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Mendelian randomization: using genetics for guiding drug discovery

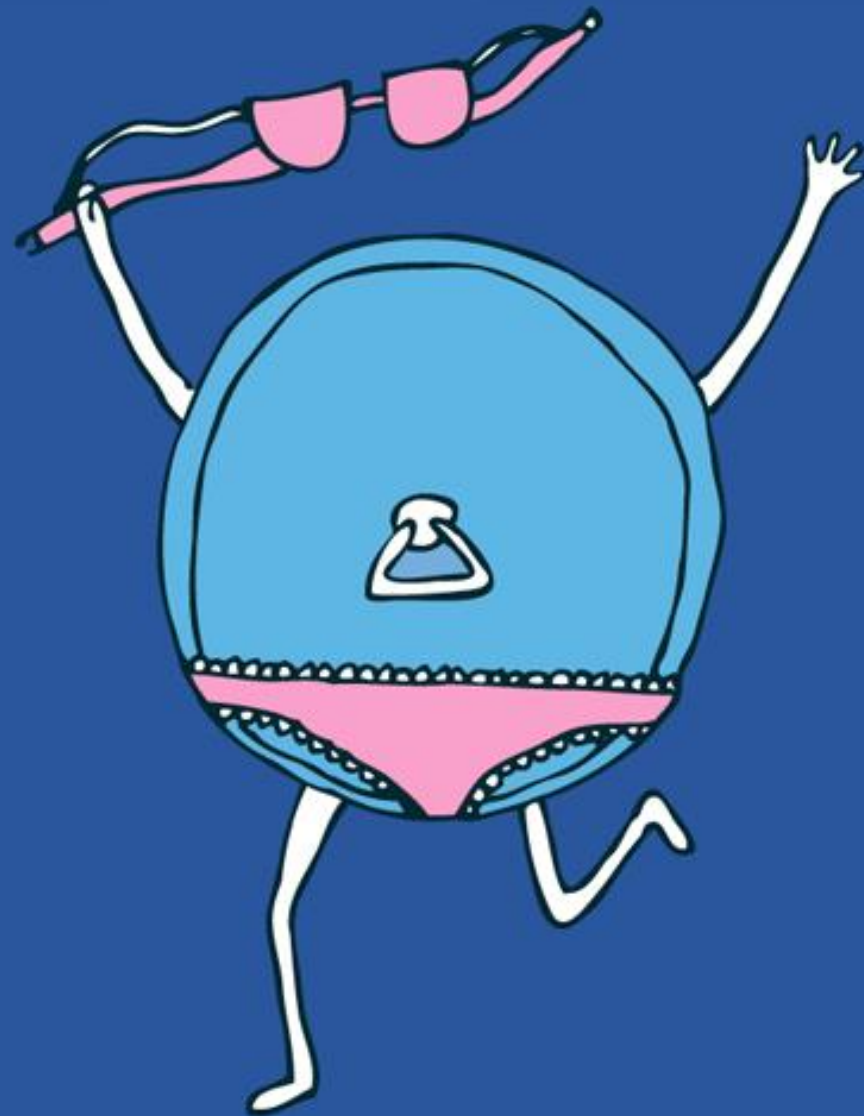
Stephen Burgess

MRC Biostatistics Unit and
Cardiovascular Epidemiology Unit,
University of Cambridge

PSI One Day Meeting, 29th January 2019



Twitter: @stevesphd
sb452@medschl.cam.ac.uk





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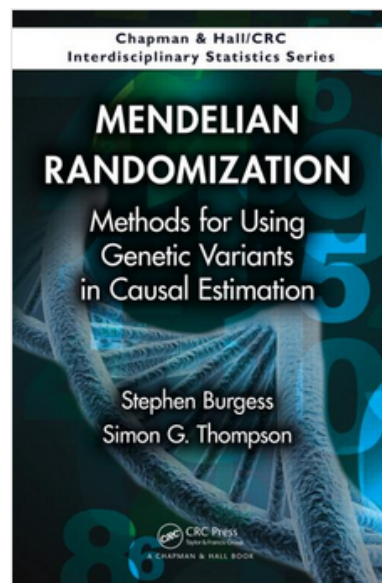
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Mendelian Randomization: Methods for Using Genetic Variants in Causal Estimation

Stephen Burgess, Simon G. Thompson

Hardback
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February 26, 2015 **Forthcoming** by Chapman and Hall/CRC

Reference - 224 Pages - 22 B/W Illustrations

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Description

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- Offers first-hand, in-depth guidance on Mendelian

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Two-day course

- Wednesday 27th and Thursday 28th March 2019
- MRC Biostatistics Unit, Cambridge
- Details at www.mendelianrandomization.com

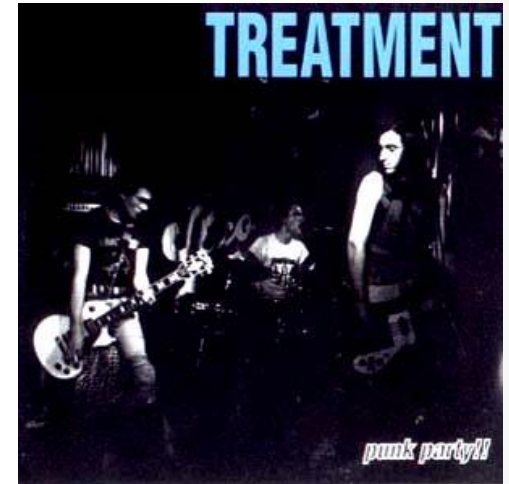
Observational data

- Correlation is not causation
- Observed associations between a risk factor and an outcome may result from:
 - Confounding
 - Reverse causation



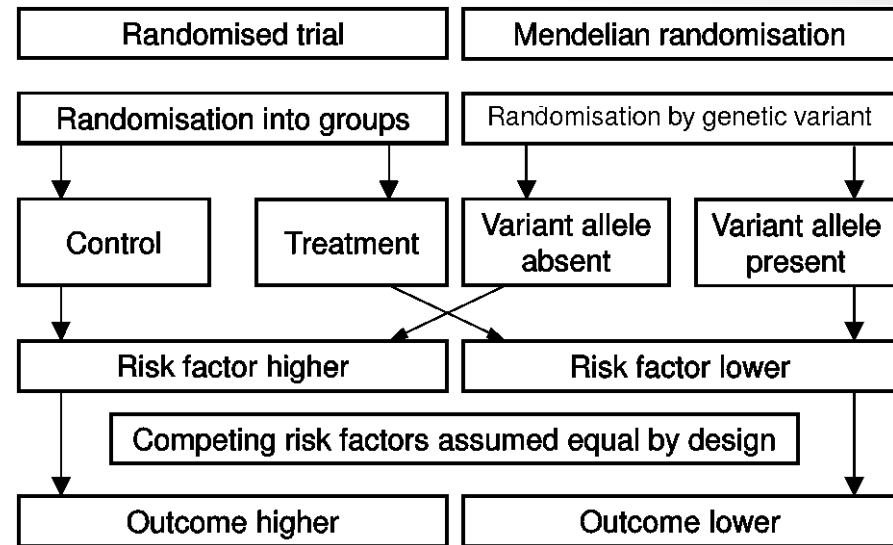
Randomized trials

- In a randomized trial, participants are randomly assigned to a treatment group
- Random allocation ensures all potential confounders are equally distributed (on average) between the groups
 - Unmeasured confounders
- Allocation of treatment gives a time-ordering
 - Reverse causation



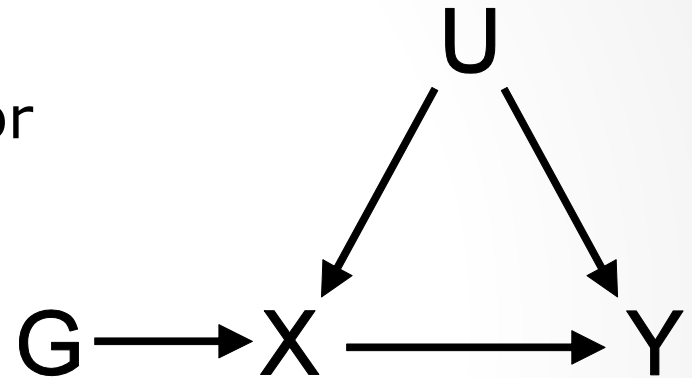
Assessment of causality

- Mendelian randomization is analogous to a randomized trial
- An association between the genetic variant and the outcome is indicative of a causal effect of the risk factor on the outcome



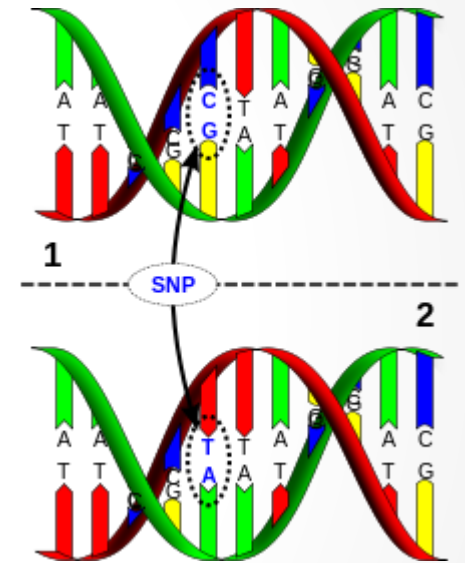
Instrumental variables

- An instrumental variable is:
 1. Associated with the risk factor of interest
 2. Not associated with any potentially confounding variable
 3. Only associated with the outcome via the risk factor
- Provides a natural experiment in observational data, similar to a randomized trial



Mendelian randomization

- Genetic variants are particularly suitable candidate instrumental variables
 - Scientific knowledge of genetic function
 - Specific association with traits
 - Genetic sequence is determined at conception
- Mendelian randomization is the use of genetic instrumental variables in observational data to obtain causal inferences

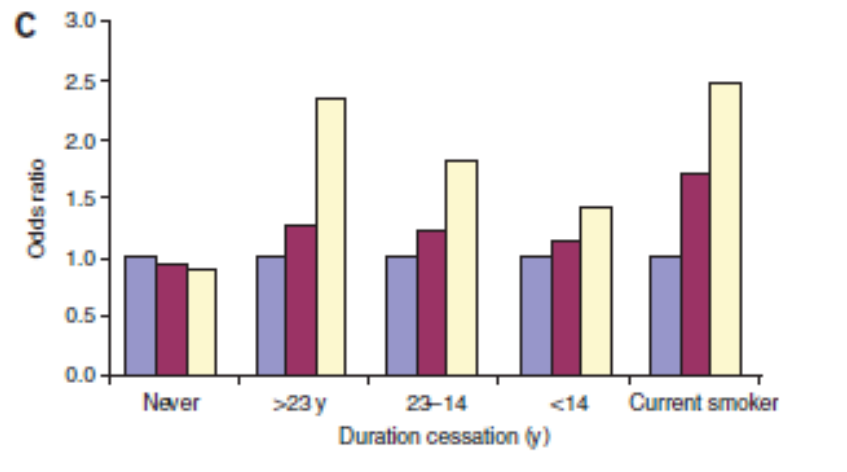


Role of Mendelian randomization in drug discovery

- Target discovery:
 - identify new targets
- Target validation: Offer insight into existing/putative targets:
 - increase or reduce enthusiasm
 - insight into safety profile
 - identify opportunities for repositioning
 - identify interactions between treatments
 - identify subgroups with greater efficacy
 - estimate impact in clinical trial

Genetics can recapitulate known causal associations

- GWAS of lung cancer identified a genomic region associated with lung cancer
 - Biological hypothesis: causal gene may regulate downstream signalling pathways and impact cell proliferation
 - Hint that SNP may not be associated in never smokers



- Genetics has “discovered” that smoking causes lung cancer!

LETTERS

A variant associated with nicotine dependence, lung cancer and peripheral arterial disease

Thorgeir E. Thorgeirsson^{1*}, Frank Geller^{1*}, Patrick Sulem^{1*}, Thorunn Rafnar^{1*}, Anna Wiste^{1,2}

Table 1 | Genotype status and SQ level of 13,945 Icelandic smokers

Parameter	Genotype of rs1051730			Total n (frequency)	Frequency of T allele
	GG	GT	TT		
Cigarettes per day (SQ level)					
1 to 10 (0)	1,743	1,558	326	3,627 (0.260)	0.305
11 to 20 (1)	2,727	2,865	810	6,402 (0.459)	0.350
21 to 30 (2)	1,145	1,416	427	2,988 (0.214)	0.380
31 and more (3)	341	448	139	928 (0.067)	0.391
All levels (frequency)	5,956 (0.427)	6,287 (0.451)	1,702 (0.122)	13,945 (1.000)	0.347
Mean SQ level (mean \pm s.d.)	1.01 \pm 0.85	1.12 \pm 0.86	1.22 \pm 0.85	1.09 \pm 0.86	

Table 4 | Association of rs1051730 allele T with LC and PAD

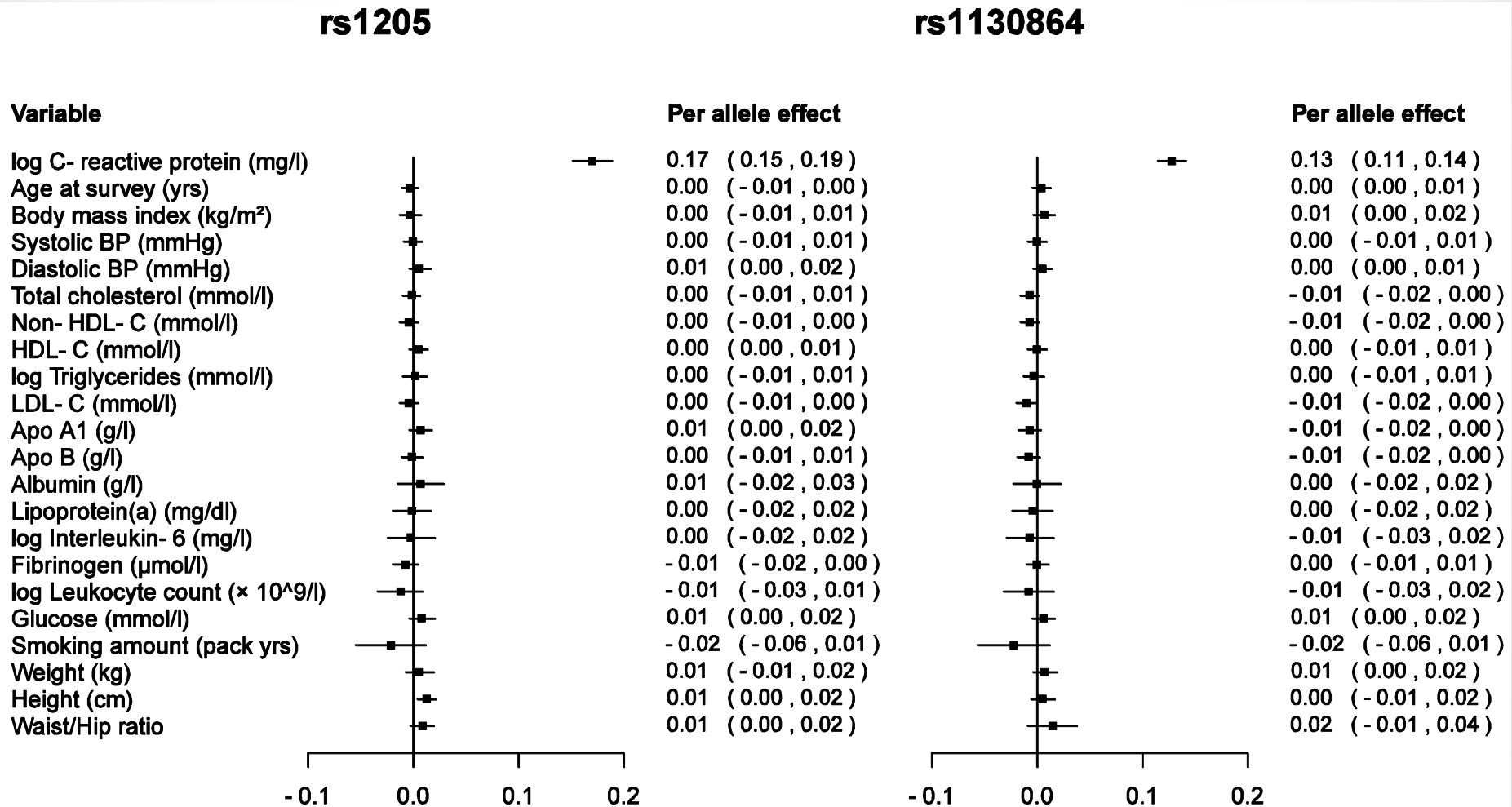
Study group	Controls		Cases		OR (95% CI)	P
	n	Frequency	n	Frequency		
LC						
Iceland	28,752	0.342	665	0.398	1.27 (1.13–1.43)	4.1×10^{-5}
Spain	1,474	0.390	269	0.483	1.46 (1.22–1.76)	5.4×10^{-5}
The Netherlands	2,018	0.314	90	0.350	1.18 (0.86–1.61)	0.31
Foreign combined	3,492	–	359	–	1.38 (1.18–1.62)	6.6×10^{-5}
All combined	32,244	–	1,024	–	1.31 (1.19–1.44)	1.5×10^{-8}
PAD						
Iceland	28,752	0.342	1,503	0.379	1.18 (1.09–1.27)	5.3×10^{-5}
New Zealand	435	0.274	441	0.337	1.35 (1.10–1.65)	0.0041
Austria	403	0.352	457	0.395	1.20 (0.99–1.46)	0.068
Sweden	140	0.304	172	0.331	1.14 (0.81–1.60)	0.46
Italy	234	0.378	165	0.412	1.15 (0.86–1.54)	0.33
Foreign combined	1,212	–	1,235	–	1.23 (1.09–1.39)	5.9×10^{-4}
All combined	29,964	–	2,738	–	1.19 (1.12–1.27)	1.4×10^{-7}

Example: C-reactive protein and coronary heart disease

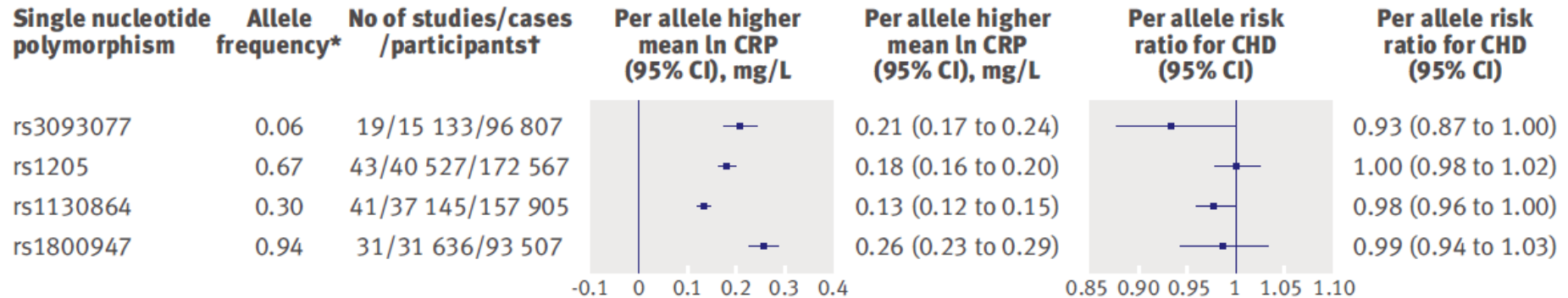
- C-reactive protein is observationally associated with coronary heart disease risk
- But that does not necessarily mean that it is a causal risk factor
- Inflammation hypothesis:
 - Does inflammation lead to heart disease?
 - Are inflammatory biomarkers raised in response to pre-clinical disease?
 - Are both processes driven by another mechanism?
- Can use common genetic variation in the *CRP* gene region to assess question of causation

Example: CRP and CHD risk

- Assessing associations with measured covariates



Example: CRP and CHD risk



Circulating usual concentrations of CRP

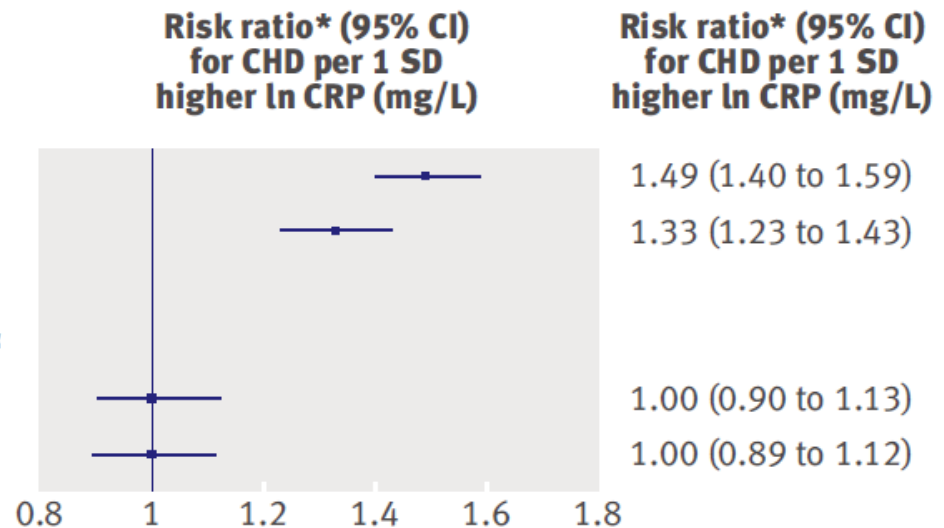
Adjusted for age, sex, and ethnicity

Further adjusted†

Genetically raised concentrations of CRP‡

SNP analyses

Haplotype analyses



Taxonomy of MR studies

1. Risk factor is a protein biomarker

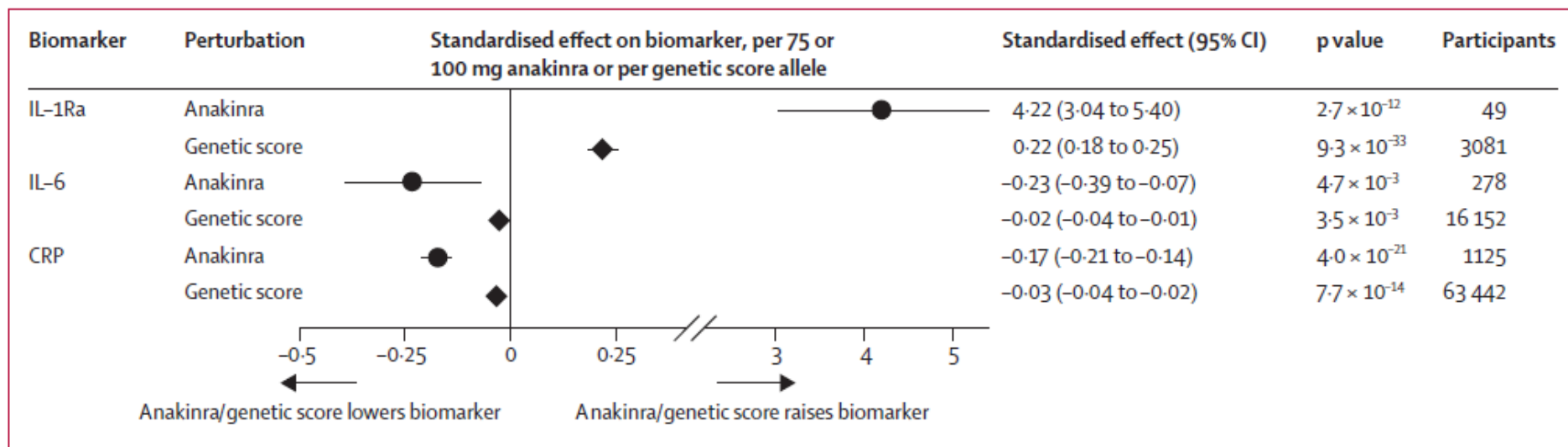
- One or a small number of genetic variants
- Usually one gene region
- Gene region has strong biological link with risk factor
- Often, we can develop drugs to act on the same biological pathway
- The most plausible assessment of causation

Example: interleukin-1 and cardiovascular outcomes

- Interleukin-1 (IL1) is another inflammatory biomarker
- We consider two genetic variants in the *IL1RN* gene region that encodes the IL-1 receptor antagonist
- Anakinra, the recombinant form of IL-1 receptor antagonist, is used in the treatment of rheumatoid arthritis

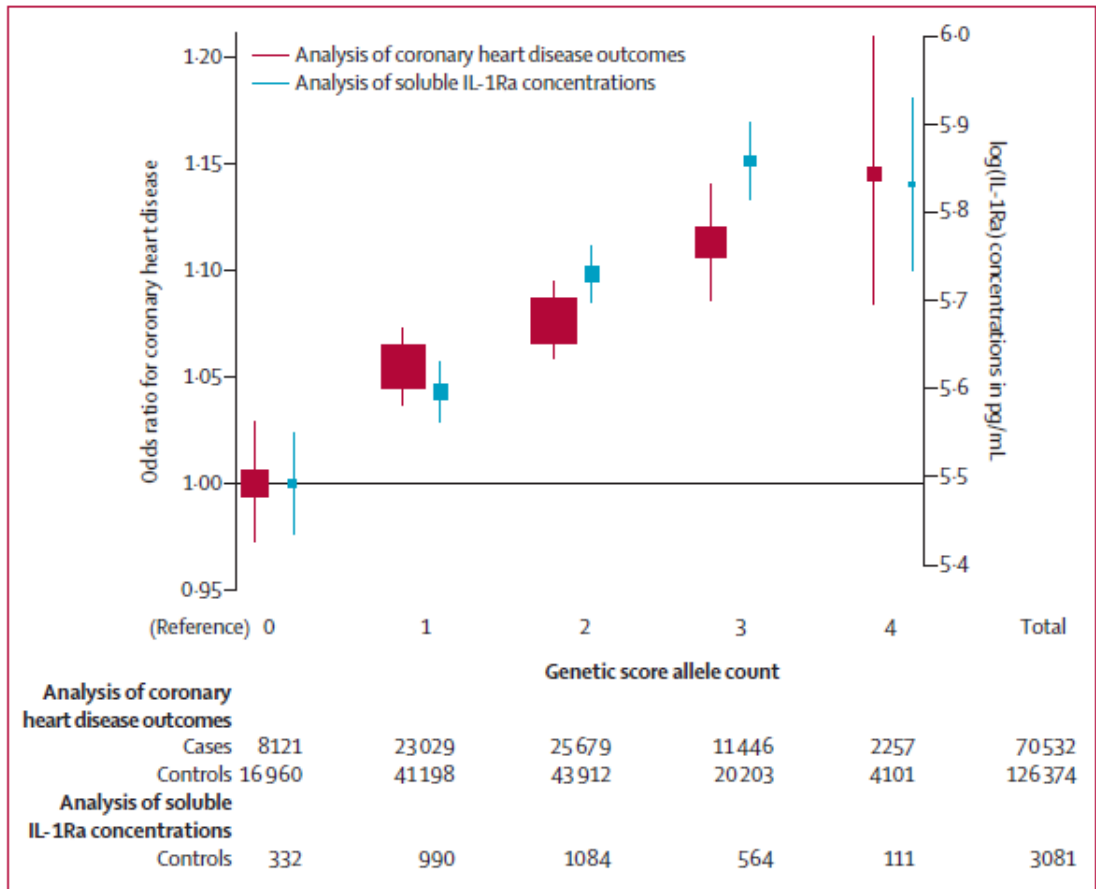
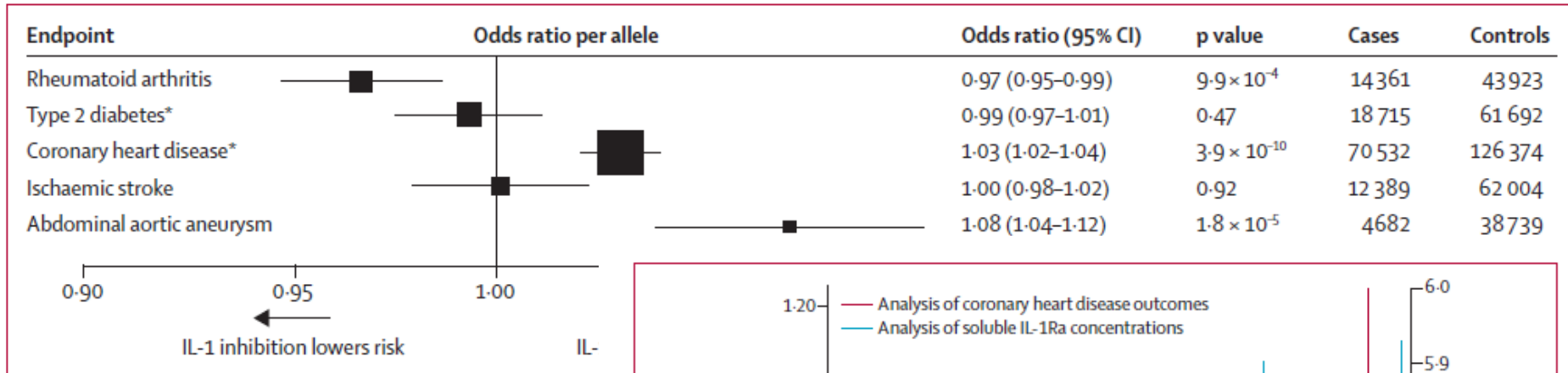
Example: IL1 and CVD risk

- We can compare known effects of anakinra on intermediate traits to the associations of the genetic variants with these traits

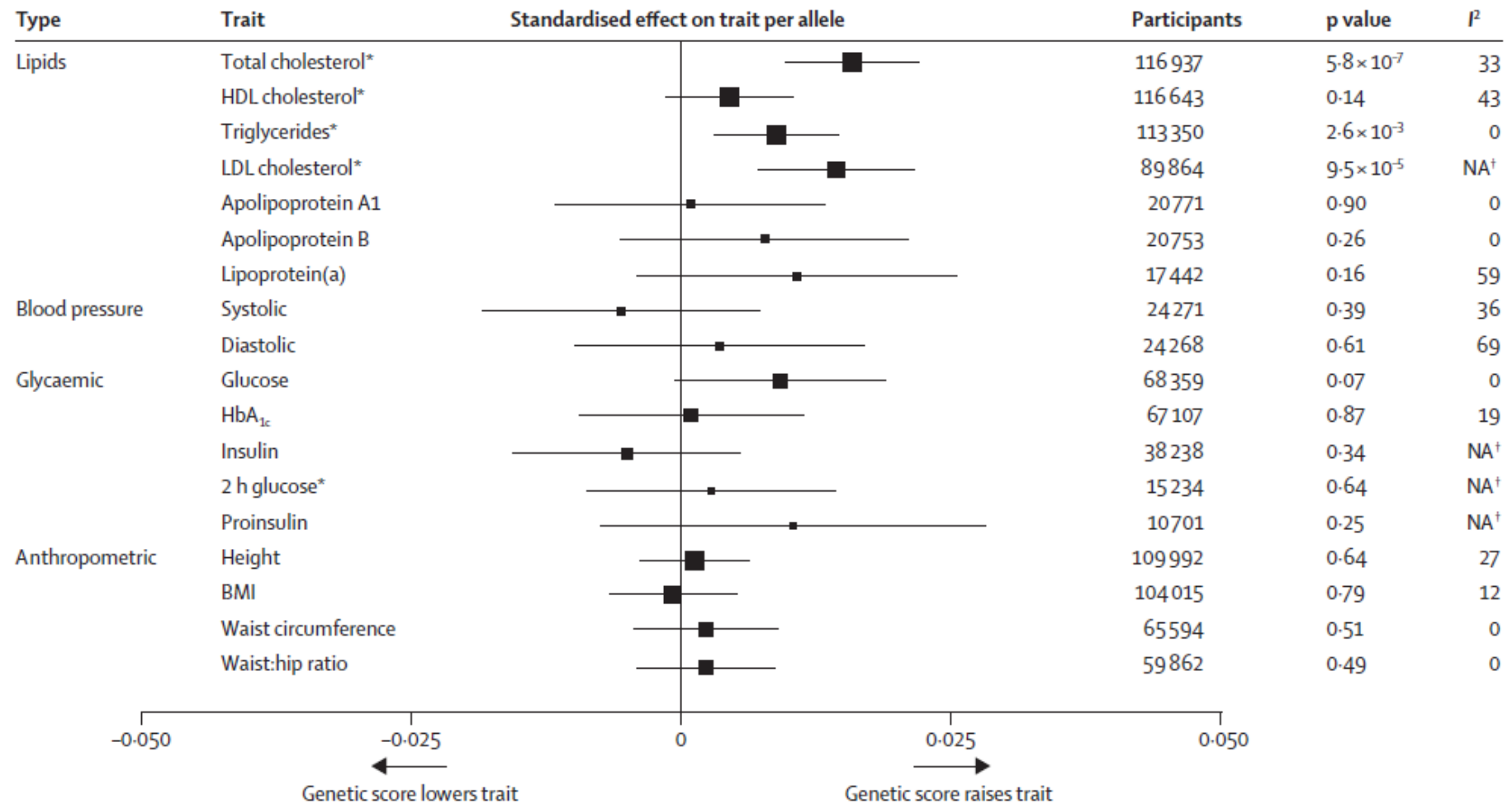


Freitag et al. "Cardiometabolic effects of genetic upregulation of the interleukin 1 receptor antagonist: a Mendelian randomisation analysis", Lancet D+E 2015

Example: IL1 and CVD risk



Example: IL1 and CVD risk

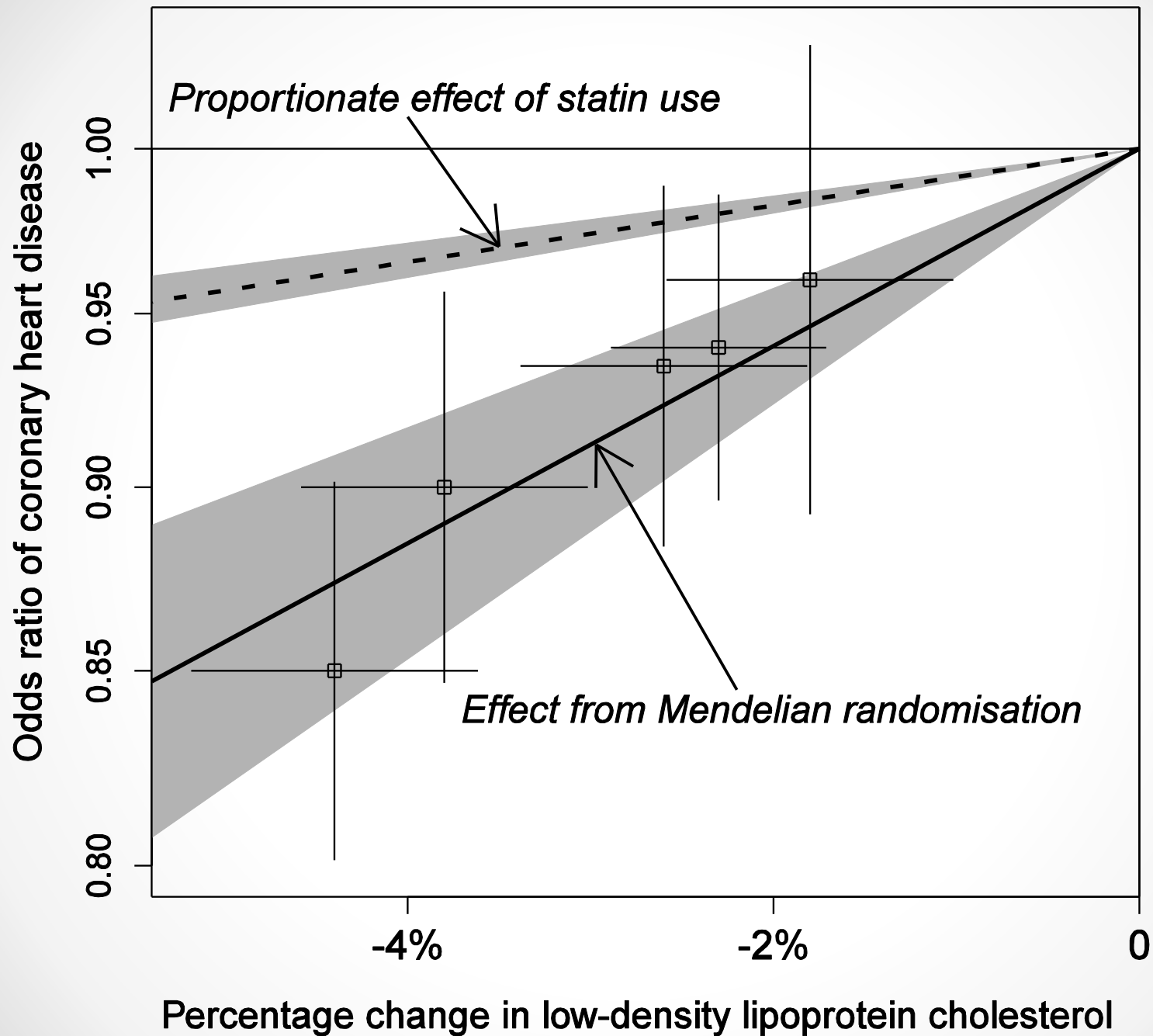


Interpretation of a causal effect

- A Mendelian randomization estimate represents the change in the outcome from a change in the risk factor corresponding to a life-long difference in the usual levels of the risk factor from conception
- This typically differs from the effect of lowering/raising the risk factor in practice, such as by a clinical/pharmacological intervention
- Why might estimates differ?
- When does this matter?

Long-term versus short-term

- Genetic effects are typically life-long, clinical interventions are often short-term
 - Atherosclerosis is a chronic condition
 - We would expect the impact of a one-day lowering in LDL-cholesterol levels to be less than that of a one-year intervention, or a five-year intervention
 - So we would expect the impact of genetically lowered LDL-c to be greater even than that of a five-year intervention
 - In contrast, we may expect there to be less difference between the MR estimate and trial estimate for the risk factor of blood pressure



Long-term versus short-term

- For lipids, the causal estimate from Mendelian randomization of the effect of LDL-c on CHD risk is 3-3.5 times the proportionate effect of statins on CHD risk
 - (assuming 5 year treatment in primary prevention: 25% lower lipids, 30% lower CHD risk)
- For blood pressure, the MR estimate is 2.5 times the effect of SBP/DBP on CHD risk

Usual versus pathological levels

- Is CRP causal for coronary heart disease risk?
 - C-reactive protein (CRP) is an acute phase reactant
 - CRP concentrations increase sharply during disease episodes
 - Genetic associations with CRP are usually estimated in healthy individuals
 - There are suggestions that genetic predictors of “usual” CRP levels are also associated with pathological concentrations
 - However, the primary question addressed in a Mendelian randomization analysis is: “do long-term elevated levels of CRP cause CHD?” and not “do acute levels of CRP cause CHD?”
 - Long-term question more relevant for disease prevention? (eg polypill)

Lipoprotein(a) and CHD risk

- Lipoprotein(a) is highly heritable
- Concentration differs by 1000 fold between individuals
- *LPA* gene region explains up to 90% of the variance in Lp(a)
 - I am somewhat sceptical of this claim, as several Lp(a) assays are isoform-sensitive, but certainly highly (60%+) heritable
- Compounds to lower Lp(a) have passed Phase 2 trials
- These compounds lower Lp(a) by up to 90%

Lipoprotein(a) and CHD risk

Genetically Elevated and Increased Risk c

Pia R. Kamstrup, MD, PhD

Anne Tybjaerg-Hansen, MD, DMSc

Rolf Steffensen, MD

Borge G. Nordestgaard, MD, DMSc

Context Hi
dial infarction

Objective
causal.

Design Co

Quartiles

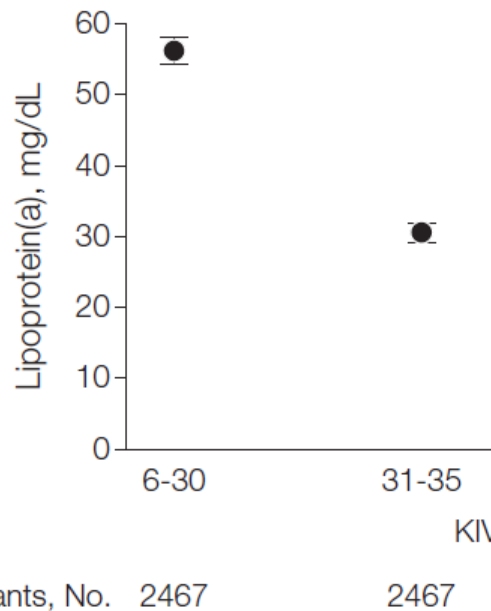
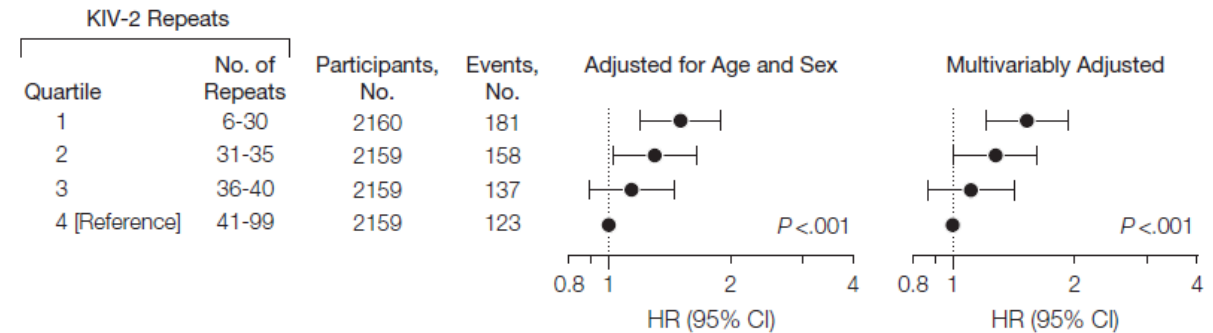
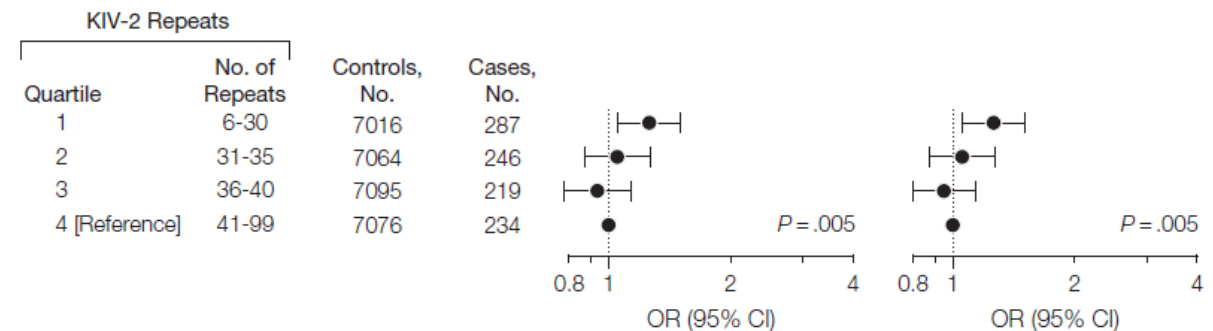


Figure 3. Risk of Myocardial Infarction by Quartiles of Apolipoprotein(a) KIV-2 Repeats in the CCHS, CGPS, and CIHDS

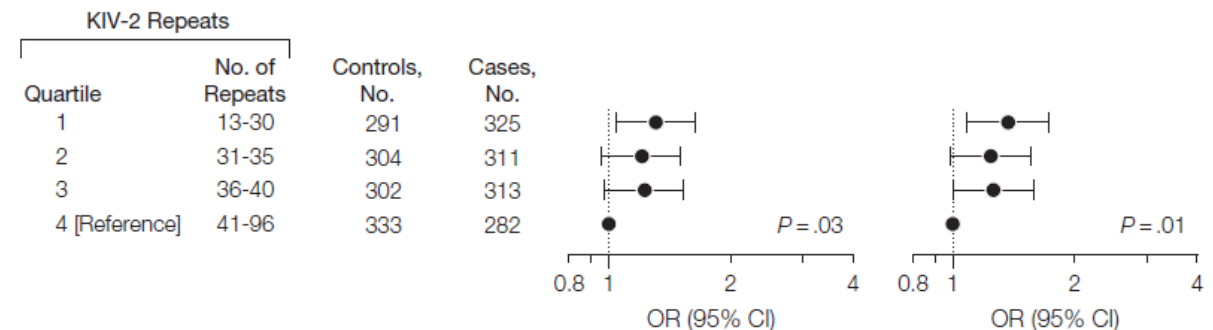
CCHS



CGPS



CIHDS



ORIGINAL ARTICLE

Genetic Variants Associated with Lp(a) Lipoprotein Level and Coronary Disease

Robert Clarke, F.R.C.P., John F. Peden, Ph.D., Jemma C. Hopewell, Ph.D., Theodosios Kyriakou, Ph.D., Anuj Goel, M.Sc., Simon C. Heath, Ph.D., Sarah Parish, D.Phil., Simona Barlera, M.S., Maria Grazia Franzosi, Ph.D., Stephan Rust, Ph.D., Derrick Bennett, Ph.D., Angela Silveira, Ph.D., Anders Malarstig, Ph.D., Fiona R. Green, Ph.D., Mark Lathrop, Ph.D., Bruna Gigante, M.D., Karin Leander, Ph.D., Ulf de Faire, M.D., Udo Seedorf, Ph.D., Anders Hamsten, F.R.C.P., Rory Collins, F.R.C.P., Hugh Watkins, F.R.C.P., and Martin Farrall, F.R.C.Path., for the PROCARDIS Consortium*

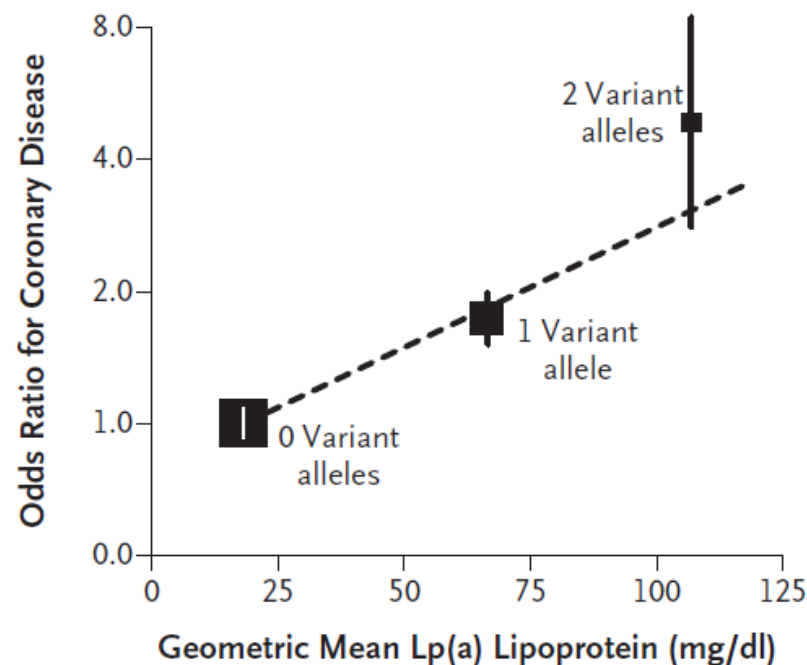
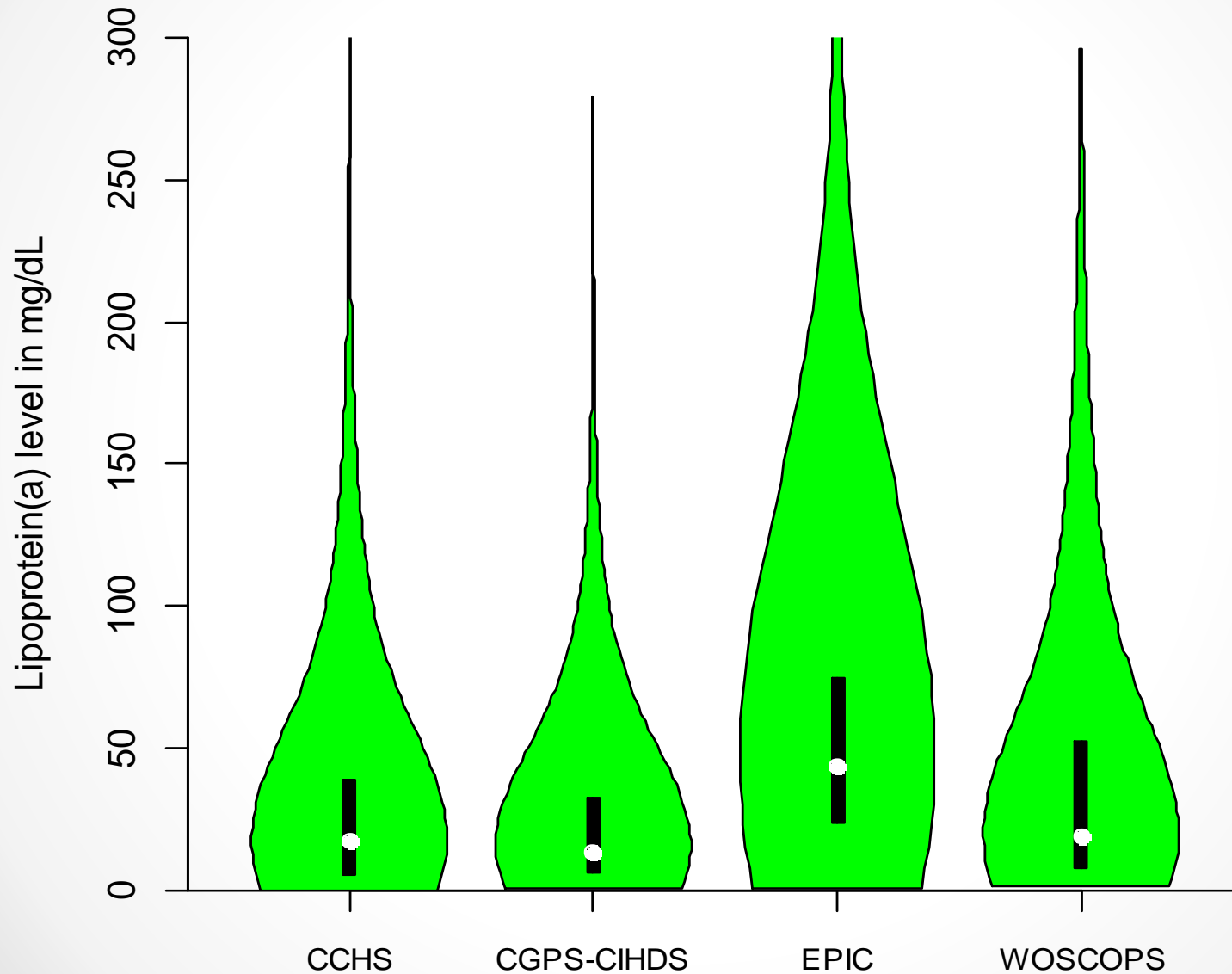


Figure 3. Association of the *LPA* Genotype Score with the Lp(a) Lipoprotein Level and the Risk of Coronary Disease in the PROCARDIS Cohort.

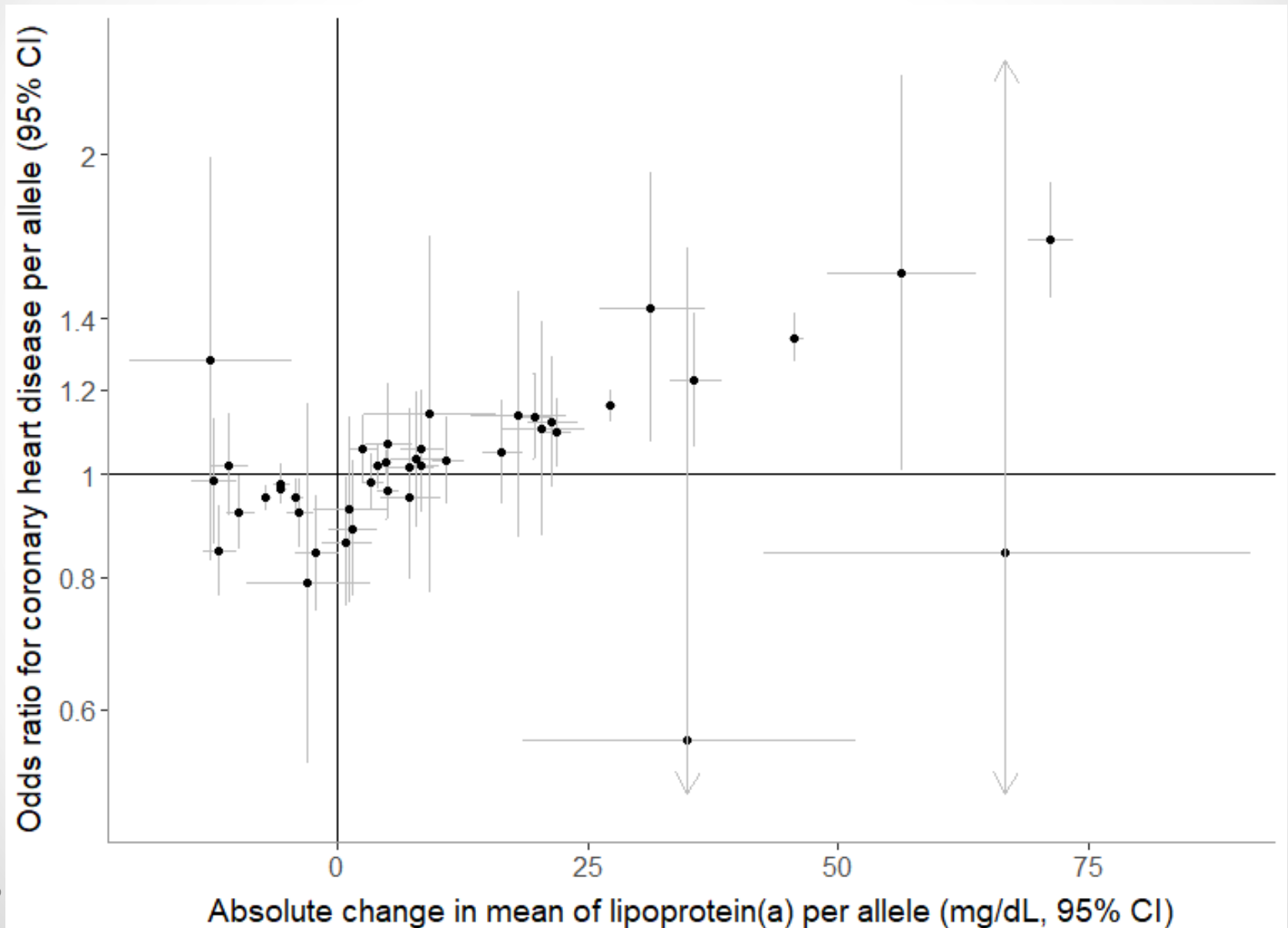
Lipoprotein(a) and CHD risk

- However, several previous treatments that lower Lp(a) by 30-40% have not been successful in Phase 3 trials
- To what extent must we lower Lp(a) to improve disease risk?
- We considered ultra-fine mapping in the LPA gene region

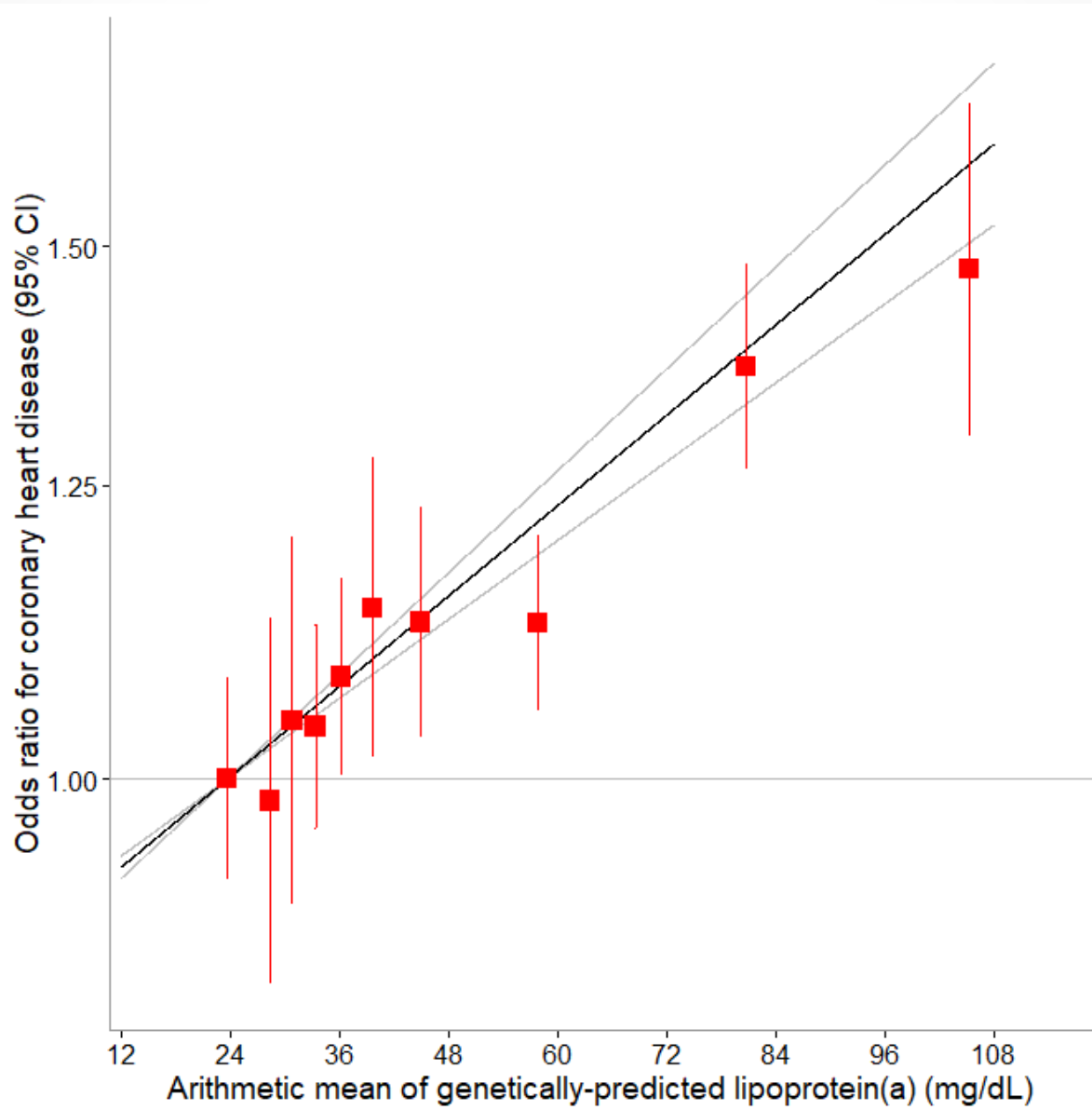
Lipoprotein(a) and CHD risk



Lipoprotein(a) and CHD risk



Lipoprotein(a) and CHD risk

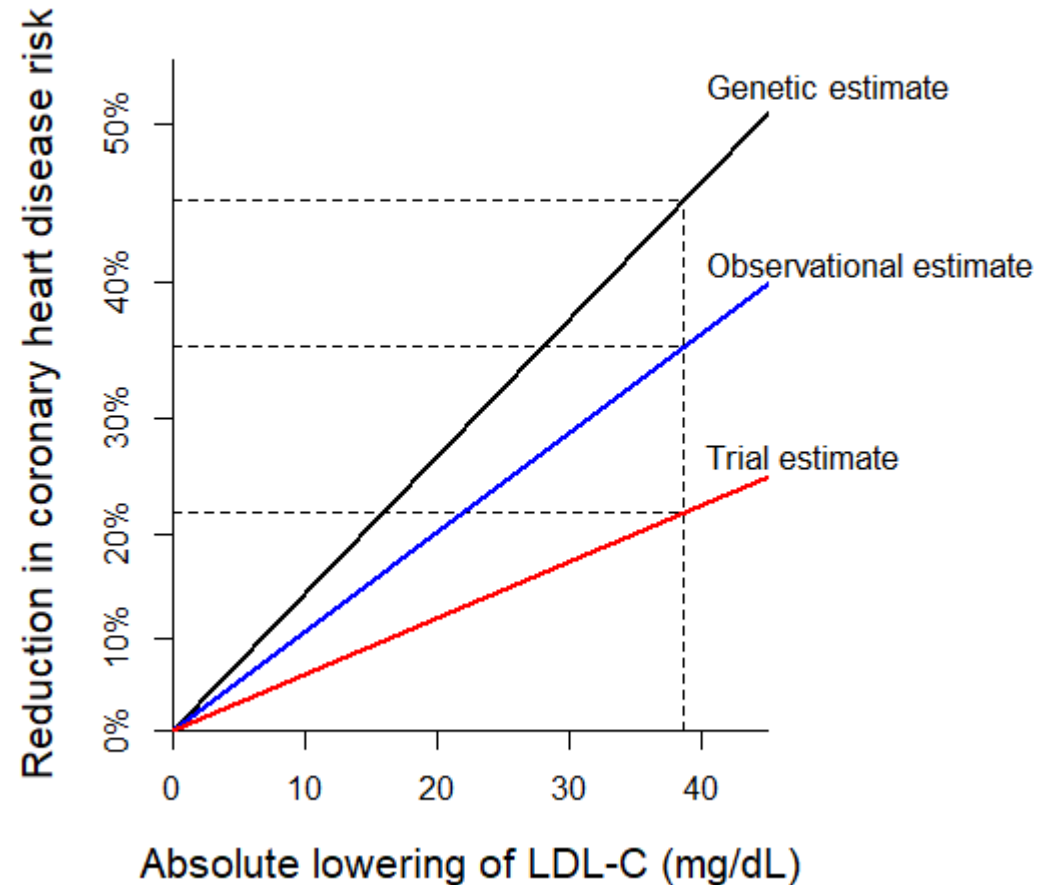


Lipoprotein(a) and CHD risk

- Linear relationship between Lp(a) and log-CHD risk
- Makes more sense to model Lp(a) on absolute scale rather than log-transformed scale
- Changes in CHD risk are proportional to absolute change in LDL-cholesterol concentration
- Need to trim/winsorize to account for outliers
- But MR estimate represents effect of life-long intervention in Lp(a)...
- ... how to find effect of short-term lowering?

Lipoprotein(a) and CHD risk

- We assume that Lp(a) and LDL-cholesterol are similar
- In that ratio of life-long (MR) to short-term effect is assumed to be same between Lp(a) and LDL-cholesterol



Lipoprotein(a) and CHD risk

Approach 1:

Step 1:
$$\frac{\text{MR estimate for lipoprotein(a) per 10mg/dL lowering}}{\text{MR estimate for LDL-cholesterol per 10mg/dL lowering}} = \text{Ratio of atherogenicity}$$

Step 2:
$$\frac{\text{Proposed magnitude of lipoprotein(a) reduction in mg/dL} \times \text{Ratio of atherogenicity}}{\text{Equivalent reduction in LDL-cholesterol in mg/dL}} =$$

Step 3:
$$\exp\left(\frac{\text{Equivalent reduction in LDL-cholesterol}}{38.67} \times \log(0.78)\right) =$$

Predicted odds ratio for CHD risk reduction in short-term trial

Approach 2:

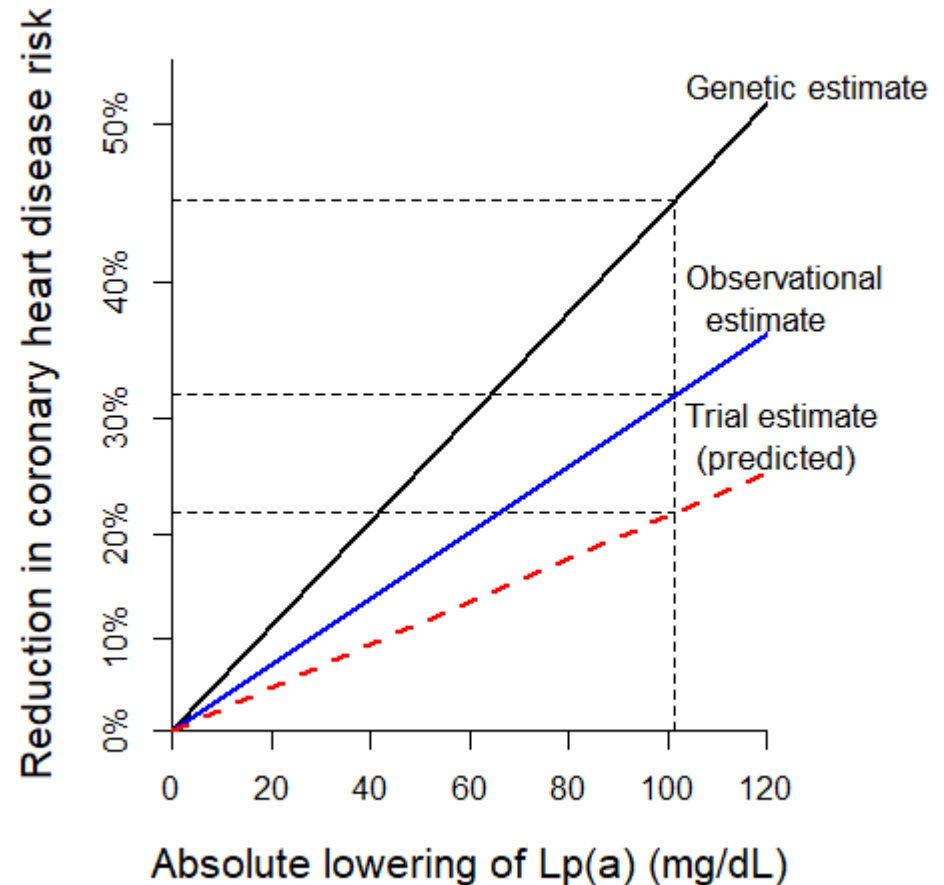
Step 1:
$$\frac{\text{Trial estimate for LDL-cholesterol per 1mmol/L lowering}}{\text{MR estimate for LDL-cholesterol per 1mmol/L lowering}} = \text{Ratio of short-term to life-long effects}$$

Step 2:
$$\exp(\text{MR estimate for lipoprotein(a) lowering} \times \text{Ratio of short-term to life-long effects}) =$$

Predicted odds ratio for CHD risk reduction in short-term trial

Lipoprotein(a) and CHD risk

- Based on the MR estimates, a ~ 100 mg/dL change in Lp(a) is equivalent to a 38.67mg/dL (1mmol/L) change in LDL-cholesterol
- Hence short-term 100 mg/dL change in Lp(a) will have similar effect to statins in a short-term trial



Lipoprotein(a) and CHD risk

Table 2: Expected Clinical Benefit of Lowering Lp(a)

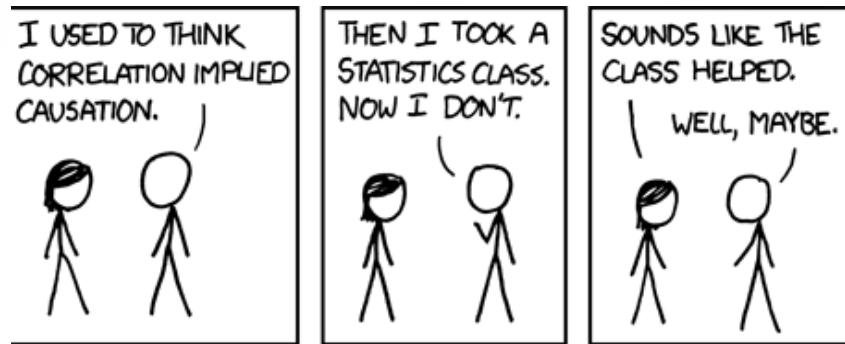
Reduction in Lp(a) (mg/dL)	Reduction in LDL-C for equivalent CHD risk reduction (mg/dL) ¹	Estimated life-long proportional risk reduction due to genetically decreased exposure (%) ²	Estimated short-term proportional risk reduction in randomized trial (%) ³
120	45.7 (34.1, 65.4)	51.1 (45.5, 56.2)	27.7 (20.9, 37.5)
100	38.1 (28.4, 54.5)	44.9 (39.7, 49.8)	23.7 (17.8, 32.4)
80	30.5 (22.7, 43.6)	38.0 (33.2, 42.3)	19.4 (14.5, 26.9)
50	19.0 (14.2, 27.3)	25.8 (22.3, 29.1)	12.6 (9.3, 17.8)
30	11.4 (8.5, 16.4)	16.4 (14.1, 18.7)	7.8 (5.7, 11.1)
20	7.6 (5.7, 10.9)	11.3 (9.6, 12.9)	5.3 (3.8, 7.5)
10	3.8 (2.8, 5.5)	5.8 (4.9, 6.7)	2.7 (1.9, 3.9)
5	1.9 (1.4, 2.7)	2.9 (2.5, 3.4)	1.3 (1.0, 1.9)

Lipoprotein(a) and CHD risk

- Only ~1% of the population have Lp(a) >100mg/dL
- For someone with Lp(a) 20mg/dL, a 30-40% reduction would not provide sufficient benefit
- Hence need to target those with high Lp(a)
- Drugs may be more effective in African descent individuals, as they have higher average Lp(a) concentrations
- Will be ineffective in East Asians
- But, the lifelong effect of Lp(a) >160mg/dL is similar OR to FH – and similar prevalence
- Screening strategy (genetic or phenotypic)

Summary

- Genetic variants can be used to obtain causal inferences about modifiable risk factors from observational data
- Mendelian randomization investigations are becoming easier to perform and more powerful due to summarized data
- Question marks:
 - how to choose variants?
 - how to interpret?
 - how to prioritize?
- Quality of genetic variants as instrumental variables (genetic proxies) is crucial to the quality of evidence
- Estimand: what are we estimating?



Thank you!

