

# **GWAS & Polygenic Risk Score**

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<u>Mission</u>: aims to deliver inter-disciplinary research programs and empower the use of data in health research and health care delivery

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### Learning outcomes

- 1. Fundamental understanding of GWAS
- 2. Fundamental understanding of PRS
- 3. Access and interpret GWAS summary statistics.
- 4. Compute PRS and interpret results.
- 5. Understand the pitfalls and limitations of GWAS and PRS studies.



# Outline of this workshop

Part 1. Genome-wide association studies (GWAS)

Part 2. Polygenic risk scores (PRS)

Part 3. Tutorial for PRS of coronary artery disease



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#### Genetic variants

What are genetic variants?

```
... CGATATTCCTATCGAATGTC...
... GCTATAAGGATAGCTTACAG...
... CGATATTCCCCATCGAATGTC...
... GCTATAAGGGTAGCTTACAG...
```



#### Genetic variants

• SNP I WILL TAKE A CAR I WILL TAKE A CAB

• Indels I WILL TAKE A CAR

I WILL TAKE A CARBONARA

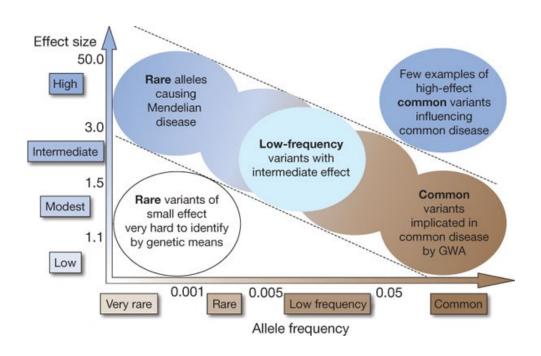
• CNVs I WILL TAKE A CAR

I WILL TAKE A CAR A CAR A CAR



# Contribution of genetic variants

Not all variants affect phenotypes equally



# Mendelian diseases vs complex traits

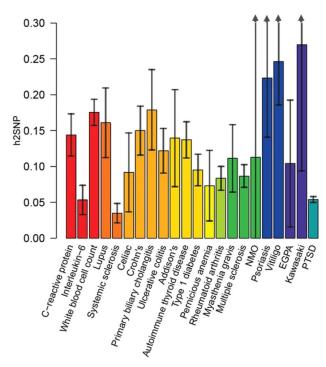
Mendelian traits	Complex traits
<ul> <li>Large effect size</li> <li>Variants are often deleterious</li> <li>Clear inheritance pattern</li> </ul>	<ul> <li>Across many genes</li> <li>Variants are often less deleterious</li> <li>Small effect size</li> <li>No distinct inheritance pattern</li> </ul>

# Mendelian diseases vs complex traits

Mendelian diseases	Complex diseases
<ul> <li>Cystic fibrosis</li> <li>Phenylketonuria</li> <li>Color blindness</li> <li>Hemophilia</li> </ul>	<ul> <li>Obesity</li> <li>Type 2 diabetes</li> <li>Type 1 diabetes</li> <li>Multiple sclerosis</li> <li>Coronary artery disease</li> <li>Schizophrenia</li> </ul>

# What is SNP-based heritability?

- How much phenotypic variance can be explained by SNPs
- High SNP-based heritability: common variants play a big role in the genetic architecture of the trait



# What is SNP-based heritability?

- SNP heritability does not capture variance explained due to:
  - Extremely rare variants
  - Structural variants
  - Epigenetic modifications
  - Gene-gene and gene-environment interactions

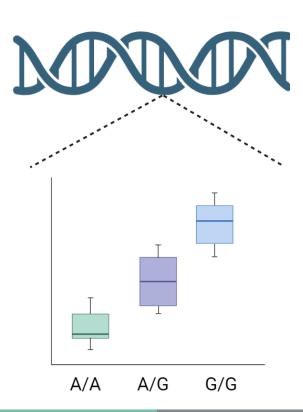
# Capturing variants across the genome

Whole-genome sequencing	Genotyping
<ul> <li>DNA is fragmented, sequenced, and aligned</li> <li>Annotate aligned reads to identify SNPs</li> </ul>	<ul> <li>DNA hybridizes to microarray chips with probes to known SNPs</li> <li>When DNA is hybridized (binds to probe), it generates a fluorescent signal</li> <li>The intensity of the signals is processed</li> <li>Variants are annotated against reference genome</li> </ul>

#### Genome-wide association studies

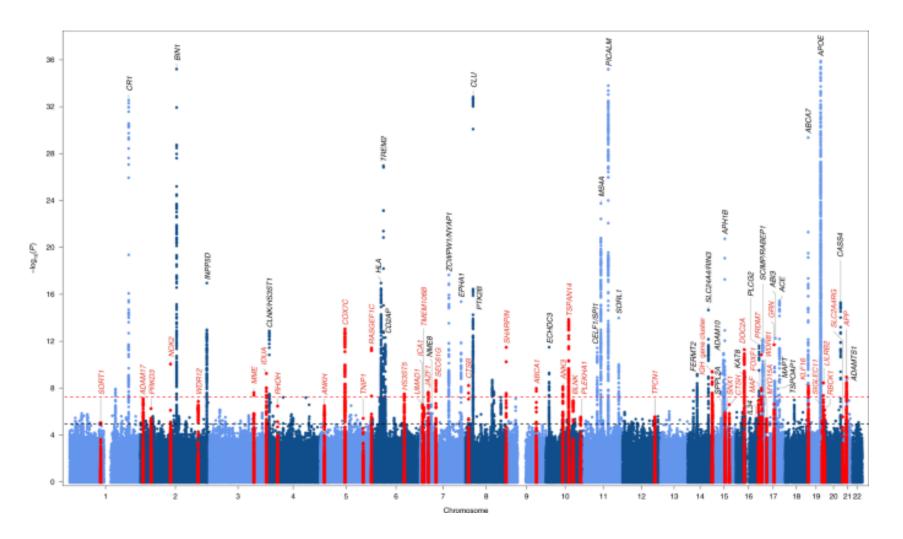
- Goal: test the effect of genetic variants on a trait/disease
- Continuous traits: linear regression
- Binary traits: logistic regression







#### Genome-wide association studies



# Genome-wide significance

- P < 0.05 cannot be used (multiple testing burden)</li>
- Adjust for number of independent LD blocks across the entire genome
- There are roughly 1 million blocks of SNPs (LD blocks) independent of each other
- Apply Bonferroni P-value correction: P = 0.05 / 1,000,000 = 5x10e-8



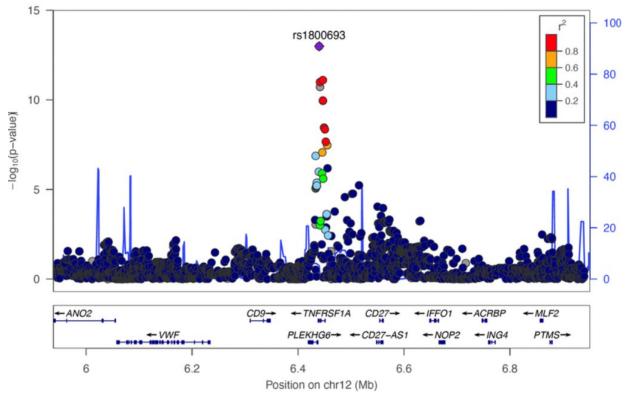
# **GWAS** summary statistics

- Each SNP has an effect size and P-value
- For binary trait: odds ratio = e ^ beta
- Each copy of C allele on rs1800693 increases the odds of multiple sclerosis by about 14%.

Chr	BP (hg19)	Effect Allele	Other Allele	Beta	SE	P-value
12	6440009	С	Т	0.127	0.0166	1.01E-13

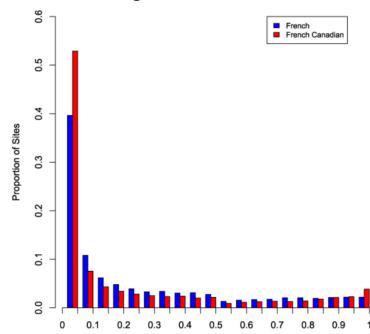
# Pitfalls of GWAS (1)

- Pitfall: raw summary statistics association cannot be used directly
- True risk variant will "tag" other variants through LD

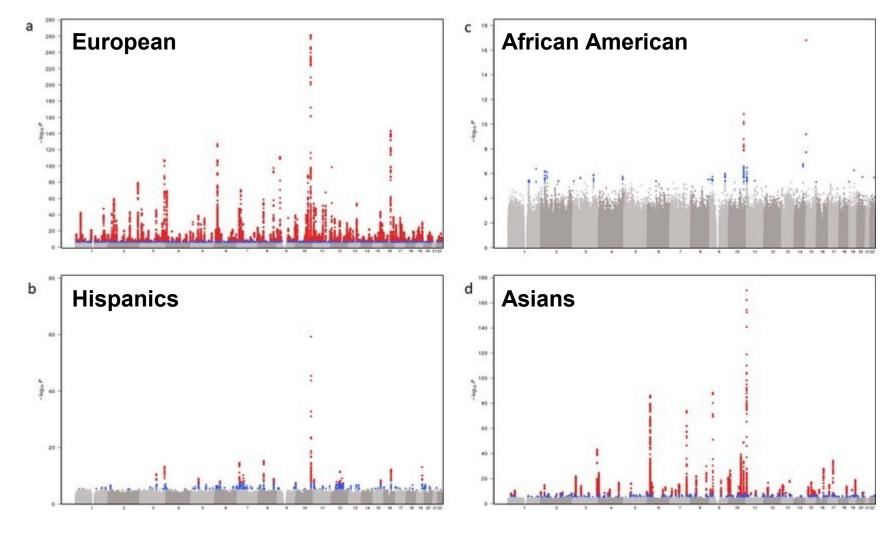


# Pitfalls of GWAS (2)

- Allele frequencies can vary between different ancestries
- Migrations, genetic drift, selection pressures:
  - Differences in allele frequencies
  - Differences in how genetic variants are inherited together (LD)



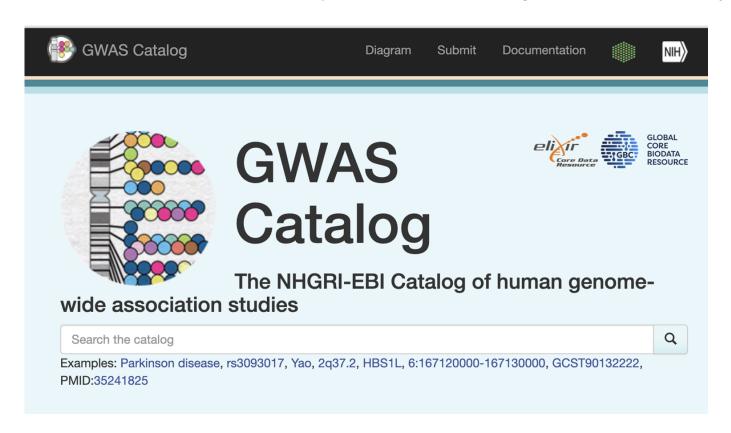
# Pitfalls of GWAS (2)



# Hands-on Activity: GWAS Sumstats

#### Where can I search for GWAS summary statistics?

Visit <a href="https://www.ebi.ac.uk/gwas/">https://www.ebi.ac.uk/gwas/</a> to look through GWAS summary statistics.





# Summary of GWAS

- 1. Genome-wide association studies identify SNPs that are associated to traits.
- 2. LD prevents direct use of GWAS for risk stratification and inference of causal gene
- 3. GWAS performed in European populations may not generalize to other ancestries.
- 4. Open sources (like GWAS catalog) contain summary statistics for GWAS.

# Outline of this workshop

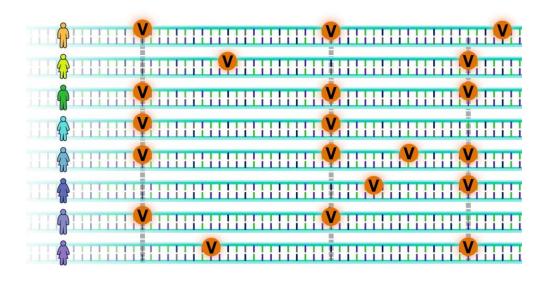
Part 1. Genome-wide association studies (GWAS)

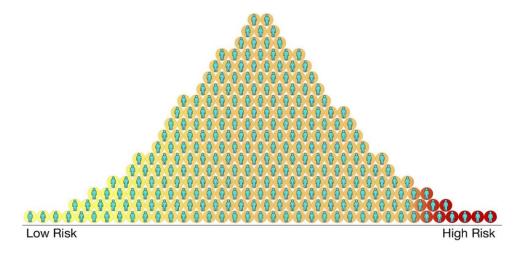
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# Polygenic risk scores

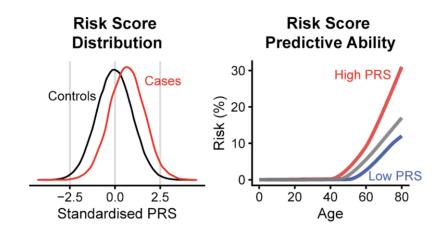






# Polygenic risk scores

#### Goal: early screening, stratification of high-risk individuals





# Polygenic risk scores

Target data
Individual-level
genotype

Process genetic data to keep high-quality common SNPs.

Ancestry/population stratification check.

Training data
GWAS summary
statistics

Samples from GWAS ≠ sample from target data

Matched ancestry with target data.

Well-powered, large sample size.

Linkage disequilibrium (LD) reference panel

Correlation between SNPs.

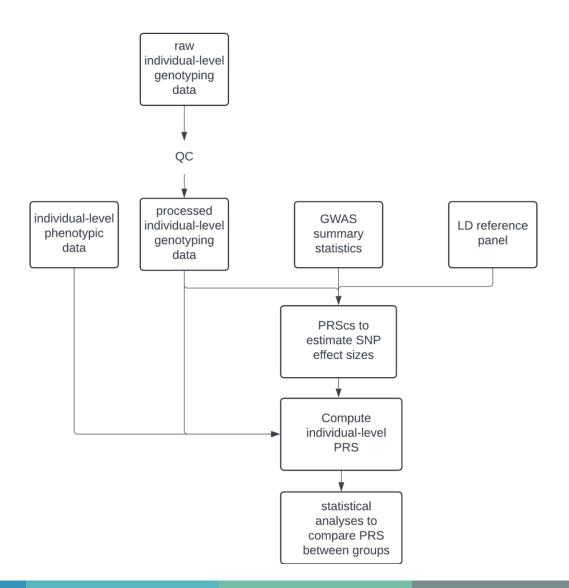
Ancestry of reference panel = ancestry of target data = ancestry of base data



# PRS strategies

Clumping and thresholding	LD prediction	Bayesian estimation of SNP effect sizes
<ul> <li>Ex.: PRSice</li> <li>Uses independent SNPs that are significant at a p- value threshold</li> </ul>	<ul> <li>Ex.: LD pred</li> <li>Predict effect sizes of SNPs based on LD correlation</li> <li>Needs an LD matrix (correlation between all SNPs)</li> </ul>	<ul> <li>Ex.: PRScs</li> <li>LD reference panel to adjust for the correlation between SNPs</li> <li>Adjust SNP effect sizes before computing PRS</li> </ul>

#### PRS flowchart





#### PRS flowchart

- Removes SNPs that are missing in more than a certain proportion of samples
- Remove SNPs that are occur at extremely low frequencies
- Remove SNPs that violate Hardy-Weinberg equilibrium
- Out of all SNPs that are correlated, select only one (LD pruning)
- Remove individuals who have more than a certain portion of missing genotype



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Part 1. Genome-wide association studies (GWAS)

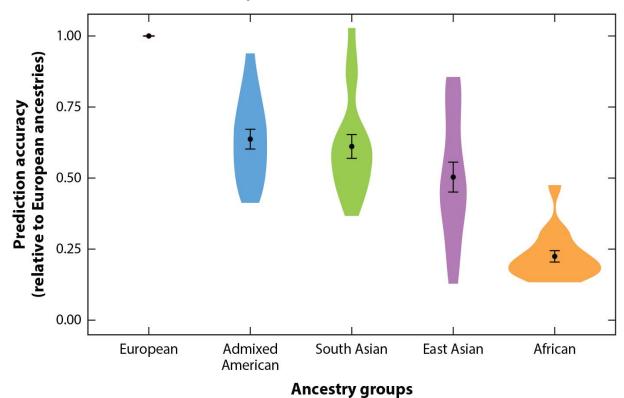
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### Pitfalls of PRS

- Limited clinical utility for non-European populations
- Exacerbate health disparities

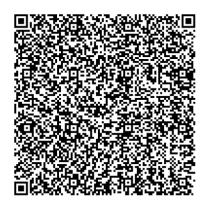


# Summary of PRS

- 1. PRS: estimated genetic susceptibility for a disease
- 2. PRS weights are typically derived from GWAS summary statistics
- 3. PRS trained on one ancestry does not perform well on another

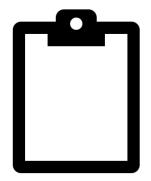
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