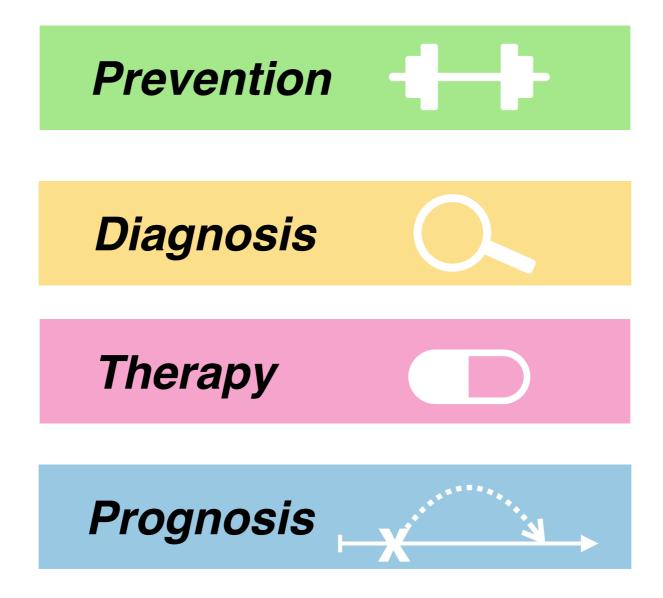
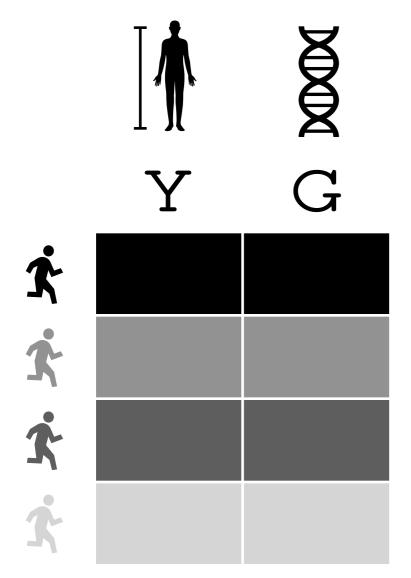
Cornerstones of medicine

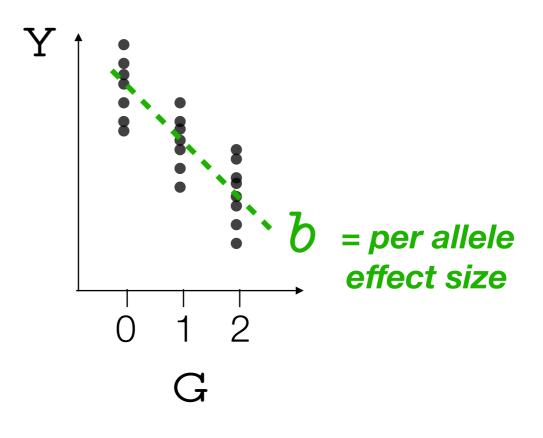


GVVASs

$$Y = Gb + E \sim N(0,s^2)$$

estimate b the effect of a risk allele of genotype G o

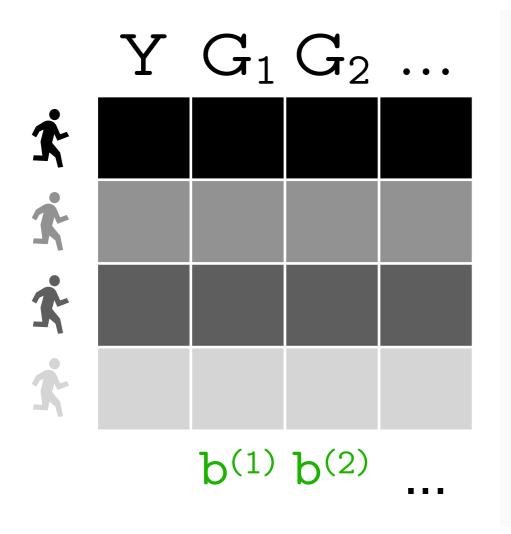


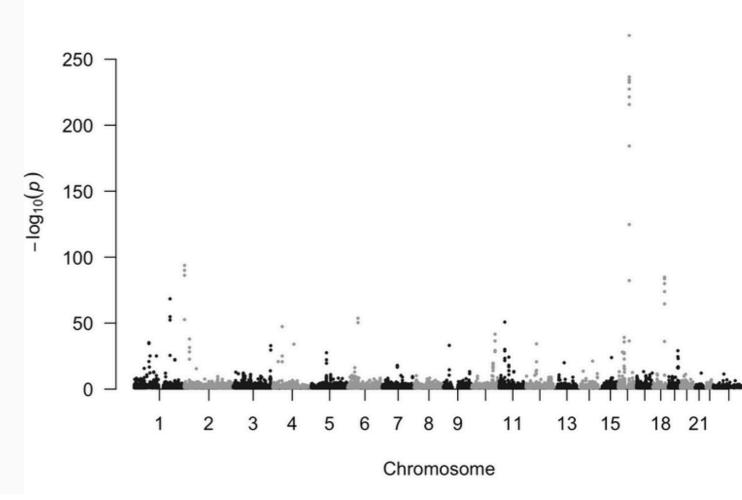


GWASs

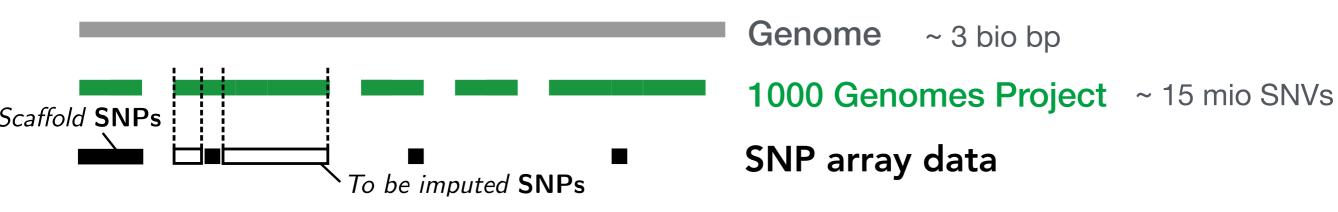
$$Y = G_i b^{(i)} + E \quad E^{-N(0,s^2)}_{i=1, \dots, \text{ \#genotypes}}$$

estimate b(i) the effect of a risk allele of genotype i on Y.





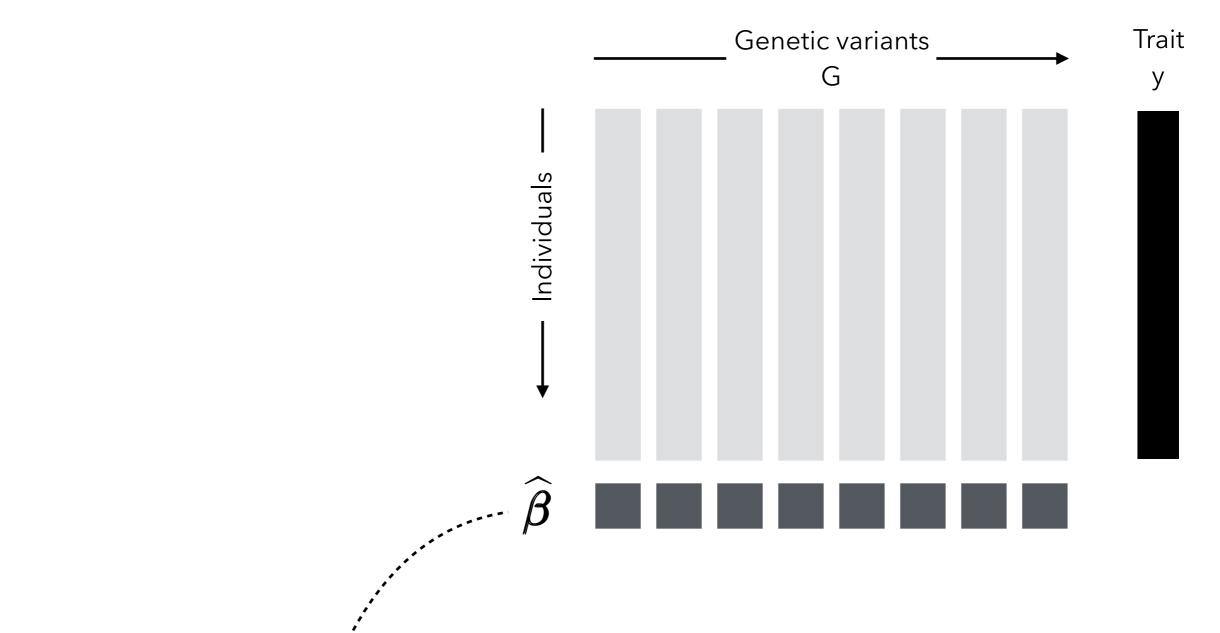
Data



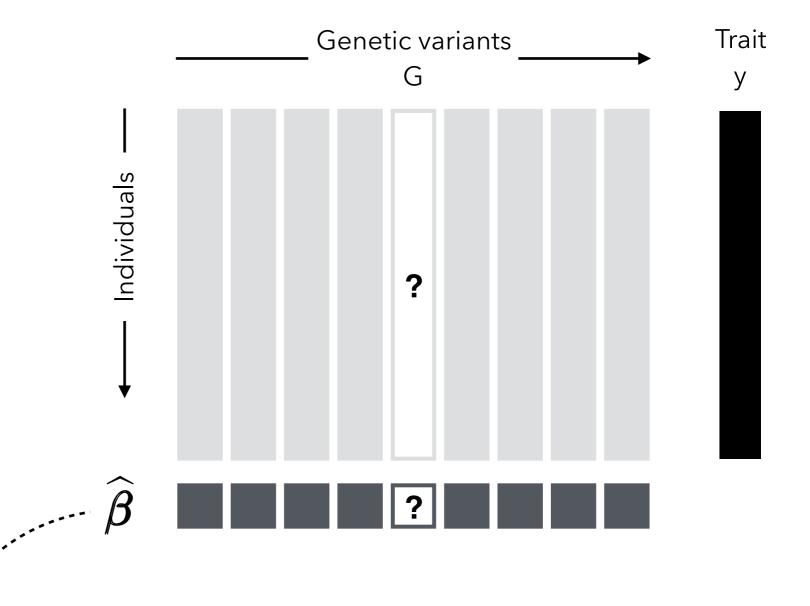


GVVASSummary statistic

SNP	effect size	standard error	sample size	allele frequency
G_1	b ⁽¹⁾			
G_2	b ⁽²⁾			
G_3	b (3)			
- -				



Summary statistics



Summary statistics

of unobserved genotype:

$$\widehat{\beta}_i = ?$$

Trait

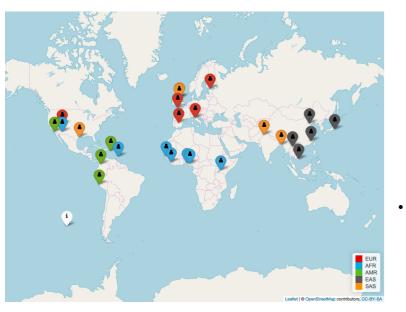
У

Genotype imputation

Step 1

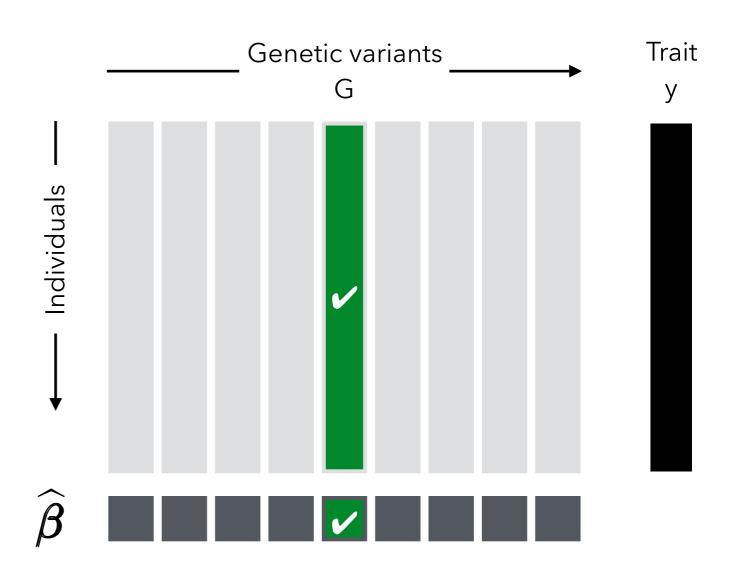
Genetic variants G Individuals Haplotype 1 Haplotype 2 Haplotype 3

1000 Genomes

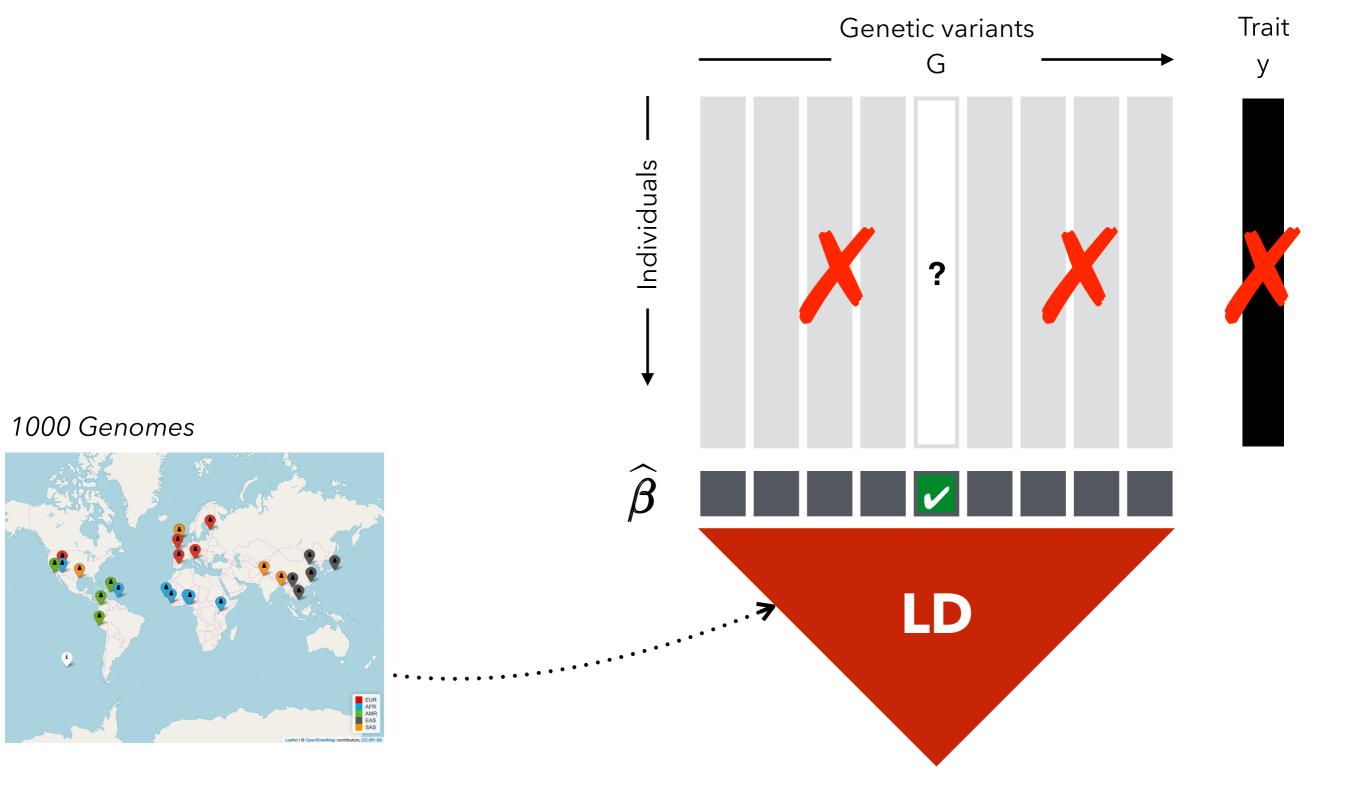


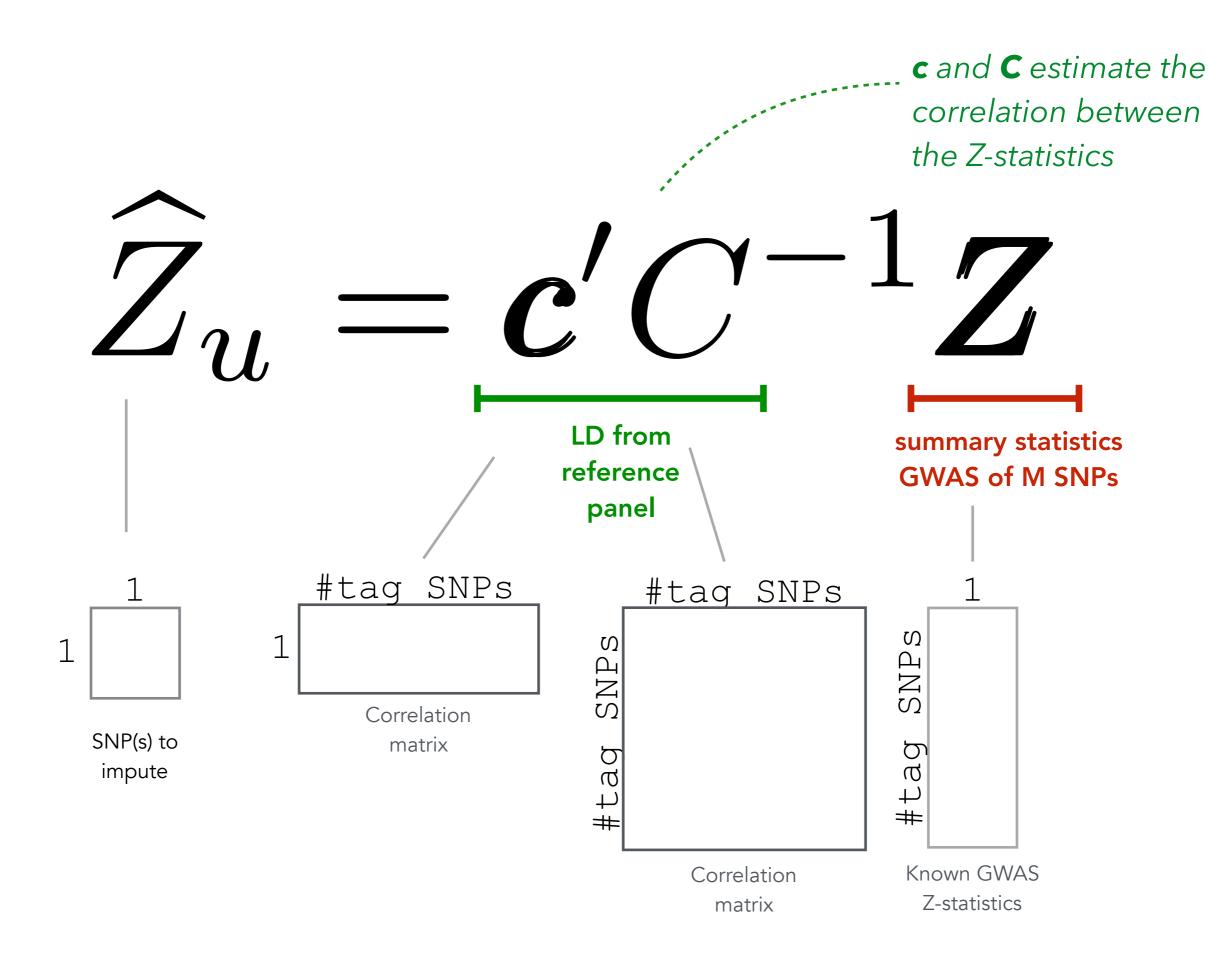
Genotype imputation

Step 2



Summary statistic imputation





SSIMP Methodology

- First described by Wen and Stephens (2010)
- Summary statistic imputation needs:
 - 1. Summary statistics for a set of genotyped markers
 - 2. LD structure
- Advantages compared to genotype imputation:
 - not bounded to privacy restrictions related to the use of individual-level data
 - lower computation time
 - less storage space
 - updating to newer reference panels easier

SSIMP *My contribution*

- Improved the **imputation quality** towards a more accurate, yet fast to compute measure.
- Incorporated **population (ad)mixture** using the weighted LD structure of subpopulations in the reference panel.
- Investigated shrinkage parameter λ .
- Compared summary statistic imputation to genotype imputation,
- Tested the utility of summary statistic imputation on a real case study on human height.

