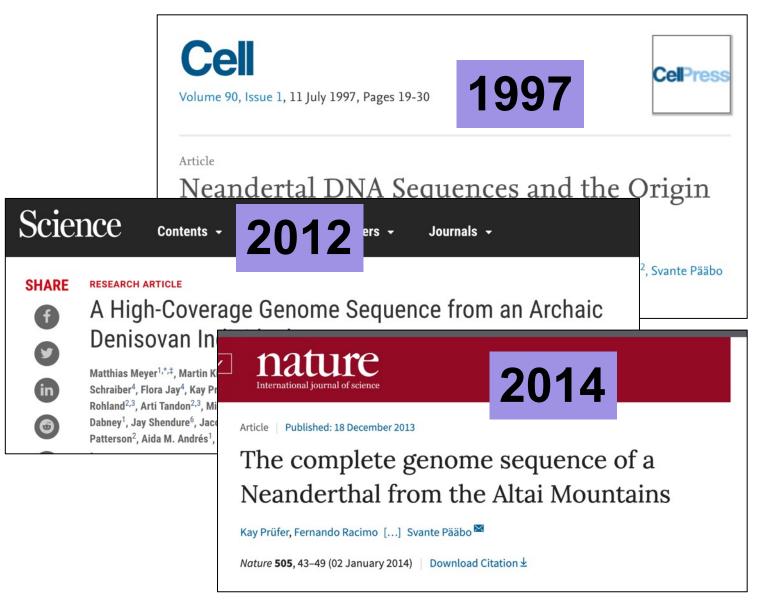
New epigenetic patterns

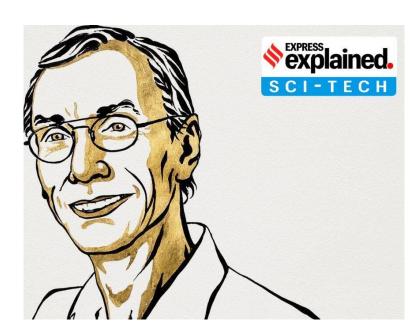


New evolutionary features

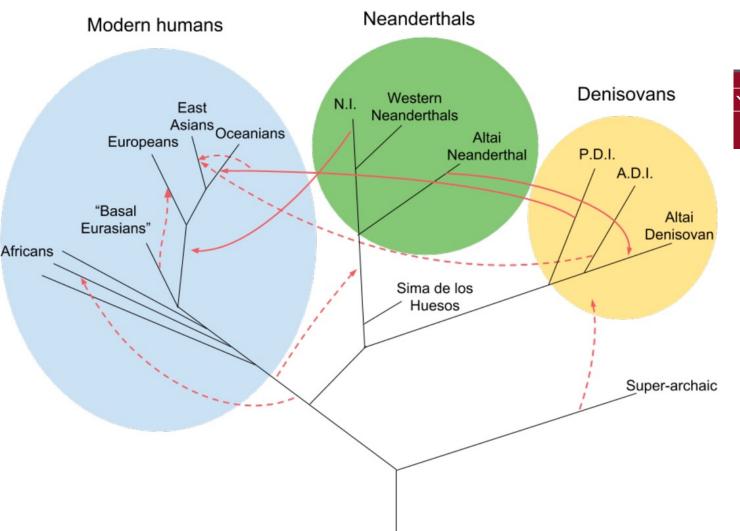
**onature** 

# Ancient human genomics





## **Archaic Hominins**





Letter Published: 22 August 2018

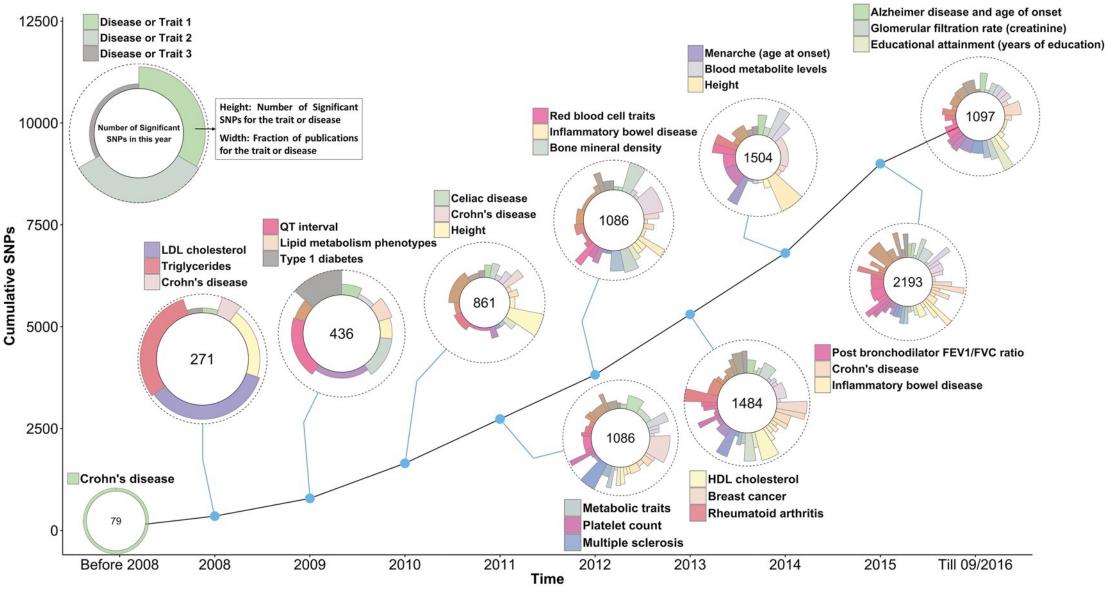
### The genome of the offspring of a Neanderthal mother and a Denisovan father

Viviane Slon , Fabrizio Mafessoni, Benjamin Vernot, Cesare de Filippo, Steffi Grote, Bence Viola, Mateja Hajdinjak, Stéphane Peyrégne, Sarah Nagel, Samantha Brown, Katerina Douka, Tom High Maxim B. Kozlikin, Michael V. Shunkov, Anatoly P. Derevianko, Janet Kelso, Matthias Meyer, Kay Prüfer & Svante Pääbo 

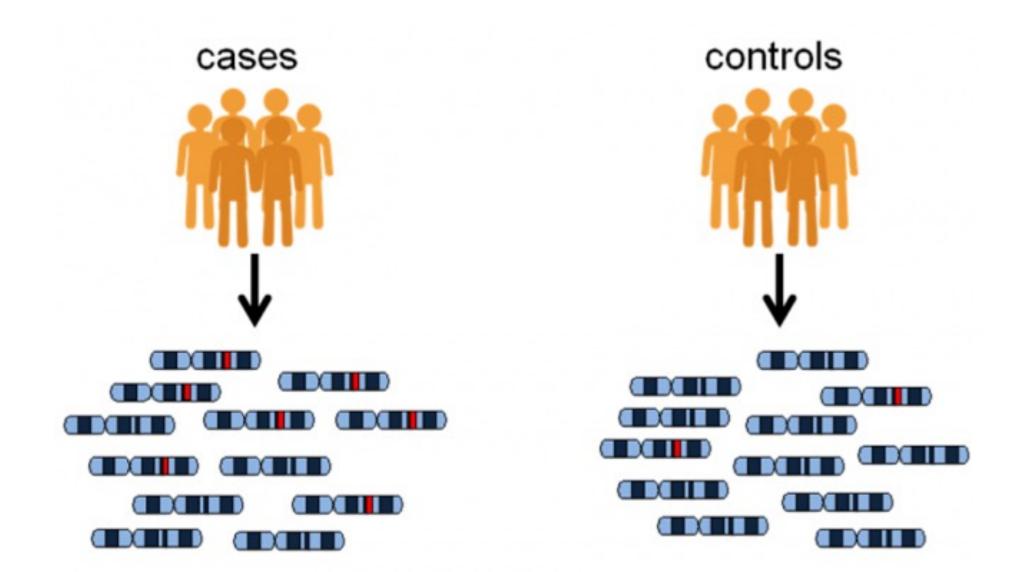
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Nature **561**, 113–116 (2018) | Download Citation ± **27k** Accesses | **45** Citations | **2542** Altmetric | Metrics ≫

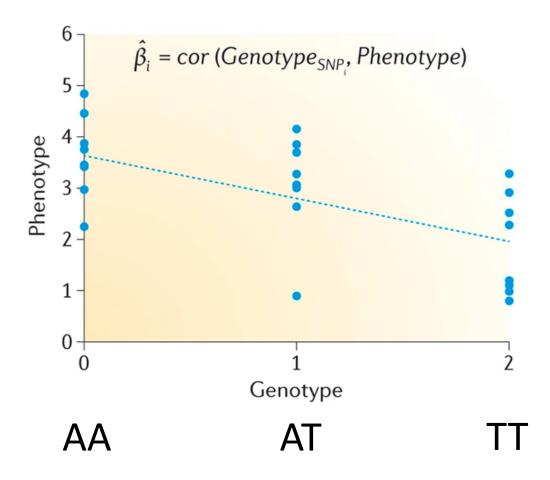
# 10 Years of genome-wide association studies



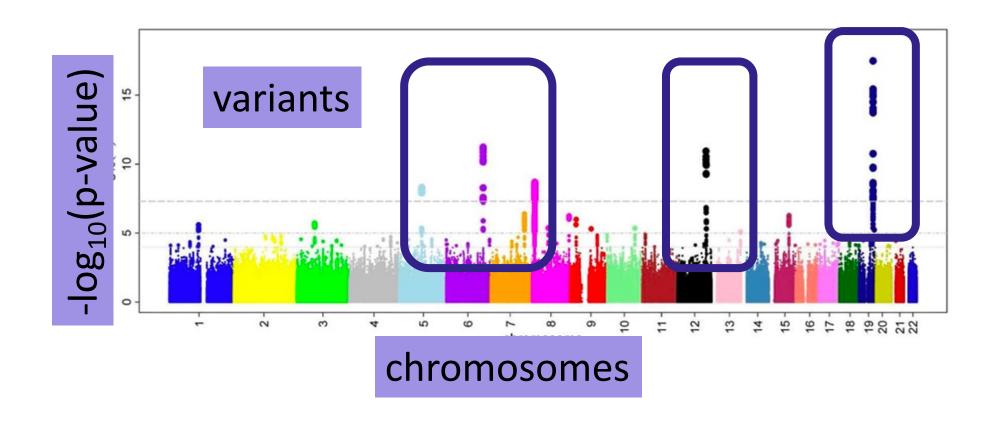
## Genome-wide association study (GWAS)



# GWAS – estimate the association between the phenotype and all observed variants



# How to read a Manhattan plot– a GWAS result representation



# What is the p-value in GWAS?

the probability of obtaining results as extreme as the observed results of a statistical hypothesis test, assuming that the <u>null hypothesis</u> is correct.

- <u>alternative hypothesis</u>: the variant is associated with the trait
- <u>null hypothesis</u>: the variant is not associated with the trait

## Multiple comparisons problem

However, When you conduct the for 100 variants, 5 of the genes can show p < 0.05 under the null hypothesis.

-> When you conduct the for 10,000 variants, 500 of the genes can show p < 0.05 under the null hypothesis (randomly, by chance).

There are various ways to "correct" this problem, considering how many times one conducts the same comparison.

#### Example:

Bonferroni correction - the local significance level is divided by the number of tests performed.

## Why genome-wide analysis is beneficial

- Most quantitative traits are influenced by many genetic variants each with small effects
- GWAS can identify these individual genetic loci with small effects in a large enough sample
- Polygenic score/breeding value predicts phenotype based on genomic information aggregating the small effect of genetic variants

## Desirable traits: mostly polygenic traits Polygenic traits vs Mendelian traits

**Mendelian traits:** shaped by a single gene alone.

example: Wet or dry earwax, Face freckles

**Polygenic traits:** influenced by multiple (~ thousands) genes.

example: Hair color, height and most of quantitative (continuous)

traits

### Polygenic trait: example (three loci, A, B and C affect color)

aabbcc (light color)

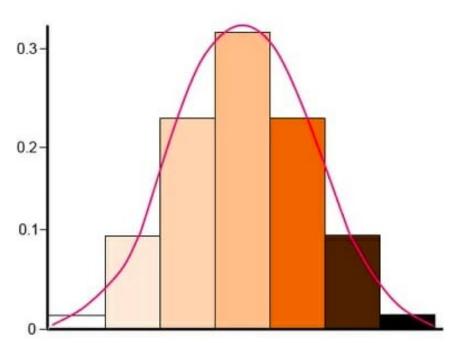
**>** 

AABBCC (dark color)

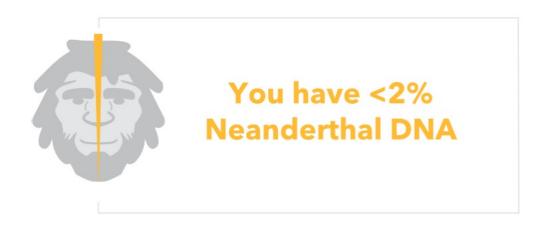
#### AaBbCc (intermediate)

#### genotype and color

|     | ABC | ABc | AbC | aBC | Abc | аВс | abC | abc |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| ABC | 6   | 5   | 5   | 5   | 4   | 4   | 4   | 3   |
| ABc | 5   | 4   | 4   | 4   | 3   | 3   | 3   | 2   |
| AbC | 5   | 4   | 4   | 4   | 3   | 3   | 3   | 2   |
| аВС | 5   | 4   | 4   | 4   | 3   | 3   | 3   | 2   |
| Abc | 4   | 3   | 3   | 3   | 2   | 2   | 2   | 1   |
| аВс | 4   | 3   | 3   | 3   | 2   | 2   | 2   | 1   |
| abC | 4   | 3   | 3   | 3   | 2   | 2   | 2   | 1   |
| abc | 3   | 2   | 2   | 2   | 1   | 1   | 1   | 0   |







| IVIS                                    |        |
|---|--------|
| MARIE SAITO                             | 100%   |
| East Asian & Native American            | 100%   |
| <ul><li>Japanese &amp; Korean</li></ul> | 100%   |
| <ul><li>Japanese</li></ul>              | 100% > |

### **DNA Relatives**

Get started with your predicted relationships, then connect and message to learn more.





#### **Light or Dark Hair**

MARIE, your genetics predict

67% chance of dark brown hair

16% chance of black hair

15% chance of **light brown** hair

2% chance of dark blond hair

<1% chance of **light blond** hair



#### **Eye Color**

MARIE, your genetics predict

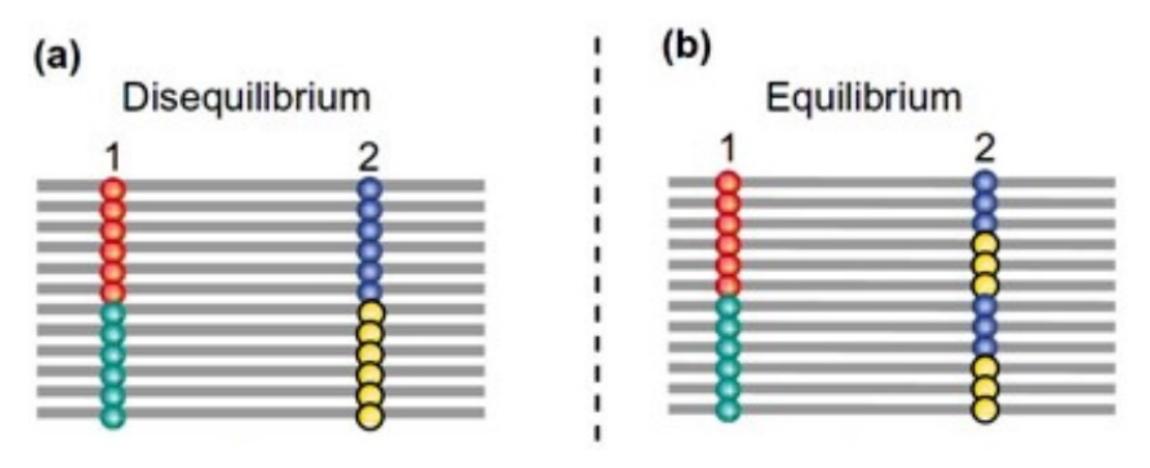
75% chance of dark brown eyes 11% chance of **light brown** eyes 9% chance of dark hazel eyes 4% chance of **light hazel** eyes 1% chance of **green** eyes <1% chance of **blue** eyes <1% chance of greenish blue eyes

## Limitation of genome-wide association studies

- The difficulty in figuring out true causal associations
- They tell us statistical associations, not biological mechanisms

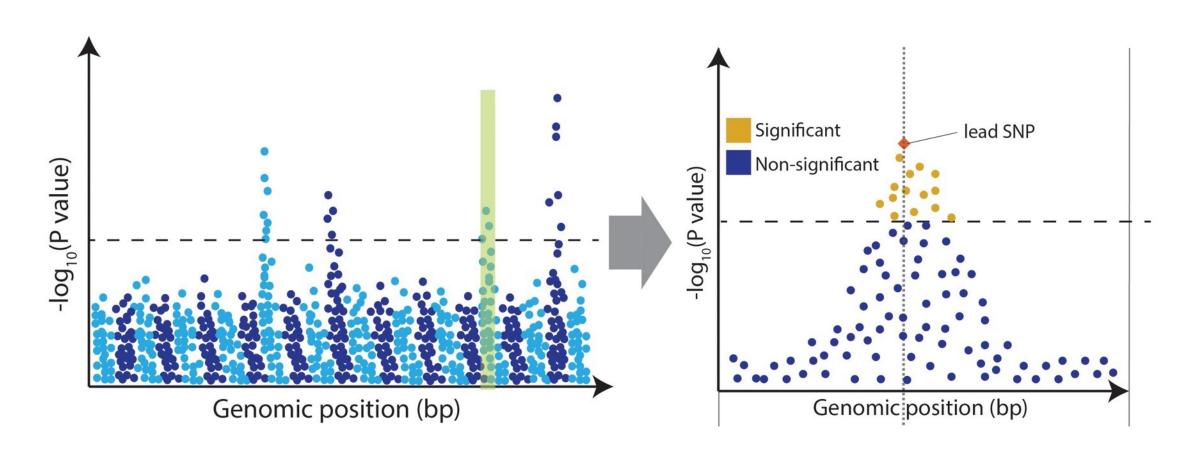
## The difficulty in figuring out true causal associations

- Linkage Disequilibrium

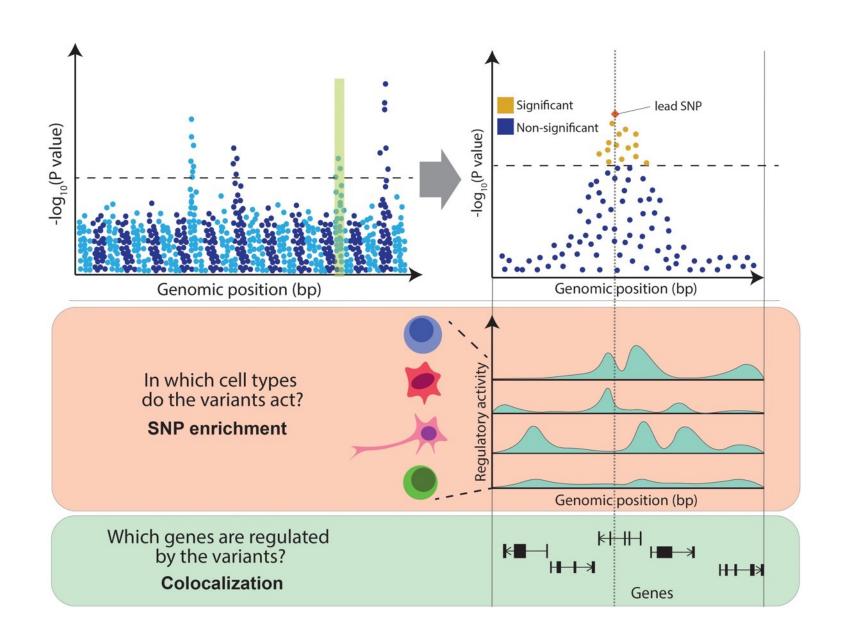


Rafalski. 2002. COPB 5: 94-100

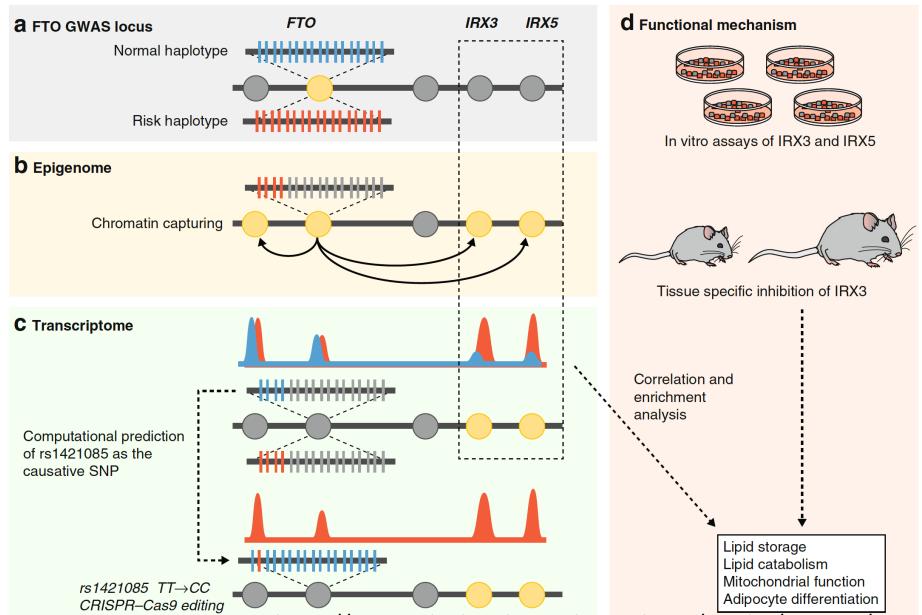
## The difficulty in figuring out true causal associations



## Multi-omics approach to narrow down causal associations



## Multi-omics approach to narrow down causal associations



https://genomebiology.biomedcentral.com/articles/10.1186/s13059-017-1215-1

# Hands-on exercise