### **Mendelian Randomization**

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University of Queensland University of Bristol













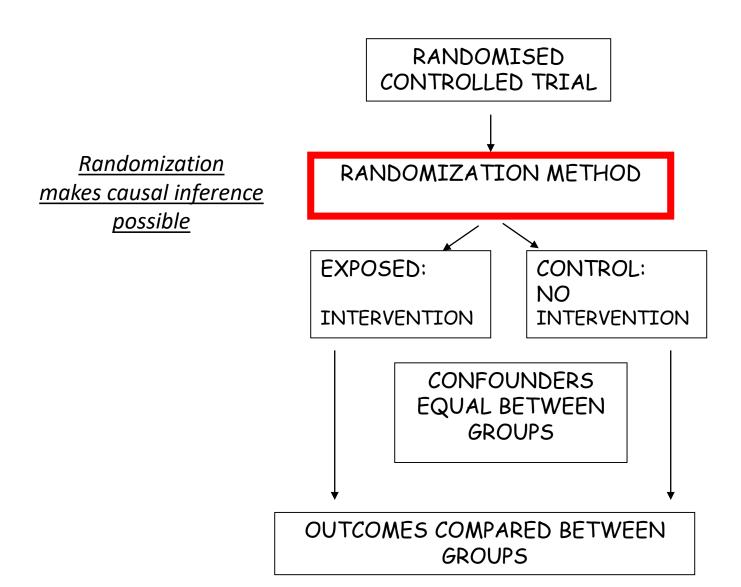
### Some Criticisms of GWA Studies...

- How do you translate the results from GWAS?
- You can't change people's genotypes (at least not yet)
- You can however modify people's environments...
- Mendelian Randomization is a method of using genetics to inform us about associations in traditional observational epidemiology and MUCH MUCH more...

### **This Session**

- Determining causality in observational studies
- Mendelian randomization (MR)
- An Example of MR
- MRBase
- MR and Drug Development
- Practical

### **RCTs: the Gold Standard in Inferring Causality**

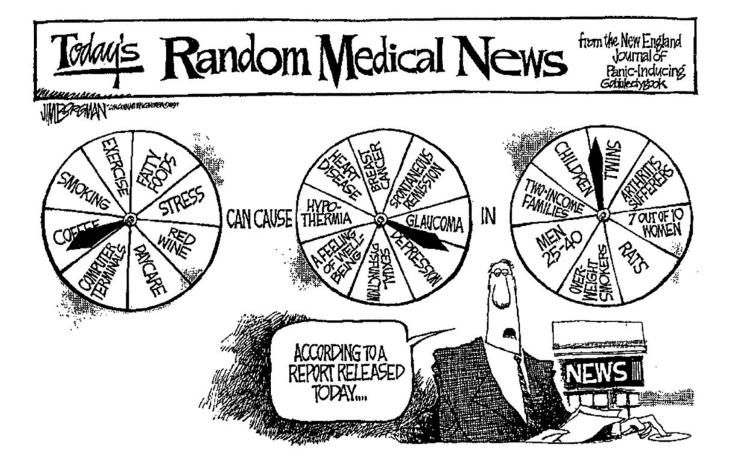


## The Need for Observational Studies

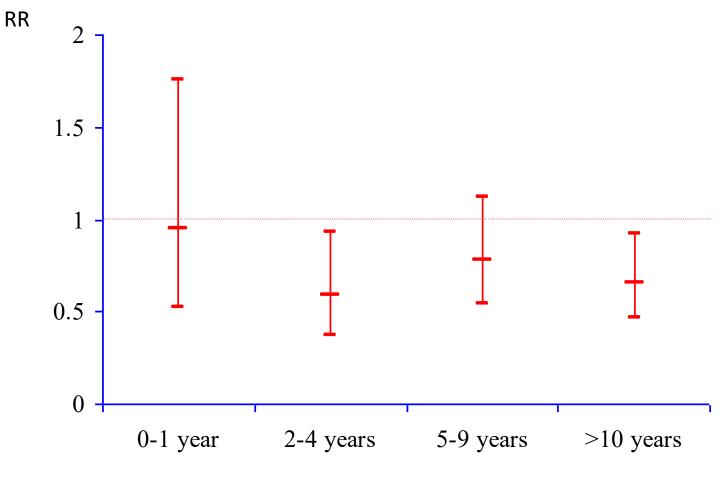
- Randomized Controlled Trials (RCTs):
  - Not always ethical or practically feasible eg anything toxic
  - Expensive, requires experimentation in humans
  - Should only be conducted on interventions that show very strong observational evidence in humans

#### Observational studies:

- Association between environmental exposures and disease measured in observational designs (non-experimental) eg case-control studies or cohort studies
- Reliably assigning causality in these types of studies is very limited



CHD risk according to duration of current Vitamin E supplement use compared to no use



Rimm et al NEJM 1993; 328: 1450-6

May 20, 1993

#### Vitamin E Greatly Reduces Risk Of Heart Disease, Studies Suggest

#### By JANE E. BRODY

Two new studies of more than 120,000 men and women strongly suggest that supplements of vitamin E can significantly reduce the risk of dise researchers and other experts cautioned against rushing out to buy the vitamin supplements before further clinical trials confirm that they are be

The studies, by researchers at the Harvard School of Public Health and Brigham and Women's Hospital in Boston, showed that initially healthy coronary disease at a rate about 40 percent lower than comparable men and women whose intake of this vitamin was lowest. The preventive e blood levels of cholesterol.

The greatest protection was found at levels of about 100 international units of vitamin E a day for more than two years. The Federal recommer consume fewer than 25 units from foods like vegetable oils, wheat germ, seeds, whole grains and nuts.

The researchers said vitamin E, as an antioxidant, might reduce heart disease by having an effect on low-density lipoprotein cholesterol, or LDI type of cholesterol damages arteries primarily after it has been oxidized.

The new findings, which appear today in The New England Journal of Medicine, are some of the first to find health benefits from taking large-of "megadoses" of vitamins as a popular remedy whose value is unproven. Expert Urge Caution

While a person might conclude from the findings that it would be wise to take large doses of vitamin E supplements daily, their long-term safety

The average American lifespan has increased nearly 3 years over the last 2 decades.\*

> We've been selling vitamins at a discount since 1977.

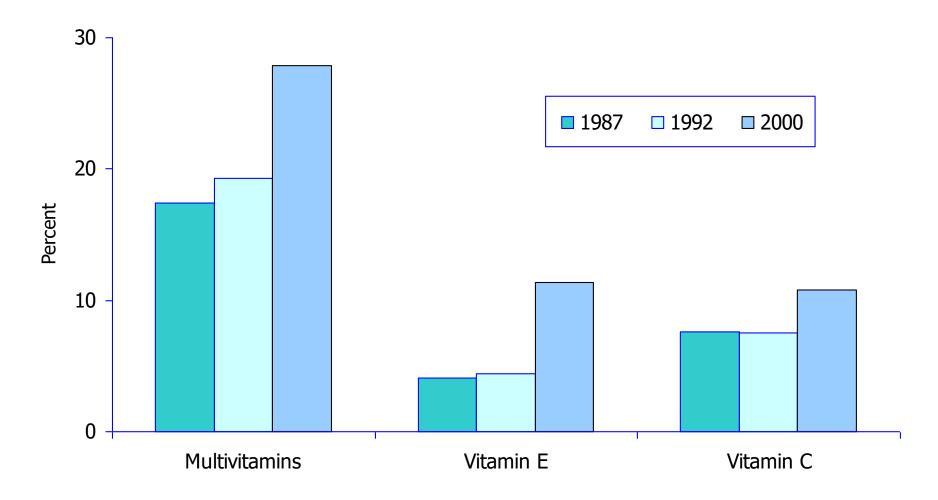
#### Coincidence? We don't think so.

At VitaminShoppe<sup>\*</sup>com we see vitamins as an essential part of a healthy life - not a luxury. And our pricing reflects that philosophy. Right now we are taking 40% off every item we stock. After 23 years in the vitamin business, we've learned how to assemble the finest vitamins, minerals, and supplements at the lowest prices...all 18,000 of them.

VitaminShoppe.com

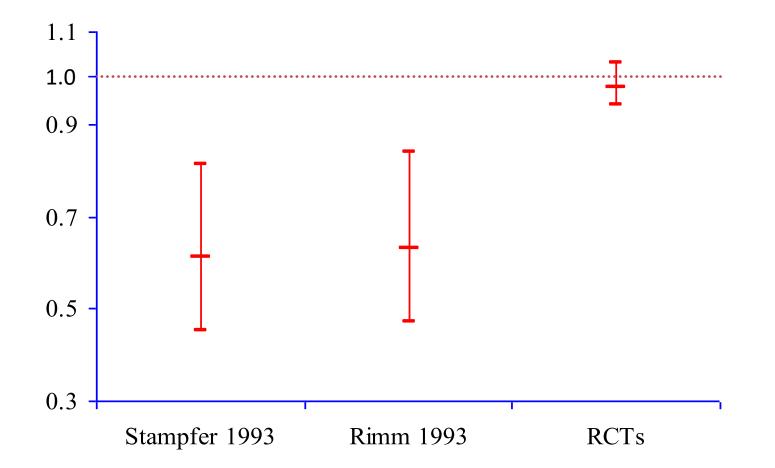
We take vitamins seriously.

# Use of vitamin supplements by US adults, 1987-2000

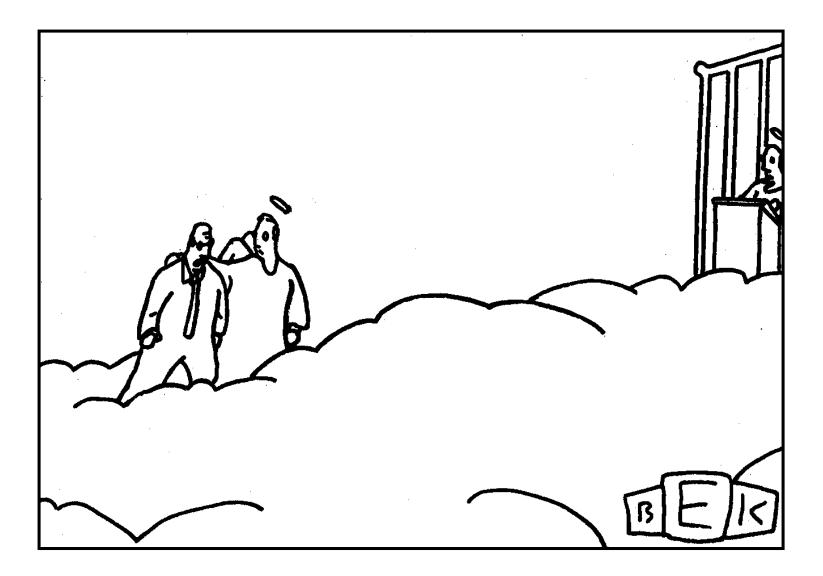


Source: Millen AE, Journal of American Dietetic Assoc 2004;104:942-950

#### Vitamin E supplement use and risk of Coronary Heart Disease



Stampfer et al NEJM 1993; 328: 144-9; Rimm et al NEJM 1993; 328: 1450-6; Eidelman et al Arch Intern Med 2004; 164:1552-6



"Well, so much for antioxidants."

#### **MANY OTHER EXAMPLES**

VITAMIN C, VITAMIN A, HRT, MANY DRUG TARGETS......

#### WHAT'S THE EXPLANATION?

### Vitamin E levels and confounding risk factors:

Childhood SES

Manual social class

No car access

State pension only

Smoker

Obese

Daily alcohol

Exercise

Low fat diet

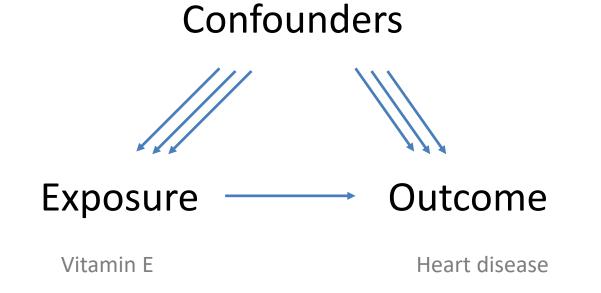
Height

Leg length

Women's Heart and Health Study Lawlor et al, Lancet 2004

## Confounding

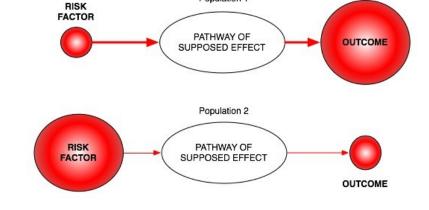
Smoking, diet, alcohol, socioeconomic position....



Classic limitations to "observational" science

• Confounding

Reverse Causation



Population 1

• Bias

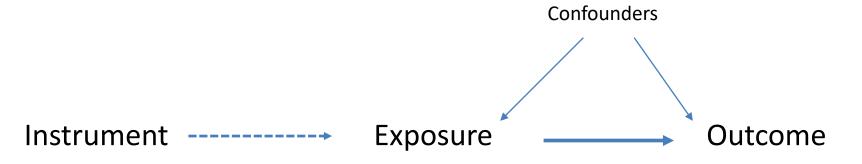
### Mendelian randomization



#### How can it help observational epidemiology?

### What is Mendelian randomization?

- Mendelian randomization (MR) is an epidemiological technique that uses genetic variants as proxy measures for an environmental exposure
- It is an application of "Instrumental Variable" (IV) analysis:
  - it uses genetic variants as 'instruments' for the exposure of interest.
     An IV is a variable that is only associated with an outcome because of its association with the exposure



### What does MR do?

- Assess causal relationship between two variables
- Estimate magnitude of causal effect

How does it do this? By harnessing Mendel's laws of inheritance

### **Mendel's Laws of Inheritance**

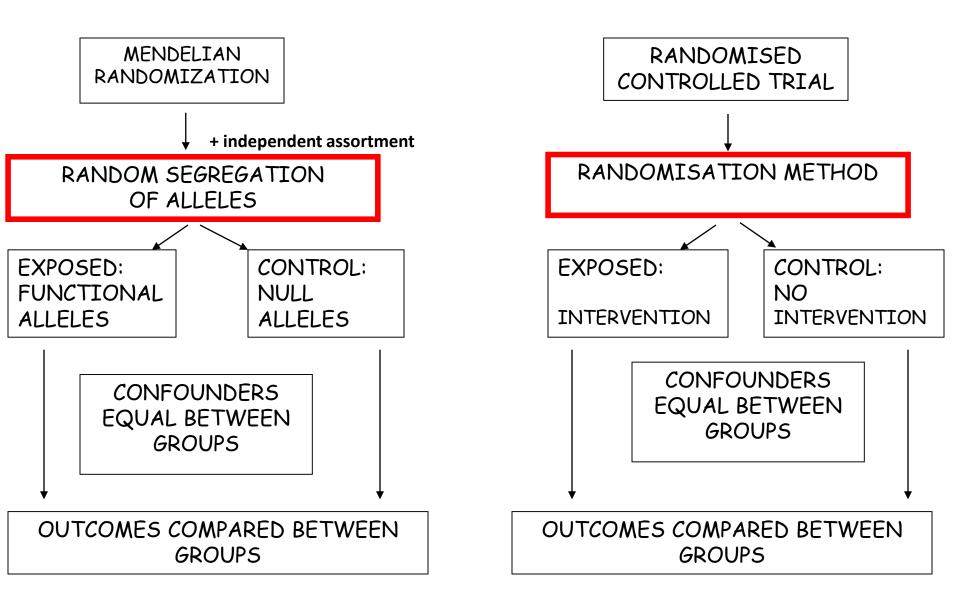


Mendel in 1862

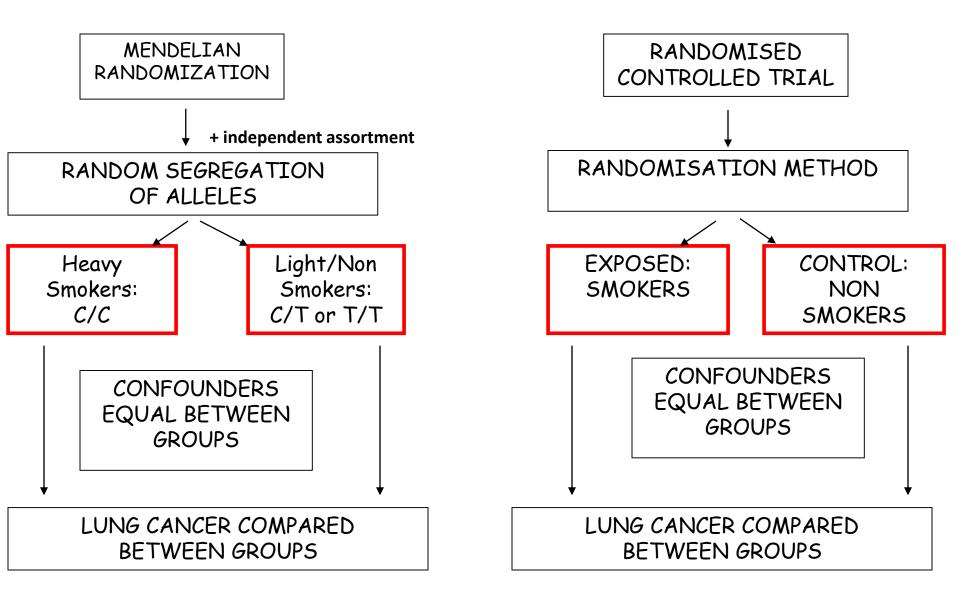
**1. Segregation:** alleles separate at meiosis and a randomly selected allele is transmitted to offspring

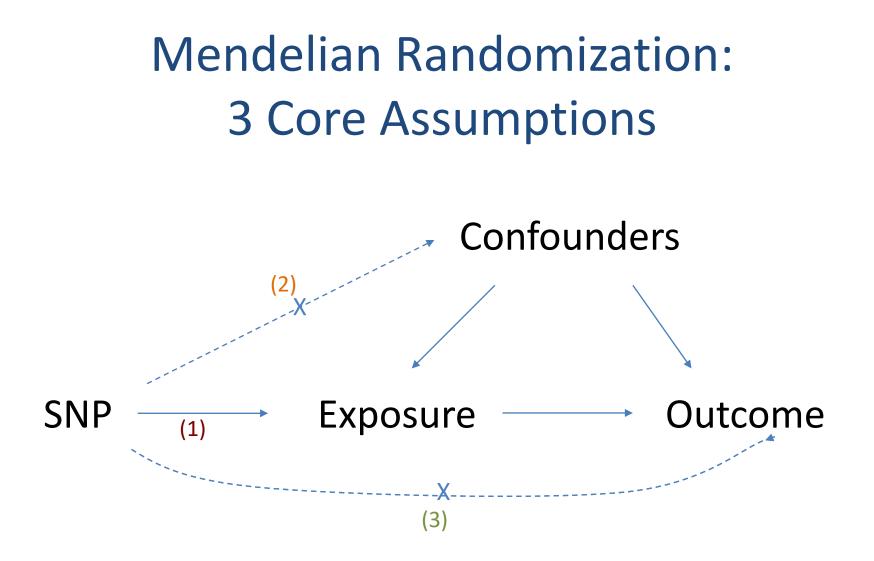
**2. Independent assortment:** alleles for separate traits are transmitted independently of one another

### **Mendelian randomization and RCTs**



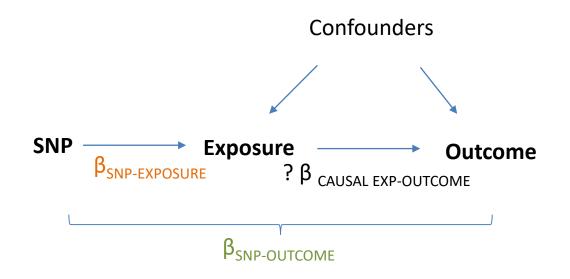
#### Mendelian randomization: Smoking and Lung Cancer





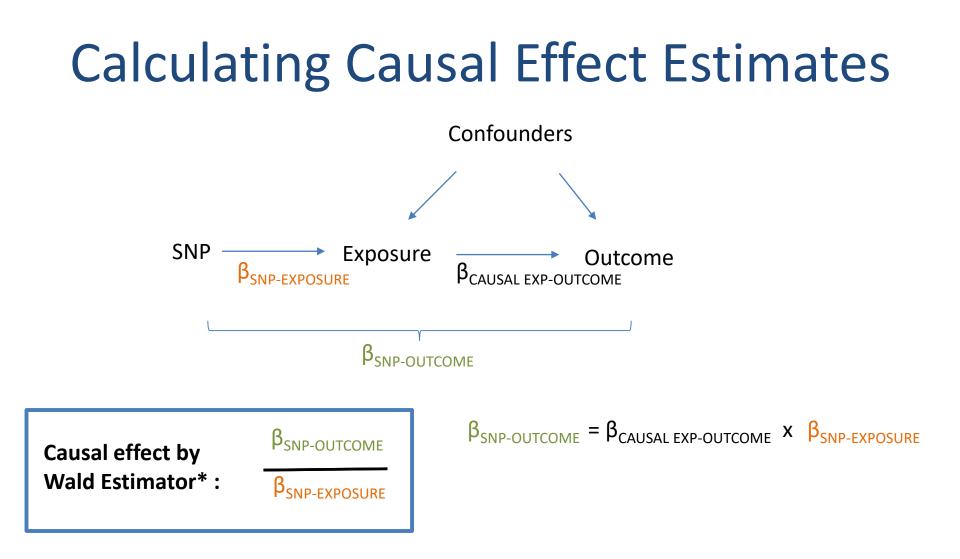
- (1) SNP is associated with the exposure
- (2) SNP is NOT associated with confounding variables
- (3) SNP ONLY associated with outcome through the exposure

### **Calculating Causal Effect Estimates**

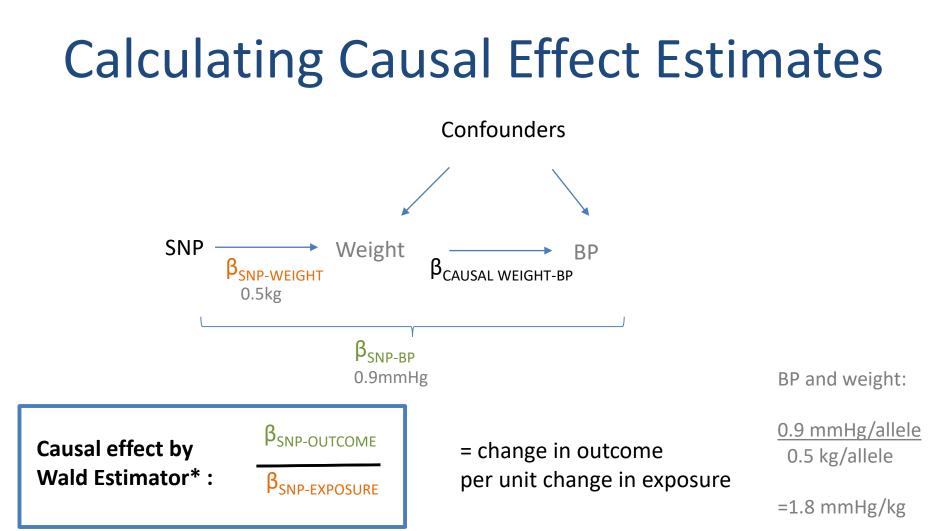


After SNP identified robustly associated with exposure of interest:

- Wald Estimator
- Two-stage least-squares (TSLS) regression

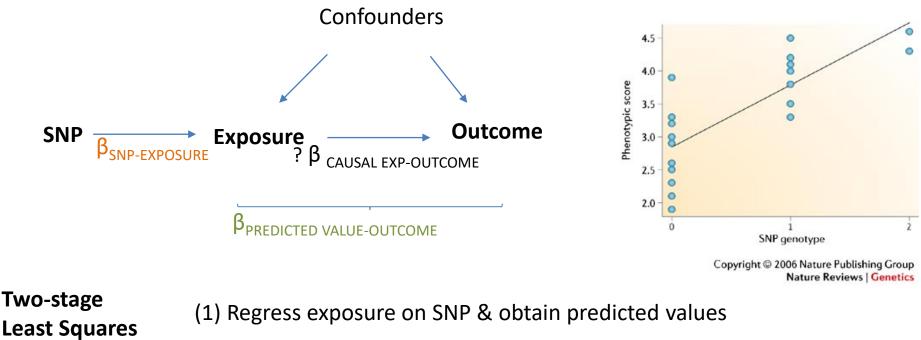


\*Can be used in different samples ("Two sample MR")



\*Can be used in different samples ("Two sample MR")

## **Calculating Causal Effect Estimates**



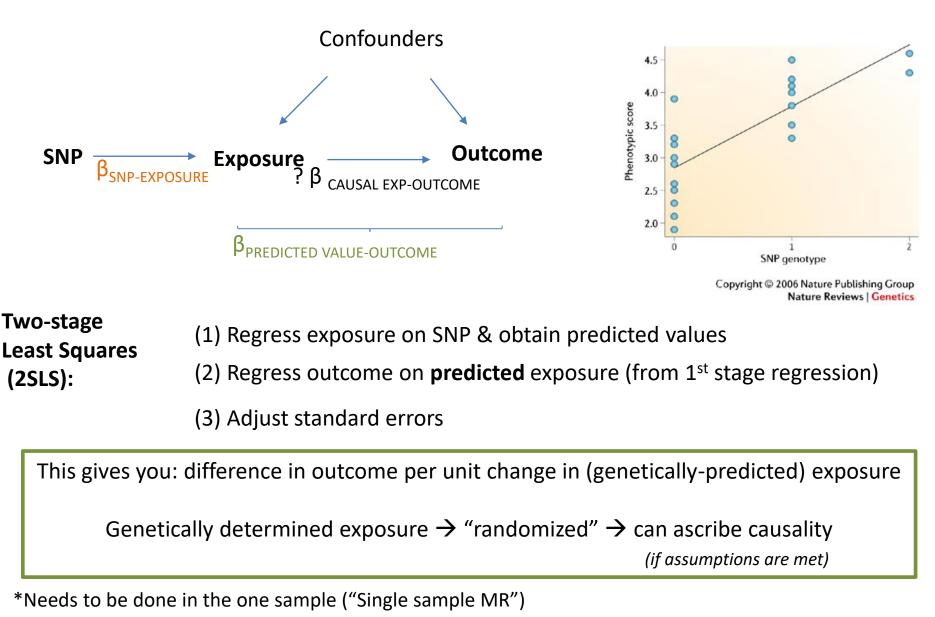
(2) Regress outcome on **predicted** exposure (from 1<sup>st</sup> stage regression)

(3) Adjust standard errors

(2SLS):

\*Needs to be done in the one sample ("Single sample MR")

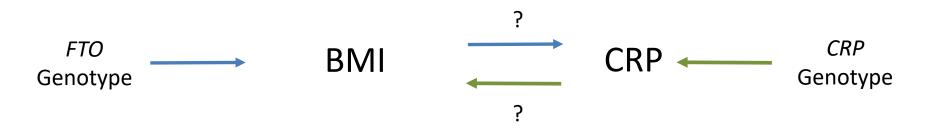
# **Calculating Causal Effect Estimates**



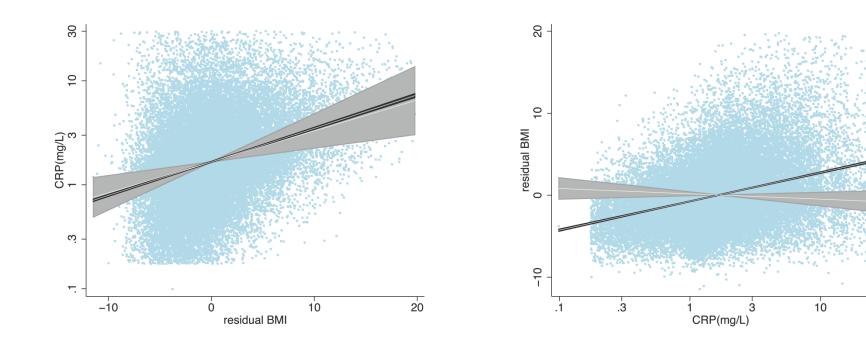
## MR Example using CRP

- C-Reactive Protein (CRP) is a biomarker of inflammation
- It is associated with BMI, metabolic syndrome, CHD and a number of other diseases
- It is unclear whether these observational relationships are causal or due to confounding or reverse causality
- This question is important from the perspective of intervention and drug development

"Bi-directional Mendelian Randomization": Testing causality and reverse causation



|   | Effect estimates        |                          |                 |                   |                |
|---|-------------------------|--------------------------|-----------------|-------------------|----------------|
| Outcome<br>/<br>explanatory<br>variable | Observational           | Instrumental<br>variable | P <sub>IV</sub> | P <sub>diff</sub> | <b>F</b> first |
| CRP/BMI                                 | 1.075<br>(1.073, 1.077) | 1.06<br>(1.02, 1.11)     | 0.002           | 0.6               | 50.2           |



### Limitations to Mendelian Randomization

- 1- Population stratification
- 2- Canalisation ("Developmental compensation")
- 3- The existence of instruments
- 4- Power (also "weak instrument bias")
- **5- Pleiotropy**

### Power and Weak Instruments

- Power:
  - Genetic variants explain very small amounts of phenotypic variance in a given trait
  - VERY large sample sizes are generally required
- Weak instruments:
  - Genetic variants that are weak proxies for the exposure
  - Results in biased causal estimates from MR
- Different impact of the bias from weak instruments:
  - Single Sample MR: to the confounded estimate
  - Two-Sample MR: to the null

#### Using Multiple Genetic Variants as Instruments

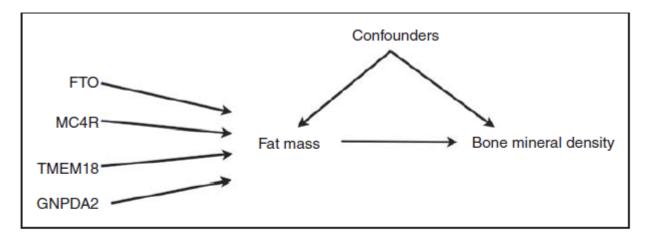


Figure 1. DAG for a Mendelian randomisation analysis using four genetic variants as instrumental variables for the effect of fat mass on bone mineral density.

Palmer et al (2011) Stat Method Res

- Allelic scores
- Testing multiple variants individually
- Meta-analyse individual SNPs



### http://cnsgenomics.com/shiny/mRnd/

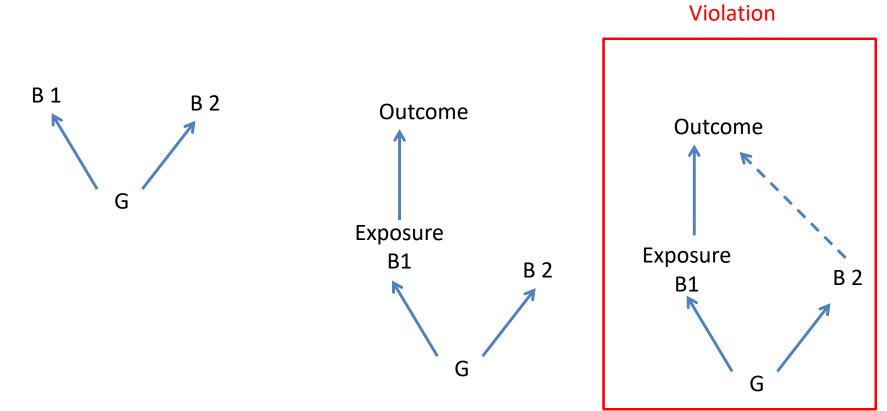
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#### mRnd: Power calculations for Mendelian Randomization

| Input                     | Continuous outcome Binary outcome derivations Citation About   |  |  |
|---------------------------|--|--|--|
| Calculate:                | Two-stage least squares  |  |  |
|                           | Power 0.05   |  |  |
| <ul> <li>Power</li> </ul> | NCP 0.00 Non-Centrality-Parameter  |  |  |
| ○ Sample size             | F-statistic 11.10 The strength of the instrument   |  |  |
| Provide:                  |  |  |  |
| Sample size               | Power or sample size calculations for two-stage least squares Mendelian Randomization studies using a genetic instrument $Z$ (a SNP                |  |  |
| 1000                      | or allele score), a continuous exposure variable X (e.g. body mass index [BMI, $\frac{kg}{m^2}$ ]) and a continuous outcome variable Y (e.g. blood |  |  |
|                           | pressure [mmHg]).  |  |  |
|                           |  |  |  |
|                           |  |  |  |
| α                         | YZ association   |  |  |
| 0.05                      |  |  |  |
| 0.05                      | Power 0.05   |  |  |
| Type-I error rate         | NCP 0.00 Non-Centrality-Parameter  |  |  |
|                           | Power or sample size calculations for the regression association of a genetic instrument $Z$ (e.g. a BMI SNP), with a continuous                   |  |  |
|                           | outcome variable $Y$ (blood pressure).   |  |  |
| $\beta_{yx}$              |  |  |  |
|                           |  |  |  |
|                           |  |  |  |

# Pleiotropy

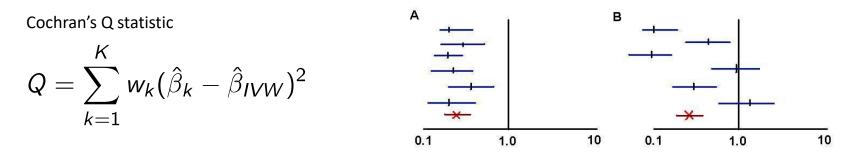
- Genetic variant influences more than one trait
- Pleiotropy only violates MR's assumptions if it involves a pathway outside that of the exposure and is a pathway that <u>affects your outcome</u>



## Tests of Heterogeneity to Identify Pleiotropy

We expect that each SNP represents an independent study, and each should give an unbiased (if imprecise) estimate of the causal effect of x on y

Heterogeneity, where effect estimates are more different than expected due to standard errors, arises because at least some of the instruments are invalid



n=6 instruments Expect Q = 5 if there is no heterogeneity Q is chi-square distributed with n-1 degrees of freedom

MR Egger regression

MR Weighted Median

**MR Modal Estimator** 

## **MR** Base



Jie "Chris" Zheng

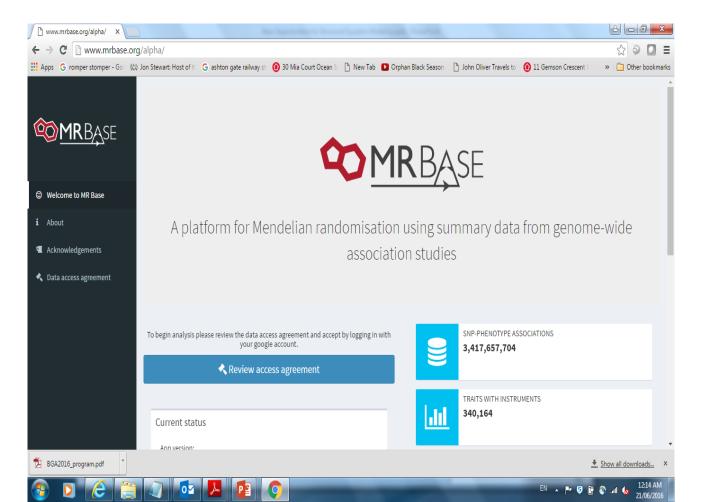
http://www.mrbase.org/





Gib Hemani

Phil Haycock



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| MR BASE   | statistics for these SNPs can be taken from a s   | URE<br>effect of an exposure on an outcome, the first step is to identify SNPs that are robustly associated with the exposure. Th<br>ample from which there is no data on the outcome.<br>one of the data sources below, or by uploading your own data. You can choose multiple exposures to be analysed, and |           | ry            |     |
| <ul> <li>Welcome to MR Base</li> <li>About</li> </ul>   | Choose instruments  | Manual file upload  |           |               |     |
| Acknowledgements  | Select exposure source<br>Manual file upload<br>NHGRI-EBI GWAS catalog  | The file must be a plain text file.<br>To do simple SNP look ups it must have at least one column with the header <b>SNP</b> .<br>To do an MR analysis it must have the following column headers:   |           |               |     |
| <ul> <li>Data access agreement</li> <li>Logged in as</li> <li>David Evans</li> <li>epxde@bristol.ac.uk</li> </ul> | <ul> <li>MR Base GWAS catalog</li> <li>Gene expression QTLs</li> <li>Protein level QTLs</li> <li>Metabolite level QTLs</li> </ul> | <ul> <li>SNP - rs IDs of the instruments for the exposure</li> <li>beta - effect sizes for each SNP</li> <li>se - standard errors</li> <li>effect_allele - Effect allele</li> </ul>   |           |               |     |
| 📽 Perform MR analysis 🛛 🗸   | Methylation level QTLs  | It's useful to have these columns too:  |           |               |     |
| 幸 Choose exposures  |   | other_allele - Other allele     eaf - Effect allele frequency   |           |               |     |
| 華 Choose outcomes<br>華 Run MR   |   | You can see an example file here: telomere_length.txt   |           |               |     |
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|   | LD clumping   | Select methods for analysis  | Submit  |
| <b>∞<u>MR</u>BA</b> SE  | Most two sample MR methods require that the instruments do not have LD between them. Linkage disequilibrium  Do not check for LD between SNPs | Many methods exist for performing two sample MR.<br>Different methods have sensitivities to different<br>potential issues, accommodate different scenarios, and<br>vary in their statistical efficiency. | Once you have selected exposures, outcomes, and analysis options you are ready to perform the analysis. |
| Welcome to MR Base  | <ul> <li>Use clumping to prune SNPs for LD</li> </ul>   | Choose which methods to use:<br>Wald ratio   | <b>9</b> Perform MR analysis  |
| i About   | LD proxies  | <ul> <li>Fixed effects meta analysis (simple SE)</li> <li>Fixed effects meta analysis (delta method)</li> </ul>  |   |
| Acknowledgements  | If a particular exposure SNP is not present in an   | <ul> <li>Random effects meta analysis (delta method)</li> <li>Maximum likelihood</li> </ul>  |   |
| 🔦 Data access agreement   | outcome dataset, should proxy SNPs be used instead through LD tagging?  | <ul> <li>MR Egger</li> <li>MR Egger (bootstrap)</li> </ul>   |   |
| Logged in as<br>David Evans                                       | Use proxies? Minimum LD Rsg value   | <ul> <li>Weighted median</li> <li>Penalised weighted median</li> </ul>   |   |
| epxde@bristol.ac.uk   | 0.6 0.8 1   | Inverse variance weighted  |   |
| 📽 Perform MR analysis 🛛 🗸   | 0.6 0.54 0.58 0.72 0.76 0.8 0.84 0.88 0.92 0.96 1   |  |   |
| <ul> <li>靠 Choose exposures</li> <li>靠 Choose outcomes</li> </ul> | Allow palindromic SNPs?   |  |   |
| 후 Run MR  | MAF threshold for aligning palindromes  |  |   |
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|                                 | 300                                      | ID<br>300 | Trait<br>LDL<br>cholesterol            | Note   | author<br>Willer CJ | Consortium<br>GLGC                               | of cases    | controls                               | size<br>173082 | variants<br>2437752 | Year<br>2013           | PubmedID<br>24097068 | Access<br>public                         | Category<br>Risk factor | Poj |   |
|                                 | 781                                      | 781       | LDL<br>cholesterol                     | Metabo-<br>chip                                      | Willer CJ           | GLGC   |             |  | 83198          | 120251              | 2013                   | 24097068             | public                                   | Risk factor             |     |   |
|                                 | 880                                      | 880       | Total<br>cholesterol in<br>large LDL   | L.LDL.C  | Kettunen            |  |             |  | 21552          | 11871461            | 2016                   | 27005778             | public                                   | Metabolites             |     |   |
|                                 | 881                                      | 881       | Cholesterol<br>esters in large<br>VLDL | L.LDL.CE   | Kettunen            |  |             |  | 19273          | 11820655            | 2016                   | 27005778             | public                                   | Metabolites             | -   |   |
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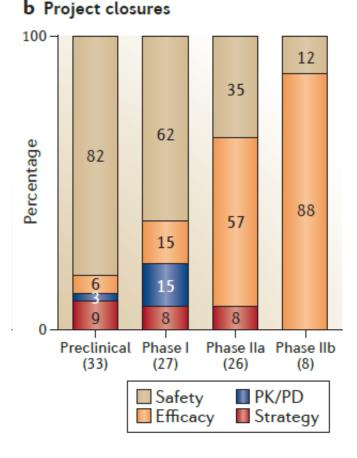
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| ← → C ① www.mrbase.org/b   | peta/?state=3DNYEn5uYpGOFV1AR1x8&code=4/c_PBiWWg   |  | ☆ ❷ ◘ :                                   |
| Welcome to MR Base   | Linkage disequilibrium <ul> <li>Do not check for LD between SNPs</li> <li>Use clumping to prune SNPs for LD</li> </ul>   | potential issues, accommodate different scenarios, and<br>vary in their statistical efficiency.<br>Choose which methods to use:<br>Wald ratio<br>Fixed effects meta analysis (simple SE)   | Perform MR analysis                       |
| <ul> <li>About</li> <li>Acknowledgements</li> <li>Data access agreement</li> <li>Logged in as</li> <li>David Evans</li> <li>epxde@bristol.ac.uk</li> </ul> | LD proxies<br>If a particular exposure SNP is not present in an<br>outcome dataset, should proxy SNPs be used instead<br>through LD tagging?<br>Use proxies?   | <ul> <li>Fixed effects meta analysis (simple SE)</li> <li>Fixed effects meta analysis (delta method)</li> <li>Random effects meta analysis (delta method)</li> <li>Maximum likelihood</li> <li>MR Egger</li> <li>MR Egger (bootstrap)</li> <li>Weighted median</li> <li>Penalised weighted median</li> </ul> |   |
| Perform MR analysis <  | Allele harmonisation<br>An important step in two sample MR is making sure<br>that the effects of the SNPs on the exposure<br>correspond to the same allele as their effects on the<br>outcome. This is potentially difficult with palindromic<br>SNPs. | Inverse variance weighted  |   |
| <b>Q</b> Quick SNP lookup <  | <ul> <li>Handling reference alleles</li> <li>All effect alleles are definitely on the positive strand</li> <li>Attempt to align strands for palindromic SNPs</li> <li>Exclude palindromic SNPs</li> </ul>  |  |   |
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## Mendelian Randomization and Drug Targets

Thanks Sek Kathiresan

## Late Stage Failure in Drug Trials

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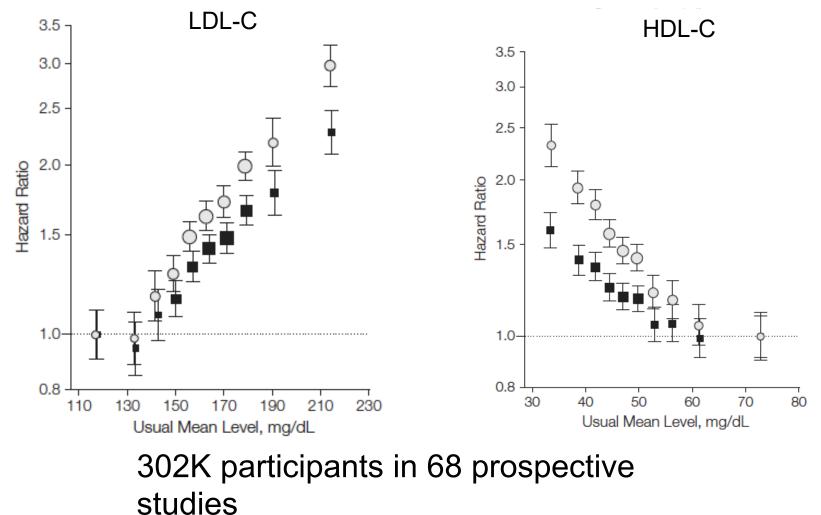
### 100 18 27 Ratio of projects (%) 57 71 82 73 43 29 0 Yes No Yes No (21) (15)(17)(7) Projects with human genetic Projects with efficacy linkage of the target to biomarkers available the disease indication at start of phase Closed Active or successful

Phase II projects

Cook et al. (2014) Nat Rev Drug Disc

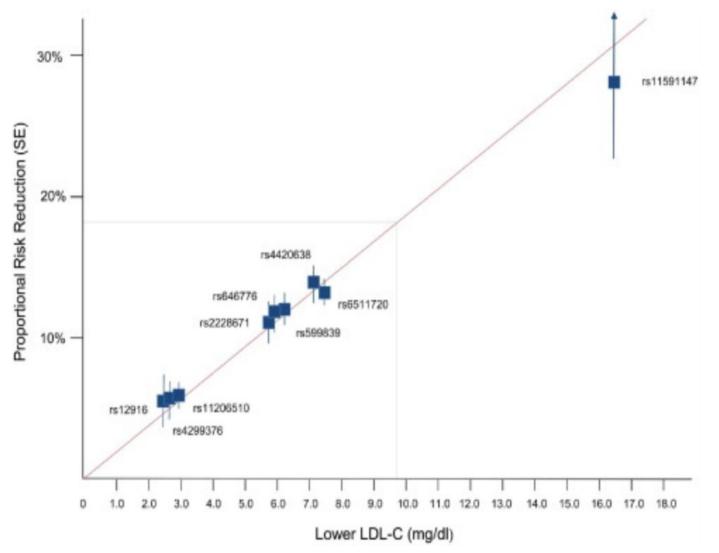
Phase IIa projects

# Association of LDL-C, HDL-C, and risk for coronary heart disease (CHD)



Emerging Risk Factors Collaboration, JAMA 2009

## LDL and CHD Risk



Ference et al, JACC 2012

## HDL: endothelial lipase Asn396Ser

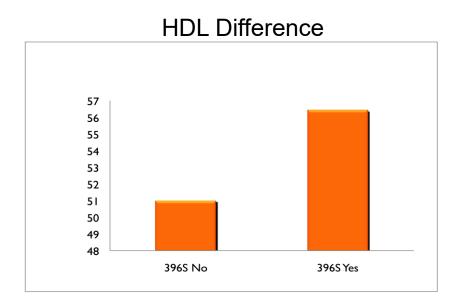
### Loss-of-function variants in endothelial lipase are a cause of elevated HDL cholesterol in humans

Andrew C. Edmondson,<sup>1</sup> Robert J. Brown,<sup>1</sup> Sekar Kathiresan,<sup>2,3</sup> L. Adrienne Cupples,<sup>4</sup> Serkalem Demissie,<sup>4</sup> Alisa Knodle Manning,<sup>4</sup> Majken K. Jensen,<sup>5</sup> Eric B. Rimm,<sup>5,6</sup> Jian Wang,<sup>7</sup> Amrith Rodrigues,<sup>1</sup> Vaneeta Bamba,<sup>1</sup> Sumeet A. Khetarpal,<sup>1</sup> Megan L. Wolfe,<sup>1</sup> Stephanie DerOhannessian,<sup>1</sup> Mingyao Li,<sup>8</sup> Muredach P. Reilly,<sup>1,9</sup> Jens Aberle,<sup>10</sup> David Evans,<sup>10</sup> Robert A. Hegele,<sup>7</sup> and Daniel J. Rader<sup>1,9</sup>

- 2.6% of population carry Serine allele
- higher HDL-C
- No effect on other lipid fractions
- No effect on other MI risk factors

Edmondson, *J Clin Invest* 2009

## LIPG N396S and plasma HDL-C



## 396S carriers have 5.5 mg/dl higher HDL-C P<10<sup>-8</sup>

# After testing in 116,320 people, summary OR for *LIPG* Asn396Ser is 0.99

|                          | Number of individuals |          |  | OR (95% CI)      |
|--------------------------|-----------------------|----------|--|------------------|
|                          | Cases                 | Controls |  |                  |
| AngioGOCARD/KORA         | 1953                  | 1482     |  | 0.76 (0.46–1.24) |
| IFS                      | 577                   | 719      |  | 1.25 (0.51-3.08) |
| deCODE                   | 729                   | 29218    | <b>_</b>   | 0.97 (0.60–1.58) |
| EPIC-NL                  | 334                   | 1827     |  | 0.64 (0.27-1.53) |
| GerMIFS-II               | 1127                  | 1874     |  | 1.36 (0.82-2.24) |
| GRACE                    | 683                   | 656      | +  | 2.48 (1.10-5.56) |
| MAHA                     | 785                   | 615      | <b></b>  | 1.08 (0.68-1.72) |
| PennCATH                 | 485                   | 489      |  | 0.82 (0.37-1.83) |
| UCP                      | 830                   | 1139     |  | 0.87 (0.44-1.72) |
| POPGEN                   | 2433                  | 1687     | _ <b>_</b>   | 0.69 (0.42-1.14) |
| PROCARDIS                | 2183                  | 3347     |  | 0.66 (0.45-0.98) |
| PROMIS                   | 1854                  | 1897     |  | 1.27 (0.74-2.16) |
| SHEEP                    | 1151                  | 1496     |  | 1.35 (0.85-2.14) |
| WTCCC                    | 1561                  | 2426     |  | 0.74 (0.49-1.12) |
| All case-control studies | 16685                 | 48872    | <b>—</b>   | 0.94 (0.82-1.09) |
| ARIC                     | 558                   | 8214     | <b>_</b> _   | 0.80 (0.45-1.40) |
| CCHS                     | 655                   | 8964     |  | 1.33 (0.73-2.43) |
| DCH                      | 933                   | 1588     |  | 1.12 (0.66-1.90) |
| FHS                      | 50                    | 1462     |  | 2.35 (0.69-8.00) |
| HPFS                     | 426                   | 869      | +  | 1.97 (0.86-4.51) |
| MDC                      | 1606                  | 25438    |  | 1.01 (0.74-1.38) |
| All cohort studies       | 4228                  | 46535    | <u> </u>   | 1.10 (0.89-1.37) |
| Overall                  | 20913                 | 95407    | A state of the | 0.99 (0.88-1.11) |

Individuals who carry the HDL-boosting variant have the same risk for heart attack as those who do not carry the variant

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### Effects of Torcetrapib in Patients at High Risk for Coronary Events

Philip J. Barter, M.D., Ph.D., Mark Caulfield, M.D., M.B., B.S., Mats Eriksson, M.D., Ph.D., Scott M. Grundy, M.D., Ph.D., John J.P. Kastelein, M.D., Ph.D., Michel Komajda, M.D., Jose Lopez-Sendon, M.D., Ph.D., Lori Mosca, M.D., M.P.H., Ph.D., Jean-Claude Tardif, M.D., David D. Waters, M.D., Charles L. Shear, Dr.P.H., James H. Revkin, M.D., Kevin A. Buhr, Ph.D., Marian R. Fisher, Ph.D., Alan R. Tall, M.B., B.S., and Bryan Brewer, M.D., Ph.D., for the ILLUMINATE Investigators\*

### RESULTS

protein cholesterol, as compared with baseline (P<0.001 for both comparisons), in addition to an increase of 5.4 mm Hg in systolic blood pressure, a decrease in serum potassium, and increases in serum sodium, bicarbonate, and aldosterone (P<0.001

showed an increased risk of death in patients treated with torcetrapib whose reduction in potassium or increase in bicarbonate was greater than the median change. ORIGINAL ARTICLE

## Effects of Dalcetrapib in Patients with a Recent Acute Coronary Syndrome

Gregory G. Schwartz, M.D., Ph.D., Anders G. Olsson, M.D., Ph.D., Markus Abt, Ph.D., Christie M. Ballantyne, M.D., Philip J. Barter, M.D., Ph.D., Jochen Brumm, Ph.D., Bernard R. Chaitman, M.D., Ingar M. Holme, Ph.D., David Kallend, M.B., B.S., Lawrence A. Leiter, M.D., Eran Leitersdorf, M.D., John J.V. McMurray, M.D., Hardi Mundl, M.D., Stephen J. Nicholls, M.B., B.S., Ph.D., Prediman K. Shah, M.D., Jean-Claude Tardif, M.D., and R. Scott Wright, M.D., for the dal-OUTCOMES Investigators\*

### RESULTS

At the time of randomization, the mean HDL cholesterol level was 42 mg per deciliter (1.1 mmol per liter), and the mean low-density lipoprotein (LDL) cholesterol

interim analysis that included 1135 primary end-point events (71% of the projected total number), the independent data and safety monitoring board recommended

per liter higher and the mean systolic blood pressure was 0.6 mm Hg higher with dalcetrapib as compared with placebo (P<0.001 for both comparisons).

## HDL—is it too big to fail?

### Dominic S. Ng, Norman C. W. Wong and Robert A. Hegele

Abstract | The HDL hypothesis has suffered damage in the past few years. Clinical trials have shown that raising HDL cholesterol levels does not improve cardiovascular disease (CVD) outcomes. In addition, Mendelian randomization studies have shown that DNA variants that alter HDL cholesterol levels in populations are unrelated to incident CVD events. Balancing this deluge of negative data are substantial basic science data supporting the concept that raising HDL cholesterol levels reduces CVD risk. Also, functionally relevant HDL subfractions might be more important determinants of risk than overall HDL cholesterol levels. But, while wobbly, the HDL hypothesis is still standing, seemingly too big to fail owing to past intellectual, economic and psychological investments in the idea.

Ng, D. S. et al. Nat. Rev. Endocrinol. 9, 308–312 (2013); published online 15 January 2013; doi:10.1038/nrendo.2012.238

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## Practical