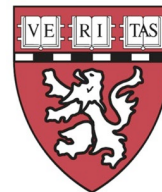


# LD Score: theory

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Analytic and Translational Genetics Unit, MGH

Stanley Center for Psychiatric Research & Program in Medical and  
Population Genetics, Broad Institute



# BGA 2025 – Atlanta!



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## 55th Annual Meeting of the Behavior Genetics Association

25 - 28 June 2025

Atlanta, Georgia, USA

### About the 2025 BGA Meeting

Get ready for BGA in ATLanta! We are looking forward to seeing you and learning about your science. The 55th annualGet ready for BGA in ATLanta! We are looking forward to seeing you and learning about your science. The annual BGA meeting will be held on the Emory University Campus. The 2025 meeting will bring together renowned researchers in the field of behavior genetics, including plenary sessions by local researchers. The theme of this year's meeting is "Harnessing broader perspectives of phenotypic and genotypic risk and liability". The meeting will include both formal presentations by faculty and trainees and multiple networking opportunities. Please join the conversation about the meeting on social media (INCLUDE LINK HERE)

### Local Hosts

[Rohan HC Palmer](#) and [Irwin Waldman](#)

### Program Chair

[Benjamin Neale](#)

### Important Dates (Event)

- 55th Annual Meeting, Atlanta, GA, USA, 25-28 June 2025
- Welcome Reception, Evening of 25 June, Location TBD
- Closing Reception, Evening of 28 June, Location TBD

[Here](#) is the abstract submission



# Francis Galton

## Twin and family studies



RATE OF REGRESSION IN HEREDITARY STATURE.  
Fig. (a)

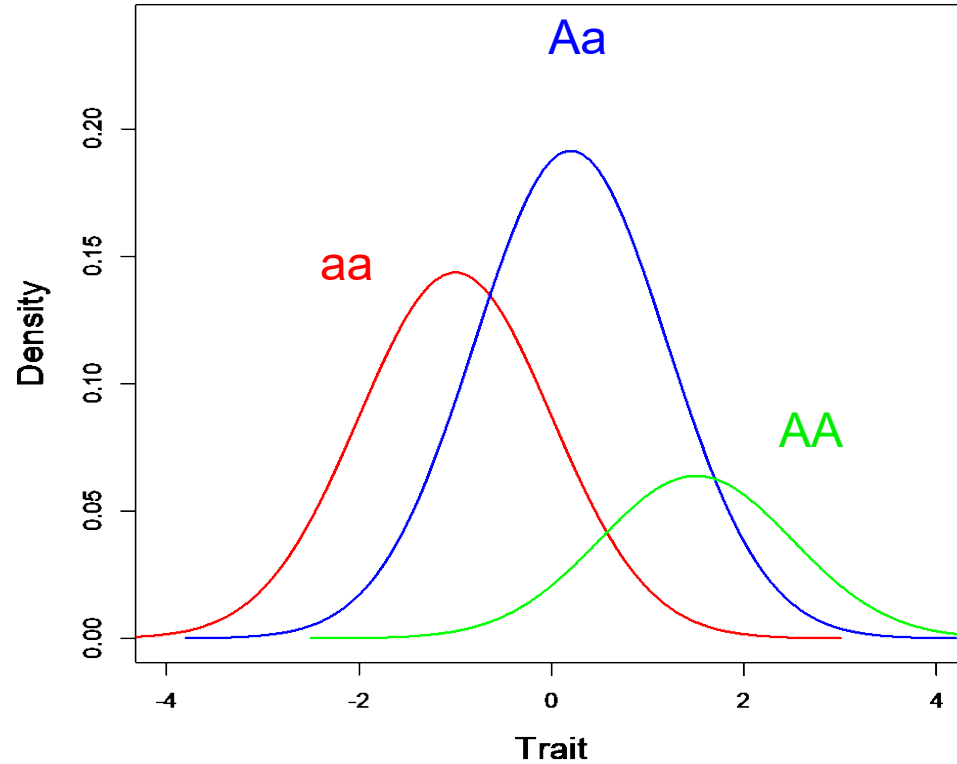
• Relatives are more similar

## Meta-analysis of the heritability of human traits based on fifty years of twin studies

Tinca J C Polderman<sup>1,10</sup>, Beben Benyamin<sup>2,10</sup>, Christiaan A de Leeuw<sup>1,3</sup>, Patrick F Sullivan<sup>4-6</sup>,  
Arjen van Bochoven<sup>7</sup>, Peter M Visscher<sup>2,8,11</sup> & Danielle Posthuma<sup>1,9,11</sup>

Average estimate of heritability 49%  
69% of twin studies support a purely additive genetic model

# Source of variation



# How much mean and variance?



## 1. Defining the Mean ( $X$ )

e.g. cholesterol levels in the population

$$\mu = \sum_i x_i f(x_i)$$

Genotypes	AA	Aa	aa
Effect, $x$	$a$	$d$	$-a$
Frequencies, $f(x)$	$p^2$	$2pq$	$q^2$

$$\text{Mean } (X) = a(p^2) + d(2pq) - a(q^2) = a(p-q) + 2pqd$$

# How much mean and variance?



## 2. Contribution of the QTL to the Variance (X)

$$Var = \sum_i (x_i - \mu)^2 f(x_i)$$

Genotypes	AA	Aa	aa
Effect, x	a	d	-a
Frequencies, f(x)	p <sup>2</sup>	2pq	q <sup>2</sup>

$$\begin{aligned} Var(X) &= (a-m)^2 p^2 + (d-m)^2 2pq + (-a-m)^2 q^2 \\ &= V_{QTL} \end{aligned}$$

$$\text{Heritability of X at this locus} = V_{QTL} / V_{\text{Total}}$$

# How much mean and variance?



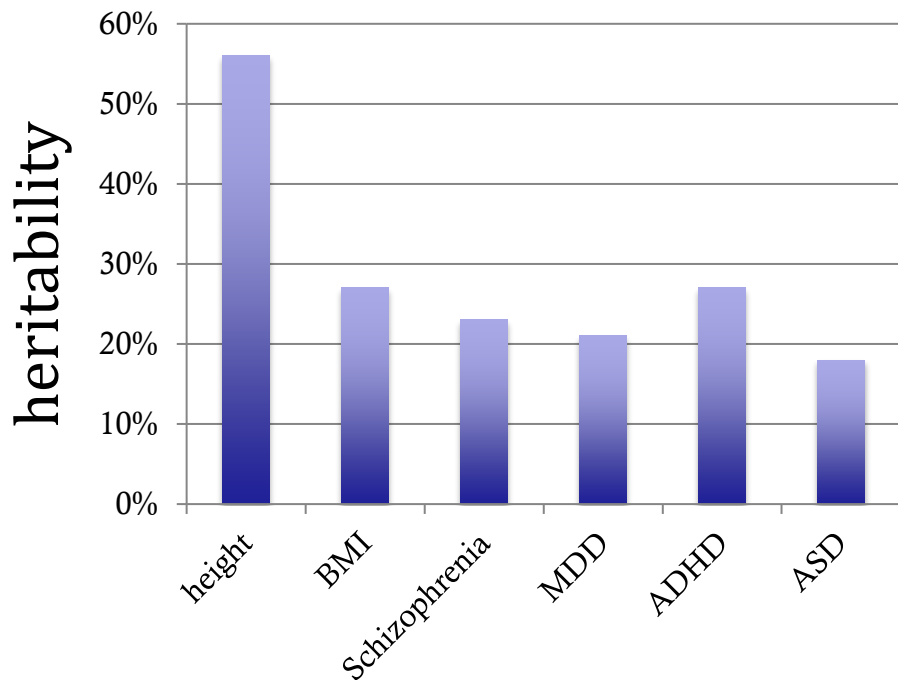
$$\begin{aligned}\text{Var}(X) &= (a-m)^2 p^2 + (d-m)^2 2pq + (-a-m)^2 q^2 \\ &= \frac{2pq[a+(q-p)d]^2}{V_{A_{QTL}}} + \frac{(2pqd)^2}{V_{D_{QTL}}} \\ &= V_{A_{QTL}} + V_{D_{QTL}}\end{aligned}$$

Additive effects: the main effects of individual alleles

Dominance effects: represent the deviation from additive effects

$$m = a(p-q) + 2pqd$$

# GREML/GCTA



- Use estimated genetic similarity

## REPORT

### GCTA: A Tool for Genome-wide Complex Trait Analysis

Jian Yang,<sup>1,\*</sup> S. Hong Lee,<sup>1</sup> Michael E. Goddard,<sup>2,3</sup> and Peter M. Visscher<sup>1</sup>

nature  
genetics

ANALYSIS

Common SNPs explain a large proportion of the heritability for human height

Jian Yang<sup>1</sup>, Beben Benyamin<sup>1</sup>, Brian P McEvoy<sup>1</sup>, Scott Gordon<sup>1</sup>, Anjali K Henders<sup>1</sup>, Dale R Nyholt<sup>1</sup>, Pamela A Madden<sup>2</sup>, Andrew C Heath<sup>2</sup>, Nicholas G Martin<sup>1</sup>, Grant W Montgomery<sup>1</sup>, Michael E Goddard<sup>2</sup> & Peter M Visscher<sup>1</sup>

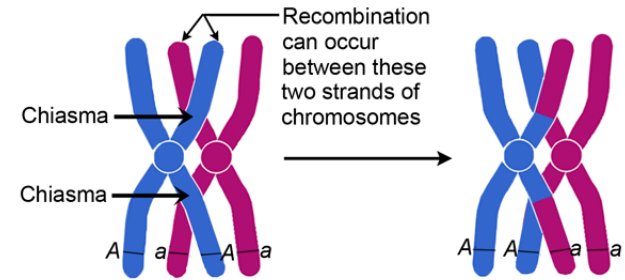
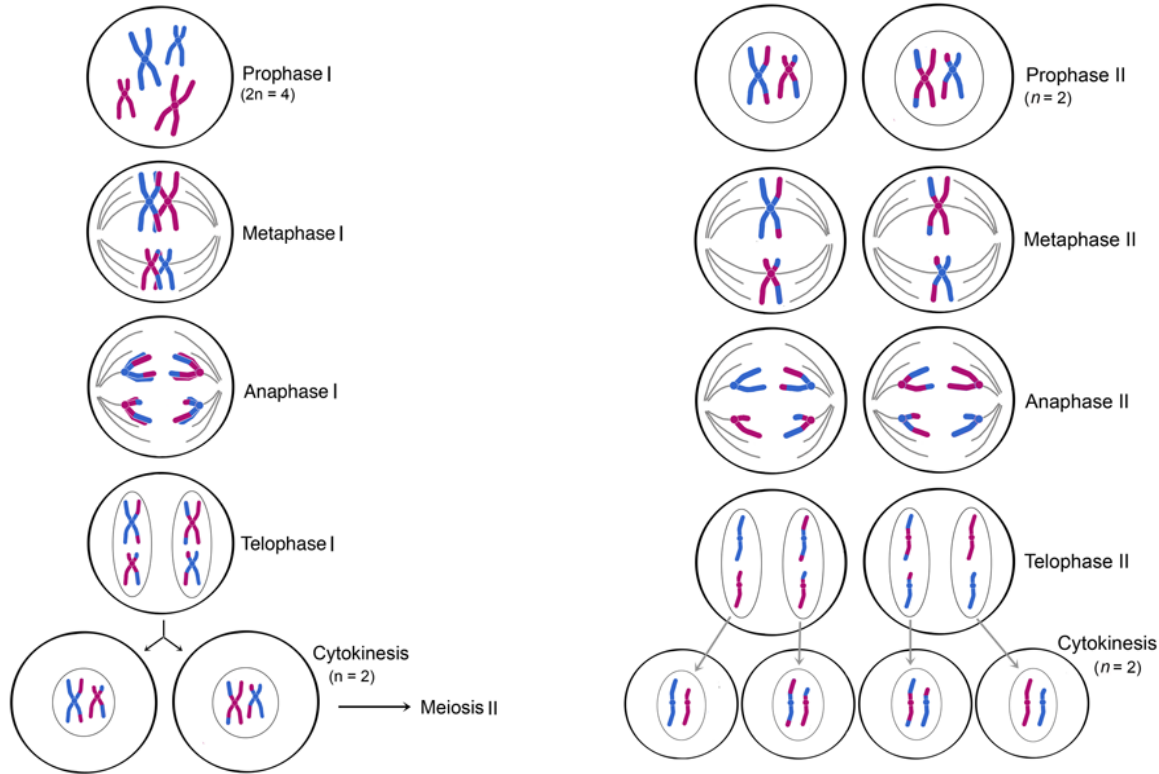
## ARTICLE

Estimating Missing Heritability for Disease from Genome-wide Association Studies

Sang Hong Lee,<sup>1</sup> Naomi R. Wray,<sup>1</sup> Michael E. Goddard,<sup>2,3</sup> and Peter M. Visscher<sup>1,\*</sup>



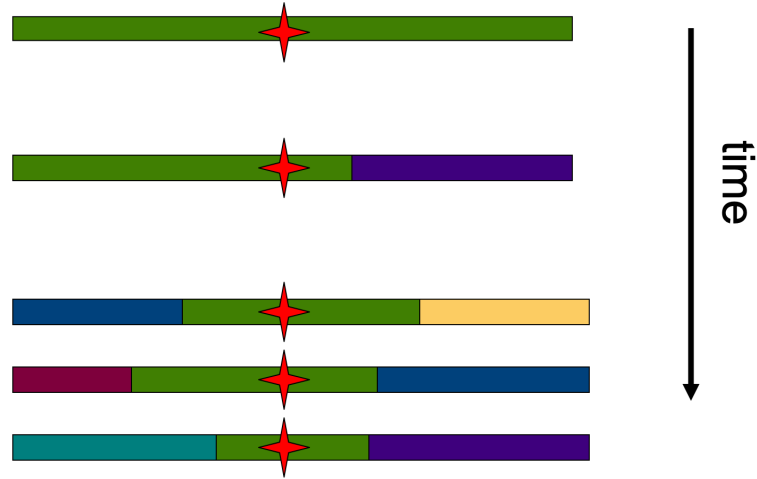
# What happens to our genomes when making sperm and egg cells?



# What happens when a new mutation arises?



# Recombination mixes up the haplotype structure



# Estimators of LD



Two SNPs      A and B  
Genotypes:    AA, Aa, aa  
                  BB, Bb, bb

$$D_{AB} = p_{AB} - p_A p_B$$

$$D' = \frac{D}{D_{\max}}$$

where

$$D_{\max} = \begin{cases} \max\{-p_A p_B, -(1-p_A)(1-p_B)\} & \text{when } D < 0 \\ \min\{p_A(1-p_B), (1-p_A)p_B\} & \text{when } D > 0 \end{cases}$$

$$r^2 = \frac{D^2}{p_A(1-p_A)p_B(1-p_B)}.$$

[nature](#) > [nature reviews genetics](#) > [review articles](#) > [article](#)

Review Article | [Published: June 2008](#)

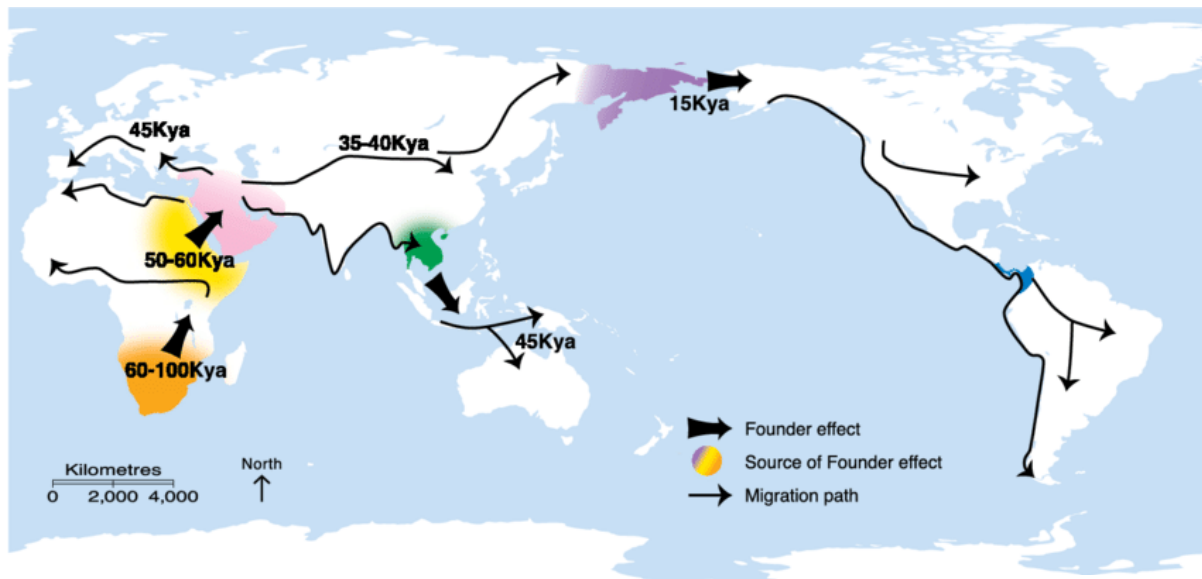
**Linkage disequilibrium – understanding the evolutionary past and mapping the medical future**

[Montgomery Slatkin](#)

[Nature Reviews Genetics](#) 9, 477–485 (2008) | [Cite this article](#)

54k Accesses | 816 Citations | 13 Altmetric | [Metrics](#)

# Genetic ancestry



PERSPECTIVE | BIOLOGICAL SCIENCES |



## The great human expansion

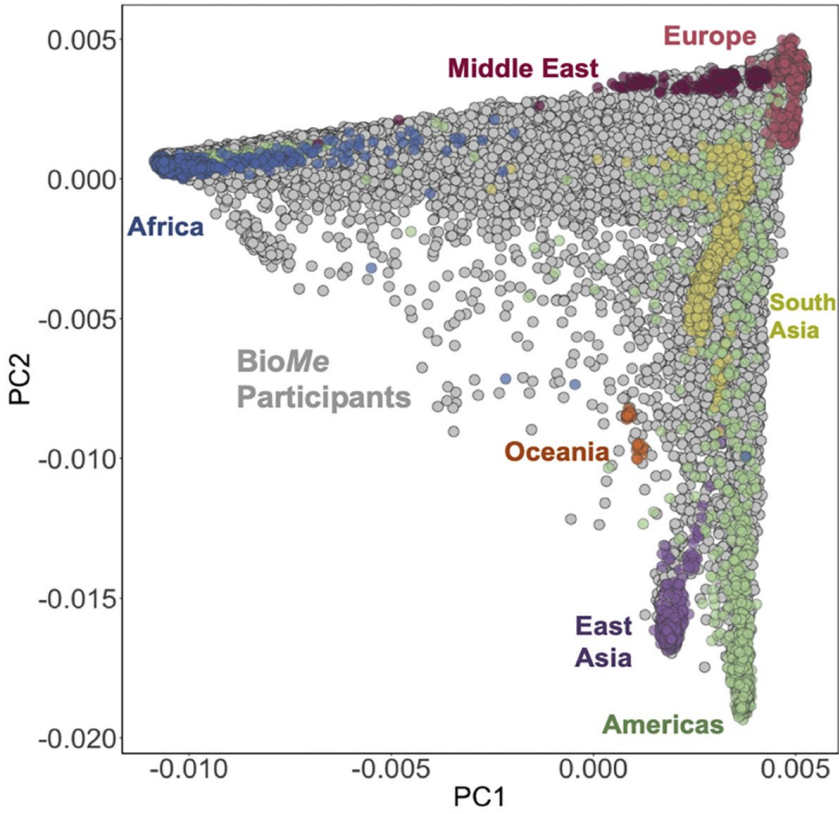
Brenna M. Henn, L. L. Cavalli-Sforza, and Marcus W. Feldman [✉](#) [Authors Info & Affiliations](#)

Edited by C. Owen Lovejoy, Kent State University, Kent, OH, and approved September 25, 2012 (received for review July 19, 2012)

October 17, 2012 | 109 (44) 17758-17764 | <https://doi.org/10.1073/pnas.1212380109>

Henn et al. (2012)

# Principal component analysis on genetic data estimates structure



🔒 | POLICY FORUM | GENETICS AND SOCIETY



## Getting genetic ancestry right for science and society

We must embrace a multidimensional, continuous view of ancestry and move away from continental ancestry categories

ANNA C. F. LEWIS, SANTIAGO J. MOLINA, PAUL S. APPELBAUM, BEGE DAUDA, ANNA DI RIENZO, AGUSTIN FUENTES, STEPHANIE M. FULLERTON, NANIBAA' A. GARRISON,

NAYANIKHA GHOSH, [...] AND DANIELLE S. ALLEN

+10 authors

[Authors Info & Affiliations](#)

# LD Score regression



With thanks



Brendan Bulik-Sullivan



Hilary Finucane



Po-Ru Loh



Mark Daly



Alkes Price

# How does LD shape association?



LD Score regression distinguishes confounding from polygenicity in genome-wide association studies

Brendan K Bulik-Sullivan, Po-Ru Loh, Hilary K Finucane, Stephan Ripke, Jian Yang, Schizophrenia Working Group of the Psychiatric Genomics Consortium, Nick Patterson, Mark J Daly, Alkes L Price & Benjamin M Neale

[Affiliations](#) | [Contributions](#) | [Corresponding author](#)

*Nature Genetics* **47**, 291–295 (2015) | doi:10.1038/ng.3211  
Received 07 March 2014 | Accepted 07 January 2015 | Published online 02 February 2015





# How does LD shape association?

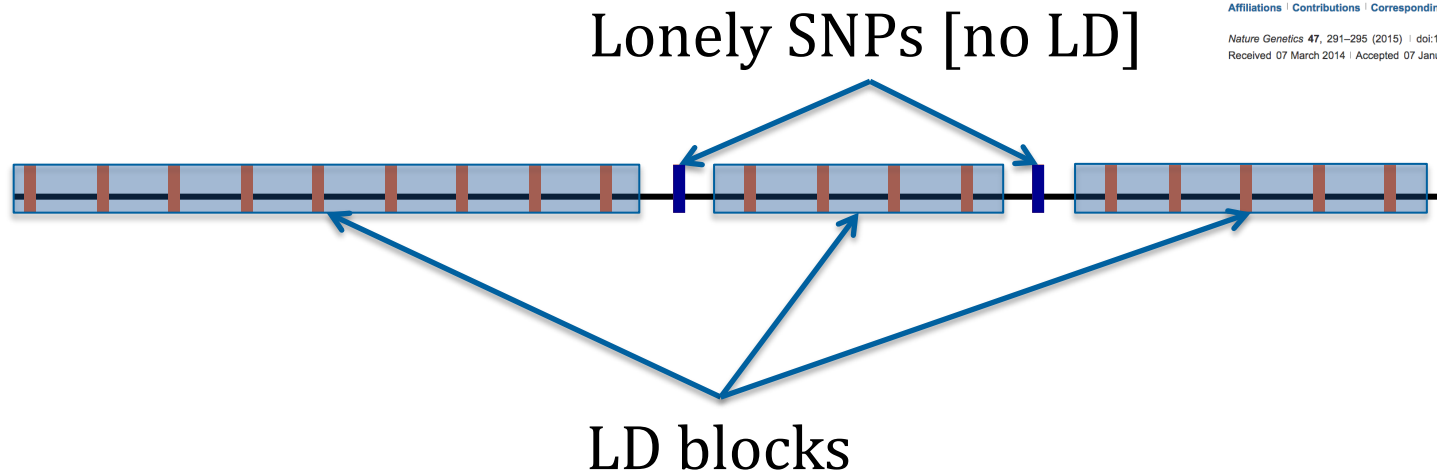


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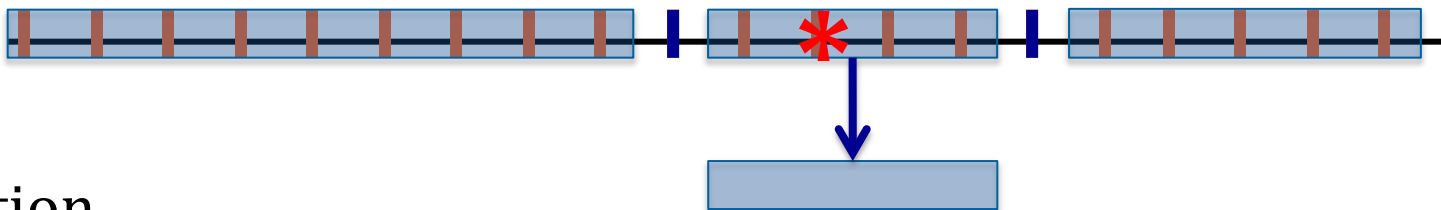
# How does LD shape association?



┃ Lonely SNPs [no LD]

▭ LD blocks

\* Causal variants



Association

All markers correlated with a causal variant show association

LD Score regression distinguishes confounding from polygenicity in genome-wide association studies

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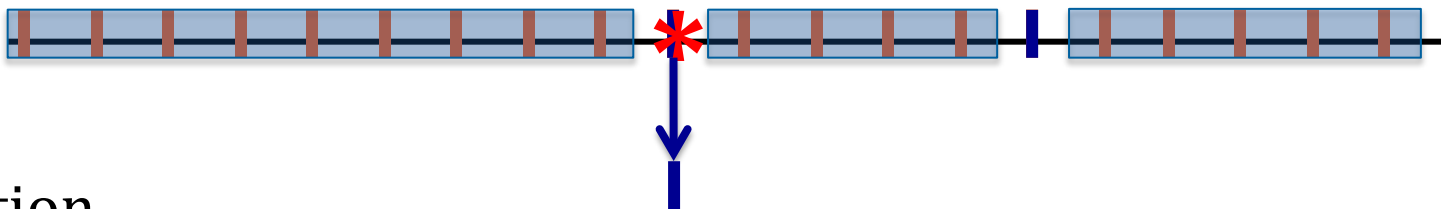
# How does LD shape association?



■ Lonely SNPs [no LD]

■ LD blocks

\* Causal variants



Association

Lonely SNPs only show association if they are causal

LD Score regression distinguishes confounding from polygenicity in genome-wide association studies

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# What happens under polygenicity?



■ Lonely SNPs [no LD]

■ LD blocks

\* Causal variants



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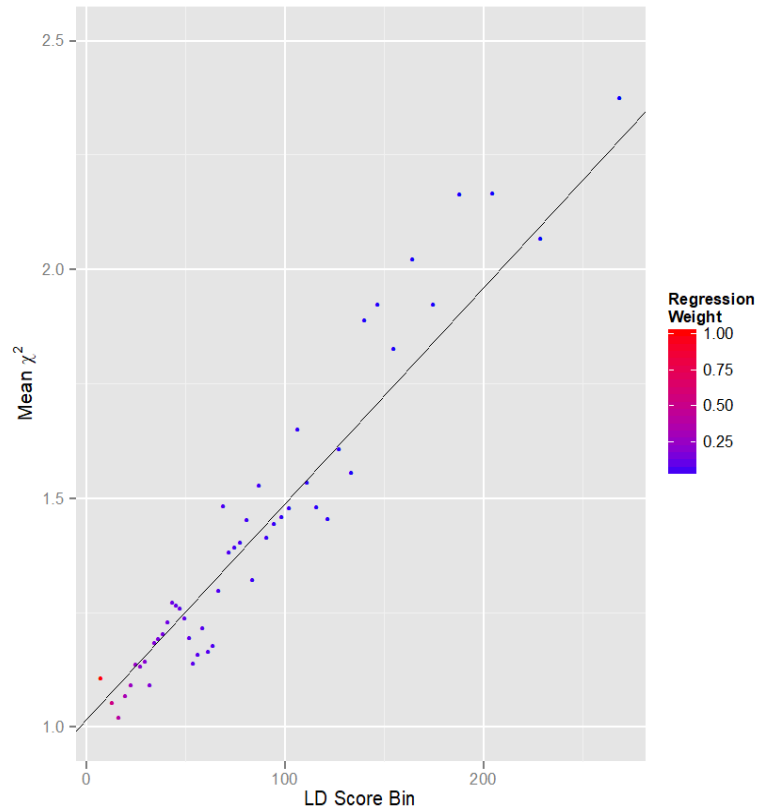
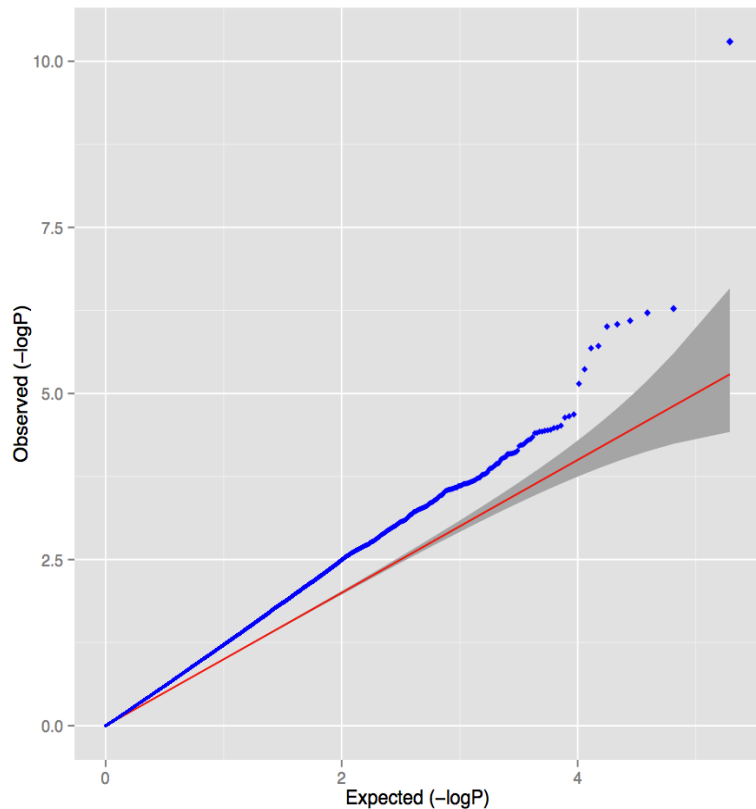
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Assuming a uniform prior, we see SNPs with more LD friends showing more association

The more you tag, the more likely you are to tag a causal variant

# Simulated polygenic architecture

Lambda = 1.30 LD score intercept = 1.02



# What happens under stratification?



■ Lonely SNPs [no LD]

■ LD blocks

\* Causal variants

LD Score regression distinguishes confounding from polygenicity in genome-wide association studies

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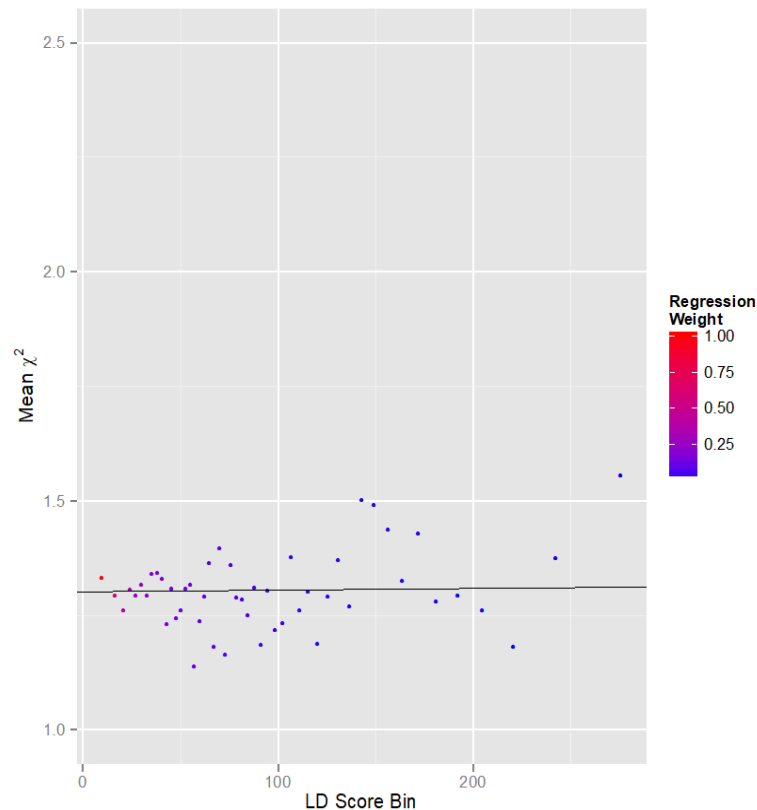
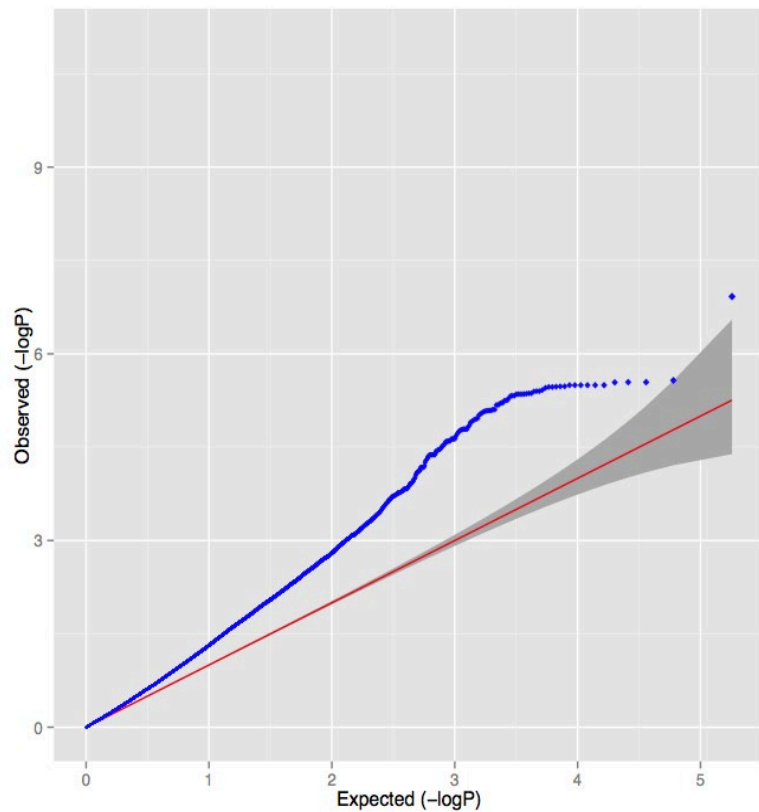
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Under pure drift we expect LD to have no relationship to differences in allele frequencies between populations

# UK controls versus Sweden controls

$\Lambda = 1.30$  LD score intercept = 1.32



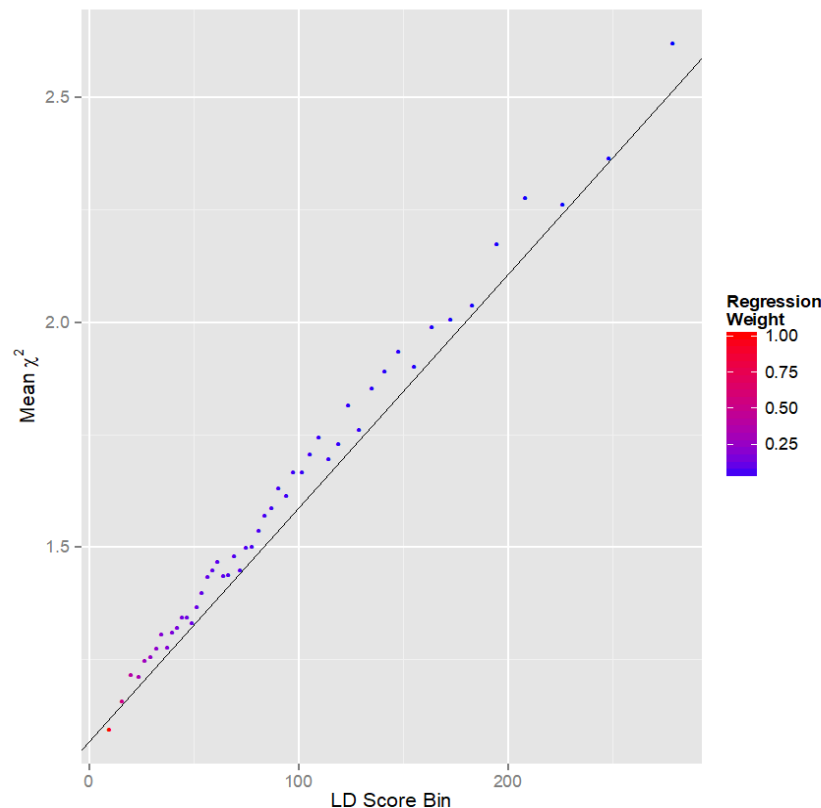
# PGC Schizophrenia

Lambda = 1.48

Intercept = 1.06

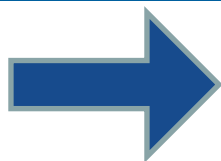
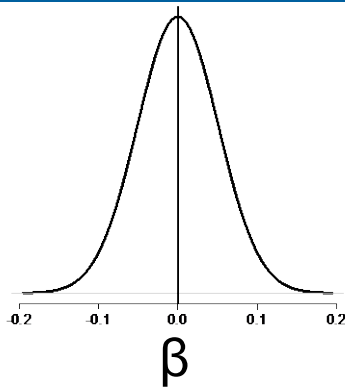
Slope  $p$ -value  $< 10^{-300}$

Overwhelming majority of  
inflation is consistent with  
polygenic architecture





# LD Score regression



Draw polygenic effects from  
 $N(0, n/m^2)$ , var =



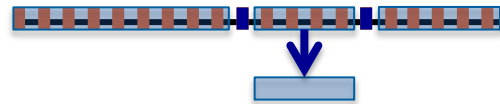
What is the  $E[\chi^2]$  for variant  $j$ ?

$$E[\chi_j^2] = 1 + Na + \frac{h_g^2 N}{M} l_j$$

New estimator of heritability

where  $N$ =sample size,  $M$ =# of SNPs,  $a$ =inflation due to confounding,  
 $h_g^2$  is heritability (total obs.) and  $l_j$  is the *LD Score*

$$l_j = \sum_{k \neq j} r_{jk}^2$$



# What isn't in LD score?



- Genetic variation that is not tagged well by common variation
- Heterogeneity of traits

Thank you!



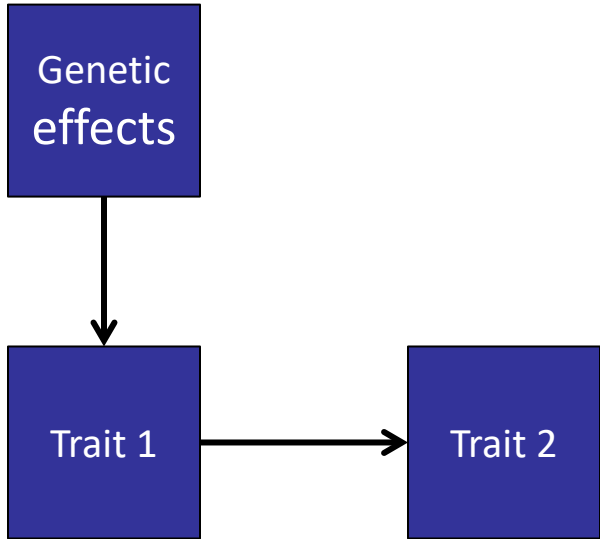


# Genetic Correlation Method in:

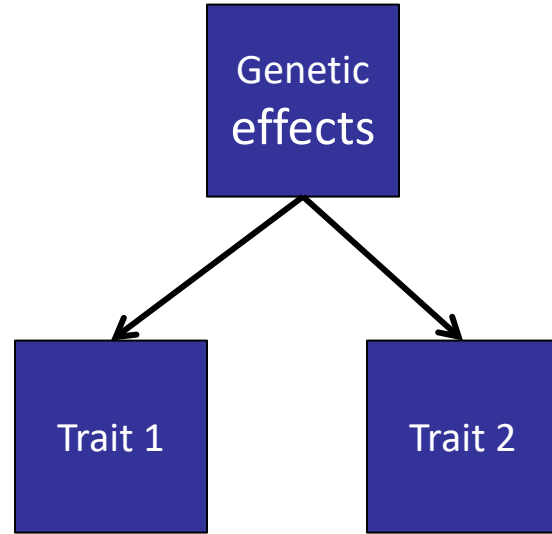
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An atlas of genetic correlations across human diseases  
and traits

# Potential sources of genetic correlation



Trait 1 exerts causal effect on Trait 2



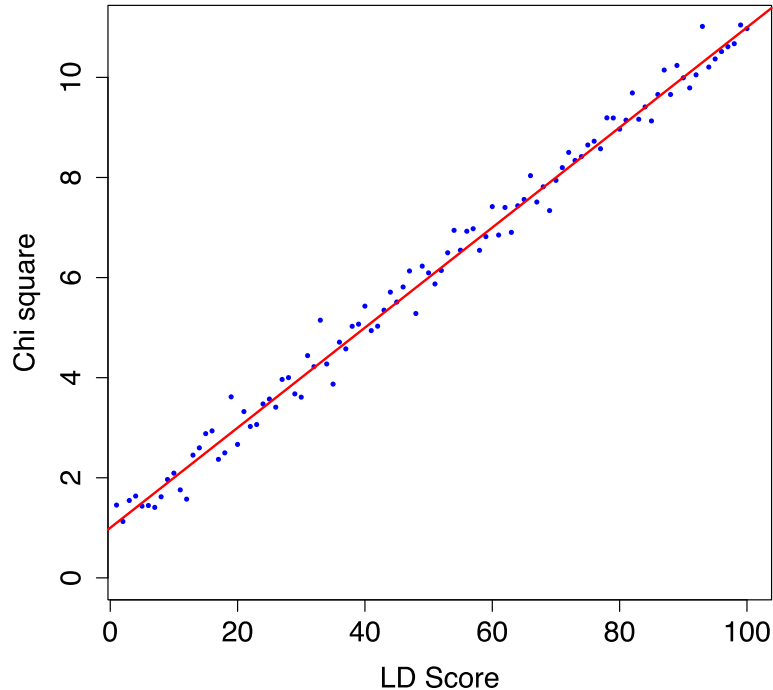
Genetic effects influence  
Trait 1 and Trait 2

# LD Score regression

## Genetic correlation



Trait 1

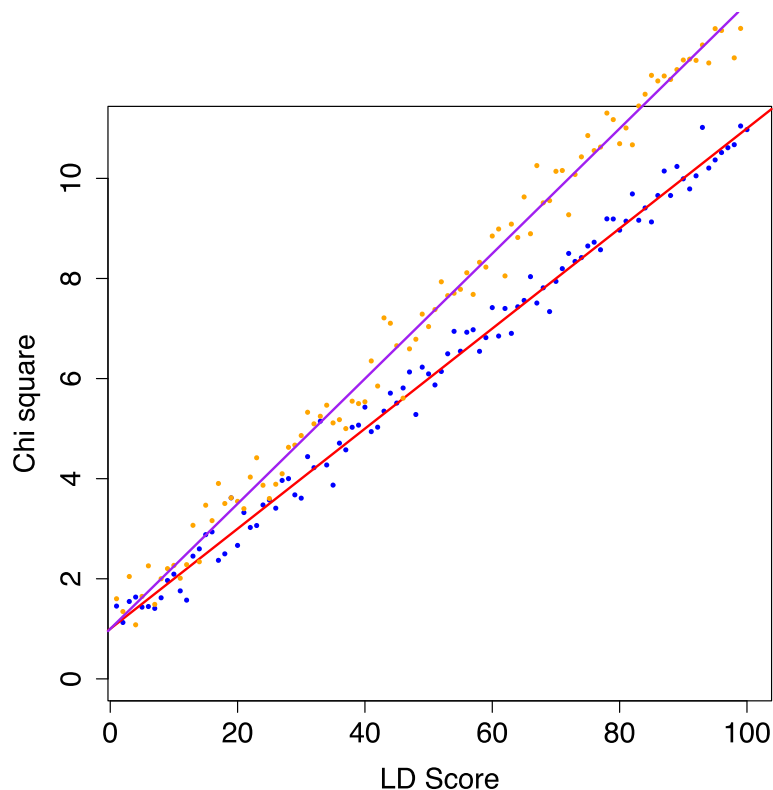


Slope estimates heritability

# LD Score regression

## Genetic correlation

Trait 1  
Trait 2

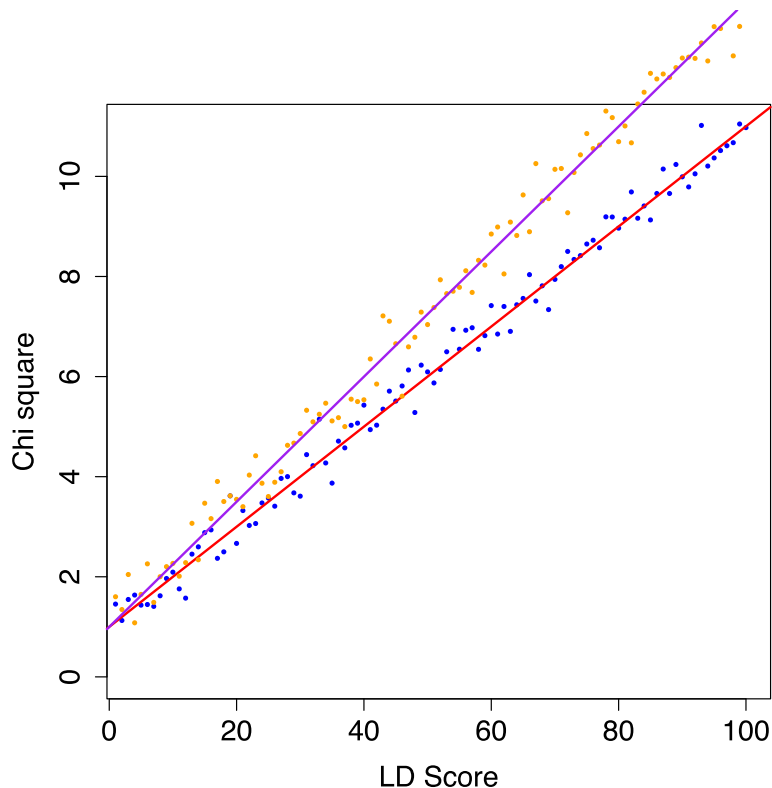


We can a second trait and  
obtain two heritability  
estimates

# LD Score regression

## Genetic correlation

Trait 1  
Trait 2



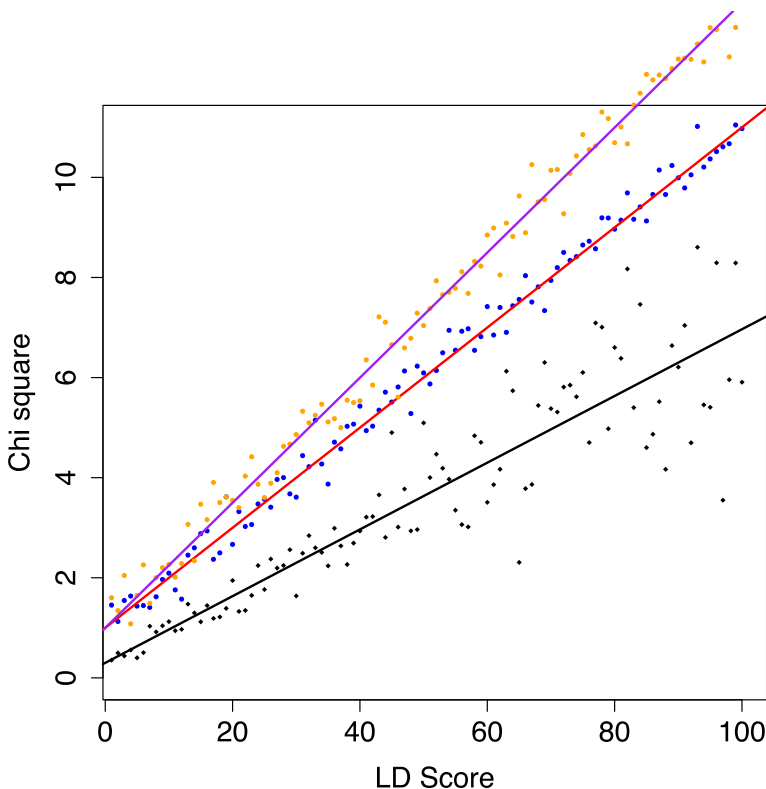
$$\mathbf{Z}^* \mathbf{Z} = \chi^2$$

So we can estimate genetic covariance from the product of the Z-scores



# LD Score regression

## Genetic correlation



Trait 1  
Trait 2  
 $R_G$

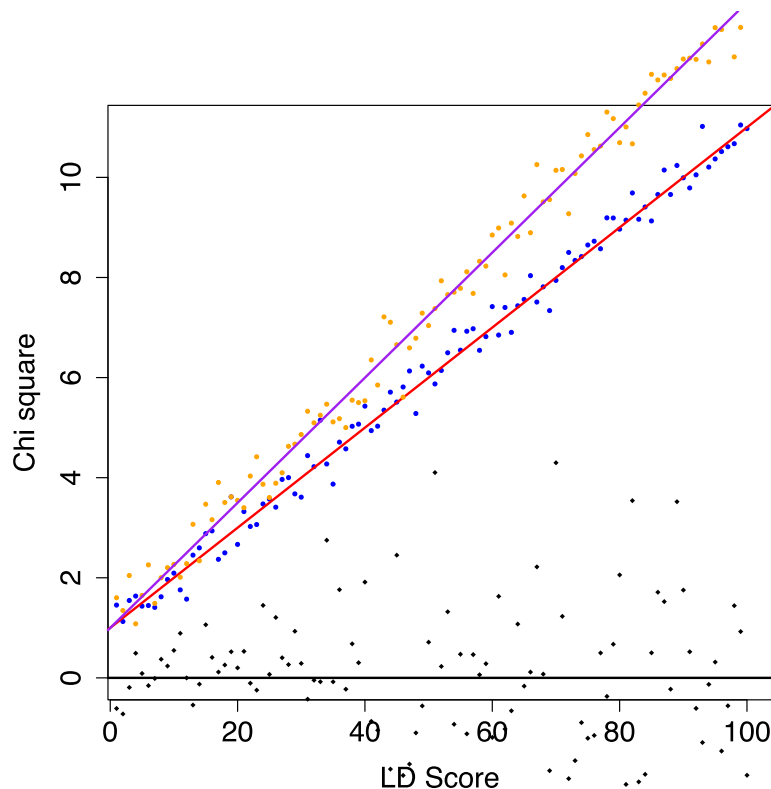
$$Z^*Z = \chi^2$$

So we can estimate genetic covariance from the product of the Z-scores for the two traits

$$R_G = 0.5$$

# LD Score regression

## Genetic correlation



Trait 1  
Trait 2  
 $R_G$

Here  $R_G = 0$

This approach is robust to sample overlap as all variants are equally inflated