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RESEARCH ARTICLE

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Phenome-wide association analysis of LDL-cholesterol lowering genetic variants in *PCSK9*

Amand F. Schmidt^{1,2,3*} , Michael V. Holmes^{4†}, David Preiss^{4†}, Daniel I. Swerdlow^{1,5}, Spiros Denaxas^{3,6,7,8}, Ghazaleh Fatemifar^{3,6,7}, Rupert Faraway¹, Chris Finan^{1,3}, Dennis Valentine^{3,9}, Zammy Fairhurst-Hunter¹⁰, Fernando Pires Hartwig¹¹, Bernardo Lessa Horta¹¹, Elina Hypponen^{12,13,14}, Christine Power¹³, Max Moldovan¹⁴, Erik van Iperen^{15,16}, Kees Hovingh¹⁷, Ilja Demuth^{18,19}, Kristina Norman^{20,21,22}, Elisabeth Steinhagen-Thiessen¹⁸, Juri Demuth²³, Lars Bertram^{24,25}, Christina M. Lill^{26,27,28}, Stefan Coassin²⁹, Johann Willeit³⁰, Stefan Kiechl³⁰, Karin Willeit^{30,31}, Dan Mason³², John Wright³², Richard Morris³³, Goya Wanamethee³³, Peter Whincup³⁴, Yoav Ben-Shlomo³⁵, Stela McLachlan³⁶, Jackie F. Price³⁶, Mika Kivimaki³⁷, Catherine Welch³⁷, Adelaida Sanchez-Galvez³⁷, Pedro Marques-Vidal³⁸, Andrew Nicolaides^{39,40}, Andrie G. Panayiotou⁴¹, N. Charlotte Onland-Moret⁴², Yvonne T. van der Schouw⁴², Giuseppe Matullo^{43,44}, Giovanni Fiorito^{43,44}, Simonetta Guarrera^{43,44}, Carlotta Sacerdote⁴⁵, Nicholas J. Wareham⁴⁶, Claudia Langenberg⁴⁶, Robert A. Scott⁴⁶, Jian'an Luan⁴⁶, Martin Bobak³⁷, Sofia Malyutina^{47,48}, Andrzej Pajak⁴⁹, Ruzena Kubinova⁵⁰, Abdonas Tamosiunas⁵¹, Hynek Pikhart³⁷, Niels Grarup⁵², Oluf Pedersen⁵², Torben Hansen⁵², Allan Linneberg^{53,54}, Tine Jess⁵⁴, Jackie Cooper⁵⁵, Steve E. Humphries⁵⁵, Murray Brilliant⁵⁶, Terrie Kitchner⁵⁶, Hakon Hakonarson⁵⁷, David S. Carrell⁶⁰, Catherine A. McCarty⁵⁸, Kirchner H. Lester⁵⁹, Eric B. Larson⁶⁰, David R. Crosslin⁶¹, Mariza de Andrade⁶², Dan M. Roden⁶³, Joshua C. Denny⁶⁴, Cara Carty⁶⁵, Stephen Hancock⁶⁶, John Attia^{66,67}, Elizabeth Holliday^{66,67}, Rodney Scott⁶⁶, Peter Schofield⁶⁸, Martin O'Donnell⁶⁹, Salim Yusuf⁶⁹, Michael Chong⁶⁹, Guillaume Pare⁶⁹, Pim van der Harst^{15,16,70,71}, M. Abdullah Said⁷¹, Ruben N. Eppinga⁷¹, Niek Verweij⁷¹, Harold Snieder⁷², Lifelines Cohort authors, Tim Christen⁷³, D. O. Mook-Kanamori⁷³, the ICBP Consortium, Stefan Gustafsson⁷⁴, Lars Lind^{75,76}, Erik Ingelsson^{74,75}, Raha Pazoki^{77,78}, Oscar Franco⁷⁷, Albert Hofman⁷⁷, Andre Uitterlinden⁷⁷, Abbas Dehghan⁷⁷, Alexander Teumer^{79,80}, Sebastian Baumeister^{79,81}, Marcus Dörr^{80,82}, Markus M. Lerch⁸³, Uwe Völker^{80,84}, Henry Völzke^{79,80}, Joey Ward⁸⁵, Jill P. Pell⁸⁵, Tom Meade⁸⁶, Ingrid E. Christophersen⁸⁷, Anke H. Maitland-van der Zee^{88,89}, Ekaterina V. Baranova⁹⁰, Robin Young⁹⁰, Ian Ford⁹⁰, Archie Campbell⁹¹, Sandosh Padmanabhan⁹², Michiel L. Bots⁴¹, Diederick E. Grobbee⁴¹, Philippe Froguel^{93,94}, Dorothée Thuillier⁹³, Ronan Roussel^{95,96,97}, Amélie Bonnefond⁹³, Bertrand Cariou⁹⁸, Melissa Smart⁹⁹, Yanchun Bao¹⁰⁰, Meena Kumari¹⁰¹, Anubha Mahajan¹⁰⁰, Jemma C. Hopewell¹⁰, Sudha Seshadri¹⁰¹, the METASTROKE Consortium of the ISGC, Caroline Dale⁹, Rui Providencia E. Costa⁹, Paul M. Ridker¹⁰², Daniel I. Chasman¹⁰², Alex P. Reiner¹⁰³, Marylyn D. Ritchie¹⁰⁴, Leslie A. Lange¹⁰⁵, Alex J. Cornish¹⁰⁶, Sara E. Dobbins¹⁰⁶, Kari Hemminki^{107,108}, Ben Kinnersley¹⁰⁶, Marc Sanson^{109,110}, Karim Labreche^{109,110}, Matthias Simon¹¹¹, Melissa Bondy¹¹², Philip Law¹⁰⁶

* Correspondence: amand.schmidt@ucl.ac.uk

[†]Amand F Schmidt, Michael V Holmes and David Preiss are Joint first authors

[†]Naveed Sattar, Richard Houlston, Juan P Casas and Aroon D Hingorani are Joint last authors

¹Institute of Cardiovascular Science, University College London, 222 Euston Road, London NW1 2DA, UK

²Department of Cardiology, Division Heart & Lungs, University Medical Center Utrecht, Utrecht University, Utrecht, the Netherlands



Helen Speedy¹⁰⁶, James Allan¹¹³, Ni Li¹⁰⁶, Molly Went¹⁰⁶, Niels Weinhold¹¹⁴, Gareth Morgan¹¹⁴, Pieter Sonneveld¹¹⁵, Björn Nilsson¹¹⁶, Hartmut Goldschmidt¹¹⁷, Amit Sud¹⁰⁶, Andreas Engert¹¹⁸, Markus Hansson^{119,120}, Harry Hemingway^{3,6,7,121}, Folkert W. Asselbergs^{1,2,3,122}, Riyaz S. Patel^{1,3,123}, Brendan J. Keating¹²⁴, Naveed Sattar⁹²⁺, Richard Houlston¹⁰⁶⁺, Juan P. Casas¹²⁵⁺ and Aroon D. Hingorani^{1,3+}

Abstract

Background: We characterised the phenotypic consequence of genetic variation at the *PCSK9* locus and compared findings with recent trials of pharmacological inhibitors of PCSK9.

Methods: Published and individual participant level data (300,000+ participants) were combined to construct a weighted *PCSK9* gene-centric score (GS). Seventeen randomized placebo controlled PCSK9 inhibitor trials were included, providing data on 79,578 participants. Results were scaled to a one mmol/L lower LDL-C concentration.

Results: The *PCSK9* GS (comprising 4 SNPs) associations with plasma lipid and apolipoprotein levels were consistent in direction with treatment effects. The GS odds ratio (OR) for myocardial infarction (MI) was 0.53 (95% CI 0.42; 0.68), compared to a PCSK9 inhibitor effect of 0.90 (95% CI 0.86; 0.93). For ischemic stroke ORs were 0.84 (95% CI 0.57; 1.22) for the GS, compared to 0.85 (95% CI 0.78; 0.93) in the drug trials. ORs with type 2 diabetes mellitus (T2DM) were 1.29 (95% CI 1.11; 1.50) for the GS, as compared to 1.00 (95% CI 0.96; 1.04) for incident T2DM in PCSK9 inhibitor trials. No genetic associations were observed for cancer, heart failure, atrial fibrillation, chronic obstructive pulmonary disease, or Alzheimer's disease – outcomes for which large-scale trial data were unavailable.

Conclusions: Genetic variation at the *PCSK9* locus recapitulates the effects of therapeutic inhibition of PCSK9 on major blood lipid fractions and MI. While indicating an increased risk of T2DM, no other possible safety concerns were shown; although precision was moderate.

Keywords: Genetic association studies, Mendelian randomisation, LDL-cholesterol, Phenome-wide association scan

Background

Statins and ezetimibe reduce the risk of major coronary events and ischemic stroke via lowering of low density lipoprotein-cholesterol (LDL-C) [1–3]. Loss-of-function mutations in *PCSK9* are associated with lower LDL-C and a reduced risk of coronary heart disease (CHD) [4, 5]. Antibodies (mAbs) inhibiting PCSK9, reduce LDL-C in patients with hypercholesterolaemia, and received market access in 2015. The FOURIER and ODYSSEY OUTCOMES trials tested the efficacy of PCSK9-inhibition versus placebo on the background of statin treatment and both found that PCSK9 inhibition led to a 15% relative risk reduction of major vascular events in patients with established CVD and recent acute coronary syndrome over a median follow up of 2.2 to 2.8 years [6, 7].

Evidence is limited on the effect of PCSK9 inhibition on clinical outcomes, and on safety outcomes that might only become apparent with prolonged use. Nor is evidence available on the efficacy and safety of PCSK9 inhibitors in subjects other than the high-risk patients studied in trials. Mendelian randomisation for target validation uses naturally-occurring variation in a gene encoding a drug target to identify mechanism-based consequences of pharmacological modification of the same target [8]. Such studies have previously proved useful in predicting success and failure in clinical trials and have assisted in delineating on-target from off-target actions of first-in-class drugs [9–

13]. For example, previous studies showed that variants in *HMGCR*, encoding the target for statins, were associated with lower concentrations of LDL-C and lower risk of coronary heart disease [9] (CHD), while confirming the on-target nature of the effect of statins on higher body weight and higher risk of type 2 diabetes (T2DM) [9].

We characterised the phenotypic consequences of genetic variation at *PCSK9* in a large, general population sample focussing on therapeutically relevant biomarkers, cardiovascular disease (CVD), individual CVD components and non-CVD outcomes such as cancer, Alzheimer's disease, and chronic obstructive pulmonary disease (COPD). Effect estimates from the genetic analysis were compared to those from intervention trials where the outcomes under evaluation overlapped.

Methods

We summarise methods briefly here as they have been previously described in detail [14].

Genetic variant selection

SNPs rs11583680 (minor allele frequency [MAF] = 0.14), rs11591147 (MAF = 0.01), rs2479409 (MAF = 0.36) and rs11206510 (MAF = 0.17) were selected as genetic instruments at the *PCSK9* locus based on the following criteria: (1) an LDL-C association as reported by the Global Lipids Genetics Consortium (GLGC) [15]; (2) low pairwise linkage

disequilibrium (LD) ($r^2 \leq 0.30$) with other SNPs in the region (based on 1000 Genomes CEU data); and (3) the combined annotation dependent depletion (CADD) score [16] which assesses potential functionality (see Additional file 1: Table S1).

Previously, we explored the between-SNP correlations (see Additional file 2: Figure S1 of Schmidt et al. 2017 [14]), revealing an r^2 of 0.26 between rs11206510 and rs11583680, confirming all other SNPs were approximately independent ($r^2 \leq 0.07$). Subsequent adjustment for the residual LD (correlation) structure did not impact results (see Appendix Figure 90 of Schmidt et al. 2017 [14]).

Individual participant-level and summary-level data

Participating studies (Additional file 1: Table S2) provided analyses of individual participant-level data (IPD) based on a common analysis script (available from AFS), submitting summary estimates to the UCL analysis centre. These data were supplemented with public domain data from relevant genetic consortia (Additional file 1: Table S3). Studies contributing summary estimates to genetic consortia were excluded from the IPD component of the analysis to avoid duplication.

Biomarker data were collected on the major routinely measured blood lipids (LDL-C, HDL-C, triglycerides [TG], total cholesterol [TC]); apolipoproteins A1 [ApoA1] and B [ApoB], and nominal lipoprotein (Lp)(a); systolic (SBP) and diastolic (DBP) blood pressure; inflammation markers C-reactive protein (CRP), interleukin-6 (IL-6), and fibrinogen; haemoglobin; glycated haemoglobin (HbA_{1c}); liver enzymes gamma-glutamyltransferase (GGT), alanine aminotransferase (ALT), aspartate transaminase (AST), and alkaline phosphatase (ALP); serum creatinine, and cognitive function (standardized to mean 0, and standard deviation 1, see Additional file 1: Table S5).

We focussed on individual clinical endpoints, rather than composites, which have been assessed in outcome trials, as well as disease end-points commonly seen in patients likely to be eligible for PCSK9 inhibitor treatment. Ischemic CVD endpoints studied were myocardial infarction (MI), ischemic stroke, revascularization, and angina. The following non-ischemic CVD events were considered: haemorrhagic stroke, heart failure, and atrial fibrillation. Non-CVD outcome data was collected on common chronic diseases: COPD, any cancer (including those of the breast, prostate, colon and lung), Alzheimer's disease, and T2DM. Study endpoints and biomarker were chosen based on a combination of 1) available sample size, 2) clinical relevance, and 3) evaluation in RCTs of PCSK9 inhibition, we did not a priori hypothesize on the likelihood of PCSK9 being associated with any of the available phenotypes. Specific cancer sites evaluated here: chronic lymphocytic leukaemia, multiple myeloma, Hodgkin, meningioma,

glioma, melanoma, colorectal cancer, prostate cancer, breast cancer, lung adenocarcinoma, and small-cell lung cancer.

Finally, aggregated trial data on the effect of monoclonal PCSK9 (13 alirocumab trials, and 4 evolocumab trials) inhibitors were compared to placebo for MI, revascularization, ischemic or haemorrhagic stroke, cancer, and T2DM abstracted from the Cochrane systematic review [6, 17], with the addition of the OUTCOMES alirocumab trial published afterwards [18]. We compared effects on biomarkers and clinical endpoints common to both the genetic analysis and trials.

Statistical analyses

In all analyses, we assumed an additive allelic effect with genotypes coded as 0, 1 and 2, corresponding to the number of LDL-C lowering alleles; model comparison tests did not show signs of non-additivity [14]. Continuous biomarkers were analysed using linear regression and binary endpoints using logistic regression. Study-specific associations were pooled for each SNP using the inverse variance weighted method for fixed effect meta-analysis. Study-specific associations were excluded if the SNP was not in Hardy-Weinberg equilibrium (see Additional file 1: Table S4, based on a Holm-Bonferroni alpha criterion), with no variants failing this test. We estimated the effect at the PCSK9 locus by combining all four SNPs in a gene centric score (GS) as the inverse variance weighted effect of the 4 variants, that were subsequently scaled by the inverse variance weighted effect on LDL-C.

Trial data were assembled as per Schmidt et al. 2017 [6]. Briefly, systematic searches were performed using the Cochrane Central Register of Controlled Trials (CENTRAL), MEDLINE, Embase, Web of Science registries, Clinicaltrials.gov and the International Clinical Trials Registry Platform databases. Data from placebo controlled trials were extracted and combined using the inverse variance weighted method for continuous data and a random-intercept logistic regression model for binary data [6].

Results are presented as mean differences (MD) or odds ratios (OR) with 95% confidence intervals (CI). Analyses were conducted using the statistical programme R version 3.4.1 [19]. For study specific estimates please contact AFS.

Results

Participant level data were available from up to 246,355 individuals, and were supplemented by summary effect estimates from data repositories, resulting in a sample size of 320,170 individuals, including 95,865 cases of MI, 16,437 stroke, 11,920 ischemic stroke, 51,623 T2DM, 54,702 cancer, 25,630 Alzheimer's disease and 12,412 of COPD.

Lipid and apolipoprotein associations

As reported previously [14], the four *PCSK9* SNPs were associated with lower LDL-C blood concentrations ranging from -0.02 mmol/L (95% CI -0.03 , -0.02) per allele for rs11583680 to -0.34 mmol/L (95% CI -0.36 ; -0.32) for rs11591147 (See Additional file 2: Figure S1). *PCSK9* SNPs associated with a lower LDL-C concentration were also associated with lower concentrations of apolipoprotein B proportionate to the LDL-C association.

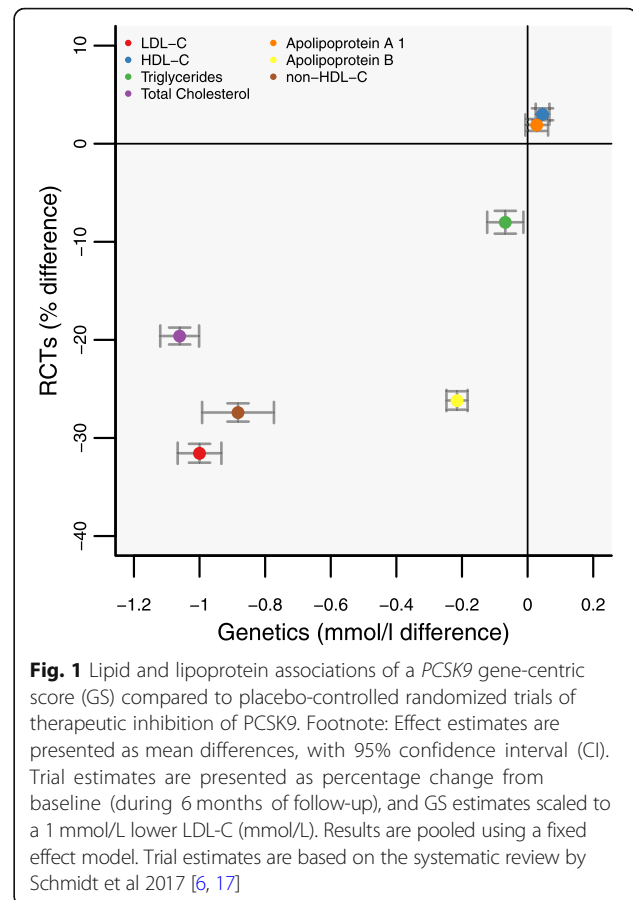
Associations of the GS with the other lipids or apolipoproteins, scaled to a 1 mmol/L lower LDL-C were (Table 1): 0.05 mmol/L (95% CI 0.02, 0.07) for HDL-C, -0.07 mmol/L (95% CI -0.12 , -0.01) for TG, -1.06 mmol/L (95% CI -1.12 , -1.00) for TC, -0.20 g/L (95% CI -0.25 , -0.18) for ApoB, 0.02 g/L (95% CI -0.01 , 0.06) for ApoA1, and -4.12 mg/dL (95% CI -8.62 , 0.38) for Lp(a).

The associations of the *PCSK9* GS with blood-based lipid markers were directionally concordant with effects from treatment trials of therapeutic inhibition of *PCSK9* (Fig. 1).

Table 1 Biomarker associations of a *PCSK9* gene centric score, effect presented as mean difference (MD) with 95% confidence interval in brackets with the effects scaled to a 1 mmol/L decrease in LDL-C

Biomarker	Total sample size	MD (95% CI)
Lipids related biomarkers		
HDL-C in mmol/L	314,078	0.05 (0.02; 0.07)
TG in mmol/L	298,069	-0.07 (-0.12 ; -0.01)
TC in mmol/L	320,170	-1.06 (-1.12 ; -1.00)
ApoA1 in g/L	55,477	0.02 (-0.01 ; 0.06)
ApoB in g/L	54,643	-0.20 (-0.25 ; -0.18)
LP [a] in mg/dL	21,181	-4.12 (-8.62 ; 0.38)
Safety related biomarkers		
SBP in mmHG	182,487	0.03 (-0.05 ; 0.10)
DBP in mmHG	182,497	0.08 (0.001; 0.15)
CRP in log (mg/L)	91,990	0.03 (-0.07 ; 0.14)
IL-6 in log (pmol/L)	22,370	-0.08 (-0.21 ; 0.04)
GGT in log (IU/L)	69,488	0.03 (-0.04 ; 0.10)
Fibrinogen in log(g/dL)	63,288	0.02 (-0.01 ; 0.04)
Hemoglobin in g/L	52,109	1.16 (-0.38 ; 2.70)
ALT in log (IU/L)	83,223	0.03 (-0.02 ; 0.08)
AST in log (IU/L)	49,556	0.01 (-0.03 ; 0.05)
ALP in log (IU/L)	60,222	-0.06 (-0.09 ; -0.02)
Creatinine in umol/L	100,206	0.06 (-1.43 ; 1.55)

Nota bene, TG triglycerides, TC Total cholesterol, ApoA1 Apolipoprotein A1, ApoB Apolipoprotein B, Lp(a) Lipoprotein a, SBP Systolic blood pressure, DBP Diastolic blood pressure, CRP C-reactive protein, IL-6 Interleukin-6, GGT Gamma-glutamyltransferase, ALT Alanine transaminase, AST Aspartate transaminase, ALP Alkaline phosphatase

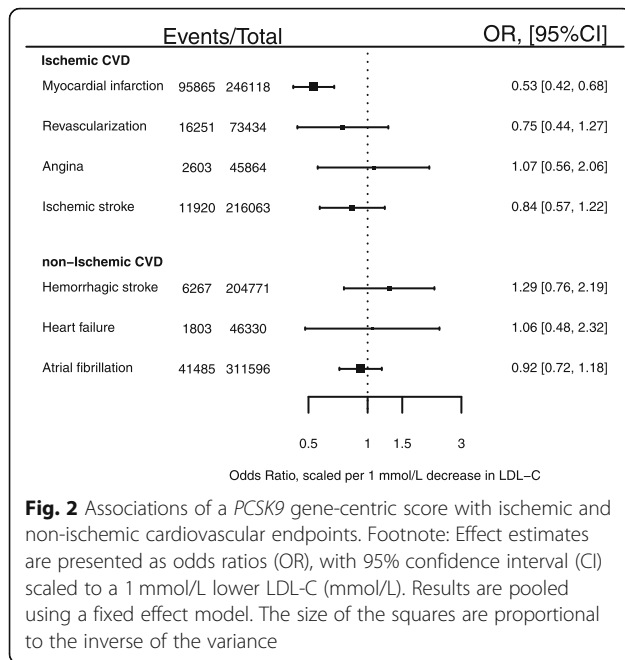


Genetic associations with other biochemical and physiological measures

The GS estimates with SBP and DBP were 0.03 mmHg (95% CI -0.05 , 0.10) and 0.08 mmHg (95% CI 0.0001, 0.15), respectively, per 1 mmol/L lower LDL-C. The *PCSK9* GS was associated with nominally lower ALP (IU/L) -0.06 (95% CI -0.09 , -0.02), but not with other liver enzymes (Table 1).

Genetic associations with ischemic cardiovascular events

The *PCSK9* GS was associated with a lower risk of MI (OR 0.53; 95% CI 0.42; 0.68; 95,865 cases), which was directionally consistent with results from placebo-controlled *PCSK9* inhibition trials: OR 0.90 (95% CI 0.86, 0.93), with both estimates scaled to a 1 mmol/L lower LDL-C (Figs. 2 and 3). The genetic effect estimate for ischemic stroke was OR 0.84 (95% CI 0.57, 1.22, 11,920 cases), concordant in direction to that of the drugs trials (OR 0.85 95% CI 0.78; 0.93). Similarly, the *PCSK9* GS association with coronary revascularization (OR 0.75 95% CI 0.44; 1.27) was directionally consistent with the *PCSK9* inhibitor trials (OR 0.90; 95% CI 0.86, 0.93) (Fig. 3).



Genetic associations with non-ischemic cardiovascular disease

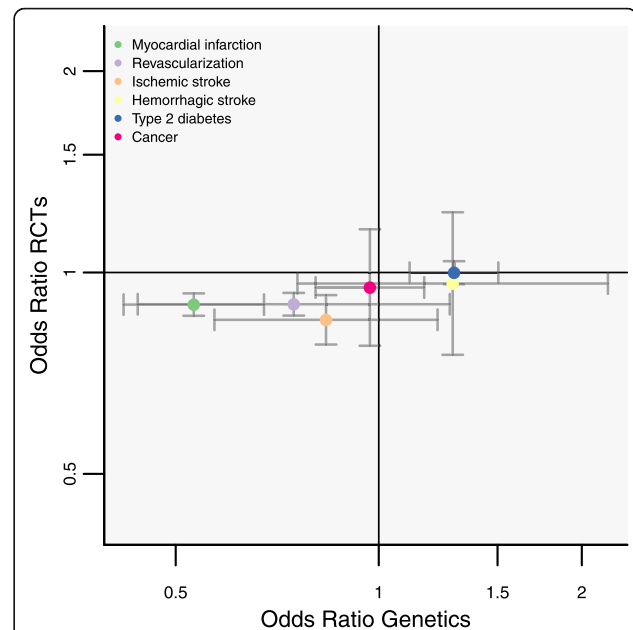
The point estimate for the GS association with hemorrhagic stroke (Fig. 2), OR 1.29 (95% CI 0.76, 2.19), was discordant to the estimate from PCSK9 inhibitor trials (OR 0.96 95% CI 0.75; 1.23) (Fig. 3), although the confidence intervals overlapped. Comparing the association of *PCSK9* GS with hemorrhagic and ischemic stroke indicated the GS had a differential effect (p -value = 0.02). No *PCSK9* GS association was observed with atrial fibrillation (OR 0.92 95% CI 0.72; 1.18; 41,485 cases), or heart failure (OR 1.06 95% CI 0.48; 2.32; 1803 cases) (Fig. 2).

Associations with non-cardiovascular disease and related biomarkers

The *PCSK9* GS was not associated with the risk of any cancer (OR 0.97; 95%CI 0.81; 1.17; 54,702 cases, see Fig. 4), nor with any of 12 specific types of cancer (Additional file 2: Figure S2). We did not observe an association with either Alzheimer's disease or cognitive performance: for Alzheimer's the OR was 0.91 (95% CI 0.55, 1.51) and for cognition (per standard deviation) -0.03 (95% CI -0.22, 0.16). As reported before [14] the GS was associated with T2DM (OR 1.29 95% CI 1.11; 1.50) (Fig. 4), higher body weight (1.03 kg, 95% CI 0.24, 1.82), waist to hip ratio 0.006 (95% CI 0.003, 0.011) and fasting glucose 0.09 mmol/L (95% CI 0.02, 0.15). The OR for COPD was 0.89 (95% CI 0.67, 1.18).

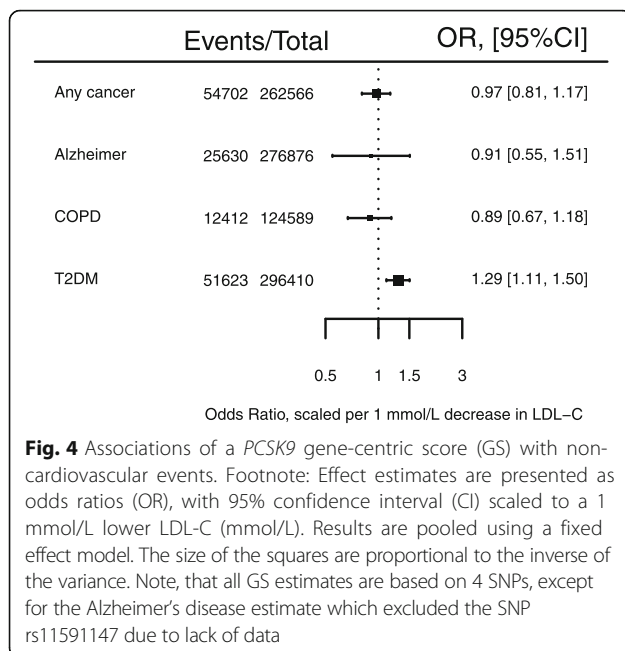
Discussion

The genetic findings presented here show that variation in *PCSK9* is associated with lower circulating



LDL-C and apoB concentrations, lower risk of MI and, with lesser confidence, the risk of ischemic stroke and coronary revascularization. These effects are consistent in direction to effects observed in PCSK9 inhibitor trial's [20].

A recent systematic review of trial data [21] indicated PCSK9 inhibition was associated with increased fasting glucose (0.17 as standardized mean difference [SMD] 95% CI 0.14; 0.19) and glycosylated haemoglobin (0.10 SMD 95% CI 0.07, 0.12, 21), although these associations were dependent on the inclusion of the terminated bococizumab trials. Recently we, and others, showed natural genetic variation *PCSK9* was associated with elevated fasting glucose and T2DM [14, 22, 23] and that variation at other LDL-C-associated loci also influence risk of T2DM [24, 25]. However, the FOURIER and ODYSSEY OUTCOMES trials, the largest treatment trials of PCSK9 inhibitors to date, did not find an association with risk of incident T2DM, at a median follow up of 2.2 and 2.8 years respectively. It is possible this reflects a genuine discordance between the findings from trials and genetic analyses. Alternatively, the exposure durations in the two largest trials may simply have been too short for subjects to develop T2DM. The risk increasing effect of statins on T2DM was only apparent after conducting a



meta-analysis of 13 statin trials in which 4278 T2DM cases were observed during an average follow-up of 4 years [26].

In general, inconsistencies between associations of variants in a gene encoding a drug target and the effects of the corresponding treatment are possible on a number of theoretical grounds. The effects of genetic variation (present from conception) may be mitigated by developmental adaptation or environmental changes. A lack of association of a genetic variant with an outcome therefore does not preclude an effect of a treatment administered in later life, when adaptive responses may no longer be available, or in the presence of a particular environment [27]. We selected a subset of all genetic variants at *PCSK9* that capture information on many others and which have some annotated function. However, other approaches to more fully capture the entire gene-centric effect are worthy of future investigation [28].

The association of *PCSK9* variants with LDL-C and MI has been reported before [5], and was a motivating factor for the development of *PCSK9* inhibiting drugs. Lotta and colleagues [22] reported a similar OR for MI of 0.60 (95% CI 0.48, 0.75) per 1 mmol/L decrease in LDL-C using the *PCSK9* rs11591147 SNP. Using a seven SNP *PCSK9* GS, Ference et al. reported a MI OR of 0.44 (95% CI 0.31, 0.64) per 1 mmol/L decrease in LDL-C [23]. These scaled genetic effects are larger than the treatment effect observed in trials which others have noted previously [29], and ascribed to the lifelong effect of genetic variation versus the short-term effect of drug treatment in later life.

The available trial data showed *PCSK9* inhibitors had a similar effect on MI (OR 0.90, 95% CI 0.86; 0.93) and ischemic stroke (OR 0.85 95% CI 0.78; 0.93). By contrast, the genetic analysis indicated a directionally concordant, but larger effect on MI (OR 0.53; 95% CI 0.42; 0.68) than ischemic stroke, (OR 0.84 95% CI 0.57; 1.22). The genetic analysis was, however, based on only 11,920 stroke cases, about one-fifth of the number of cases available for the genetic analysis of MI and as such confidence interval overlapped. We did observe a differential association between *PCSK9* SNPs and ischemic and hemorrhagic stroke (interaction p -value = 0.02). Findings from statin trials previously suggested LDL-C lowering through inhibition of HMG-coA reductase is associated with a reduced risk of ischemic but potentially increased risk of hemorrhagic stroke [30–32]. Our findings suggest that a different effect on ischemic and hemorrhagic stroke subtypes may be eventually identified for *PCSK9* inhibitors.

Despite previous concerns about a potential effect of this class of drugs on cognition [33], the genetic analysis did not reveal a significant association of the *PCSK9* variants with cognitive function or Alzheimer's disease, nor with COPD or cancer, though this does not preclude an effect on such outcomes from drug treatment given in later life. While we explored the associations with any cancer (54,702 events) as well as individual cancer sites (Additional file 2: Figure S2), we did not have data on some clinically relevant cancer types such as endometrial cancer.

This neutral effect on cognition has been recently reported by the EBBINGHAUS study, nested within the FOURIER trial, which reported a non-significant *PCSK9* inhibitor effect on multiple measures of cognition confirming (using a non-inferiority design) an absence of effect [33]; it should be noted that similar to the FOURIER, the EBBINGHAUS follow-up time was limited. The absence of an effect on cognition during *PCSK9* inhibitor treatment was also observed in the ODYSSEY OUTCOMES trial, which had a median follow-up [7] of 2.8 years.

Drugs (even apparently specific monoclonal antibodies) can exert actions on more than one protein if such targets belong to a family of structurally similar proteins. *PCSK9*, for example, is one of nine related proprotein convertases [34]. Such 'off-target' actions, whether beneficial or deleterious, would not be shared by variants in the gene encoding the target of interest. In addition, monoclonal antibodies prevent interaction between circulating *PCSK9* and LDL-receptor and should not, in theory, influence any intracellular action of the protein [35].

Genetic association studies of the type conducted here tend to examine the risk of a first clinical event, whereas

clinical trials such as ODYSSEY OUTCOMES focus on patients with established disease, where mechanisms may be modified. Proteins influencing the risk of a first event may also influence the risk of subsequent events, as observed in the case of the target of statin drugs that are effective in both primary and secondary prevention [1]. For this and other reasons [36–38], examination of the effects of *PCSK9* variants on the risk of subsequent CHD events in patients with established coronary atherosclerosis is the subject of a separate analysis led by the GENIUS-CHD consortium [38].

Conclusions

PCSK9 SNPs associated with lower LDL-C predict a substantial reduction in the risk of MI and concordant associations with a reduction in risk of ischemic stroke, but with a modestly increased risk of T2DM. In this preliminary analysis we did not observe associations with other non-cardiovascular safety outcomes such as cancer, COPD, Alzheimer's disease or atrial fibrillation.

Additional files

Additional file 1: Supplemental tables. (XLSX 62 kb)

Additional file 2: Supplemental figures and study acknowledgments. (PDF 154 kb)

Abbreviations

ALP: Alkaline phosphatase; ALT: Alanine aminotransferase; ApoA1: Apolipoproteins A1; ApoB: Apolipoproteins B; AST: Aspartate transaminase; CADD: Combined annotation dependent depletion; CHD: Coronary heart disease; CI: Confidence interval; COPD: Chronic obstructive pulmonary disease; CRP: C-reactive protein; CVD: Cardiovascular disease; DBP: Diastolic blood pressure; GGT: Gamma-glutamyltransferase; GLGC: Global lipids genetics Consortium; GS: Gene-centric score; HbA1c: Glycated haemoglobin; IL-6: Interleukin-6; IPD: Individual participant-level data; LD: Linkage disequilibrium; LDL-C: Low density lipoprotein-cholesterol; LpA: Lipoprotein a; mAbs: Monoclonal antibodies; MAF: Minor allele frequency; MD: Mean difference; MI: Myocardial infarction; OR: Odds ratio; SBP: Systolic blood pressure; SMD: Standardized mean difference; T2DM: Type 2 diabetes mellitus; TC: Total cholesterol; TG: Triglycerides

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Authors' contributions

AFS, DIS, MVH, RSP, FWA, JPC, BJK, ADH, DP, NS contributed to the idea and design of the study. AFS, DIS, MVH, designed the analysis scripts shared with individual centres. AFS performed the meta-analysis and had access to all the data. The authors jointly drafted the manuscript, and contributed to subsequent critical revisions. All authors have approved the submitted manuscript, and take responsibility for the integrity and the accuracy of the data and presented results.

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Availability of data and materials

The datasets used and/or analysed during the current study are available from the corresponding author on reasonable request.

Ethics approval and consent to participate

Local ethics committees for studies contributing data to these analyses granted approval for the work.

Consent for publication

Not applicable

Competing interests

Dr. Holmes has collaborated with Boehringer Ingelheim in research, and in accordance with the policy of the The Clinical Trial Service Unit and Epidemiological Studies Unit (University of Oxford), did not accept any personal payment. David Preiss consulted for Amgen on a single occasion but, in accordance with the policy of the Clinical Trial Service Unit (University of Oxford), did not accept any personal payment. He is an investigator on a clinical trial of the PCSK9 synthesis inhibitor, inclisiran, funded by a grant to the University of Oxford by the Medicines Company, but he receives no personal fees from this grant. Daniel I Swerdlow is an employee of BenevolentAI Ltd. Aroon Hingorani and Harry Hemingway are National Institute for Health Research Senior Investigators. Naveed Sattar consulted for Amgen and Sanofi related to PCSK9 inhibitors; and was an investigator on clinical trials of PCSK9 inhibition funded by Amgen. Naveed Sattar has also consulted for Boehringer Ingelheim, Janssen, Eli-Lilly and NovoNordisk. Daniel Swerdlow has consulted to Pfizer for work unrelated to this paper. Folkert W. Asselbergs is supported by UCL Hospitals NIHR Biomedical Research Centre. Dr. Patel has received honoraria and speaker fees from Sanofi, Amgen and Bayer. Kees Hovingh or his institution (AMC) received honoraria for consultancy, ad boards, and/or conduct of clinical trials from: AMGEN, Aegerion, Pfizer, Astra Zeneca, Sanofi, Regeneron, KOWA, Ionis pharmaceuticals and Cerenis. Bertrand Cariou has received research funding from Pfizer and Sanofi, received honoraria from AstraZeneca, Pierre Fabre, Janssen, Eli-Lilly, MSD Merck & Co., Novo-Nordisk, Sanofi, and Takeda, and has acted as a consultant/advisory panel member for Amgen, Eli Lilly, Novo-Nordisk, Sanofi, and Regeneron. Andrzej Pajak acted as a consultant/advisory panel member for Amgen. Erik Ingelsson is a scientific advisor for Precision Wellness and Olink Proteomics for work unrelated to this paper. JCH is a scientific advisor to a clinical trial of PCSK9 inhibition. AE Honoraria: Takeda, BMS, Amgen; Consulting: Takeda, BMS, Amgen. SEH acknowledges BHF funding (PG008/08) and support from the UCL BRC. All other authors declare no competing interests.

Author details

¹Institute of Cardiovascular Science, University College London, 222 Euston Road, London NW1 2DA, UK. ²Department of Cardiology, Division Heart & Lungs, University Medical Center Utrecht, Utrecht University, Utrecht, the Netherlands. ³UCL's BHF Research Accelerator Centre, London, UK. ⁴Medical Research Council Population Health Research Unit, Clinical Trial Service Unit & Epidemiological Studies Unit (CTSUs), Nuffield Department of Population Health, University of Oxford, Oxford, UK. ⁵Department of Medicine, Imperial College London, London, UK. ⁶Health Data Research UK, University College London, 222 Euston Road, London NW1 2DA, UK. ⁷Institute of Health Informatics, University College London, 222 Euston Road, London NW1 2DA, UK. ⁸The Alan Turing Institute, British Library, 96 Euston Rd, London NW1 2DB, UK. ⁹University College London, Farr Institute of Health Informatics, London, UK. ¹⁰Clinical Trial Service Unit & Epidemiological Studies Unit (CTSUs), Nuffield Department of Population Health, University of Oxford, Richard Doll Building, Old Road Campus, Roosevelt Drive, Oxford OX3 7LF, UK. ¹¹Postgraduate Program in Epidemiology, Federal University of Pelotas, Pelotas, Brazil. ¹²Centre for Population Health Research, Sansom Institute for Health Research, University of South Australia, Adelaide, Australia. ¹³Population, Policy and Practice, UCL GOS Institute of Child Health, London, UK. ¹⁴South Australian Health and Medical Research Institute, Adelaide, Australia. ¹⁵Durrer Center for Cardiovascular Research, Netherlands Heart Institute, Utrecht, The Netherlands. ¹⁶Department of Clinical Epidemiology, Biostatistics And Bioinformatics, Academic Medical Center Amsterdam, Amsterdam, the Netherlands. ¹⁷Department of vascular medicine, Academic Medical Center Amsterdam, Amsterdam, the Netherlands. ¹⁸Charité – Universitätsmedizin Berlin, corporate member of Freie Universität Berlin, Humboldt-Universität zu Berlin, and Berlin Institute of Health, Lipid Clinic at the Interdisciplinary Metabolism Center, Berlin, Germany. ¹⁹Charité - Universitätsmedizin Berlin, BCRT - Berlin Institute of Health Center for Regenerative Therapies, Berlin, Germany. ²⁰Institute of Nutritional Science, University of Potsdam, 14558 Nuthetal, Germany. ²¹Geriatrics Research Group,

Charité - Universitätsmedizin Berlin, 13347 Berlin, Germany. ²²Department of Nutrition and Gerontology, German Institute of Human Nutrition Potsdam-Rehbruecke, 14558 Nuthetal, Germany. ²³CA Economics GmbH, Berlin, Germany. ²⁴Lübeck Interdisciplinary Platform for Genome Analytics (LIGA), Institutes of Neurogenetics & Cardiogenetics, University of Lübeck, Lübeck, Germany. ²⁵Center for Lifespan Changes in Brain and Cognition (LCBC), Dept. Psychology, University of Oslo, Oslo, Norway. ²⁶Genetic and Molecular Epidemiology Group, Lübeck Interdisciplinary Platform for Genome Analytics (LIGA), Institutes of Neurogenetics & Cardiogenetics, University of Lübeck, Lübeck, Germany. ²⁷Institute of Human Genetics, Lübeck, Germany. ²⁸Ageing Epidemiology Research Unit, School of Public Health, Imperial College, London, UK. ²⁹Institute of Genetic Epidemiology, Department of Genetics and Pharmacology, Medical University of Innsbruck, 6020 Innsbruck, Austria. ³⁰Department of Neurology, Medical University Innsbruck, Innsbruck, Austria. ³¹Department of Neurology, Inselspital, University Hospital Bern, University of Bern, Bern, Switzerland. ³²Bradford Institute for Health Research, Bradford Royal Infirmary, Bradford, UK. ³³Department Primary Care & Population Health, University College London, London, UK. ³⁴Population Health Research Institute, St George's, University of London, London, UK. ³⁵Population Health Sciences, University of Bristol, Bristol, UK. ³⁶Centre for Population Health Sciences, The Usher Institute, University of Edinburgh, Edinburgh, UK. ³⁷Department of Epidemiology and Public Health, UCL Institute of Epidemiology and Health Care, University College London, London, UK. ³⁸Department of Medicine, Internal Medicine, Lausanne University Hospital, Lausanne, Switzerland. ³⁹Department of Vascular Surgery, Imperial College, London, United Kingdom. ⁴⁰Department of Surgery, Nicosia Medical School, University of Nicosia, Nicosia, Cyprus. ⁴¹Cyprus International Institute for Environmental and Public Health, Cyprus University of Technology, Limassol, Cyprus. ⁴²Julius Center for Health Sciences and Primary Care, University Medical Center Utrecht, Utrecht University, Utrecht, the Netherlands. ⁴³Italian Institute for Genomic Medicine (IIGM), Turin, Italy. ⁴⁴Department of Medical Sciences, University of Turin, Turin, Italy. ⁴⁵Unit of Cancer Epidemiology, Città della Salute e della Scienza University-Hospital and Center for Cancer Prevention (CPO), Turin, Italy. ⁴⁶MRC Epidemiology Unit, Institute of Metabolic Science, University of Cambridge School of Clinical Medicine, Cambridge Biomedical Campus, Addenbrooke's Hospital, Cambridge, UK. ⁴⁷Novosibirsk State Medical University, Novosibirsk, Russian Federation. ⁴⁸Institute of Internal and Preventive Medicine, Siberian Branch of the Russian Academy of Medical Sciences, Novosibirsk, Russian Federation. ⁴⁹Department of Epidemiology and Population Studies, Faculty of Health Sciences, Jagiellonian University Medical College, Kraków, Poland. ⁵⁰National Institute of Public Health, Prague, Czech Republic. ⁵¹Lithuanian University of Health Sciences, Kaunas, Lithuania. ⁵²Novo Nordisk Foundation Center for Basic Metabolic Research, Faculty of Health and Medical Sciences, University of Copenhagen, Copenhagen, Denmark. ⁵³Department of Clinical Medicine, Faculty of Health and Medical Sciences, University of Copenhagen, Copenhagen, Denmark. ⁵⁴Center for Clinical Research and Prevention, Bispebjerg and Frederiksberg Hospital, Copenhagen, The Capital Region of Denmark, Denmark. ⁵⁵Centre for Cardiovascular Genetics, Department of Medicine, University College London, London, UK. ⁵⁶Center for Human Genetics, Marshfield Clinic Research Institute, Marshfield, USA. ⁵⁷Children's Hospital of Philadelphia, Philadelphia, USA. ⁵⁸University of Minnesota, Minneapolis, USA. ⁵⁹Geisinger, Danville, USA. ⁶⁰Kaiser Permanente Washington Health Research Institute, Seattle, WA, USA. ⁶¹Department of Biomedical Informatics and Medical Education University of Washington Seattle, Seattle, WA, USA. ⁶²Mayo Clinic, Rochester, USA. ⁶³Department of Medicine, Department of Pharmacology, Department of Biomedical Informatics, Vanderbilt University School of Medicine, Nashville, TN, USA. ⁶⁴Vanderbilt University, Nashville, USA. ⁶⁵WHI, Seattle, USA. ⁶⁶University of Newcastle, Newcastle, NSW, Australia. ⁶⁷Public Health Program, Hunter Medical Research Institute, Newcastle, NSW, Australia. ⁶⁸Hunter New England Local Health District, Newcastle, NSW, Australia. ⁶⁹Population Health Research Institute, Hamilton, Ontario, Canada. ⁷⁰Department of Genetics, University of Groningen, University Medical Center Groningen, Groningen, The Netherlands. ⁷¹Department of Cardiology, University of Groningen, University Medical Center Groningen, Groningen, The Netherlands. ⁷²Department of Epidemiology, University of Groningen, University Medical Center Groningen, Groningen, Netherlands. ⁷³Department of Clinical Epidemiology, Leiden University Medical Center, Leiden, The Netherlands. ⁷⁴Department of Medical Sciences, Molecular Epidemiology, Uppsala University, Uppsala, Sweden. ⁷⁵Department of Medicine, Division of Cardiovascular Medicine, Stanford University School of Medicine, Stanford, CA 94305, USA. ⁷⁶Department of Medical Sciences, Molecular

Epidemiology and Science for Life Laboratory, Uppsala University, Uppsala, Sweden. ⁷⁷Department of Epidemiology, Erasmus University Medical Center, Rotterdam, the Netherlands. ⁷⁸Department of Epidemiology and Biostatistics, Imperial College London, London, UK. ⁷⁹Institute for Community Medicine, University Medicine Greifswald, Greifswald, Germany. ⁸⁰DZHK (German Centre for Cardiovascular Research), partner site Greifswald, Greifswald, Germany. ⁸¹Chair of Epidemiology, Ludwig-Maximilians-Universität München, UNIKA-T Augsburg, Augsburg, Germany. ⁸²Department of Internal Medicine B, University Medicine Greifswald, Greifswald, Germany. ⁸³Department of Internal Medicine A, University Medicine Greifswald, Greifswald, Germany. ⁸⁴Interfaculty Institute of Genetics and Functional Genomics, University Medicine Greifswald, Greifswald, Germany. ⁸⁵Institute of Health and Wellbeing, University of Glasgow, Glasgow G12 8RZ, Scotland, UK. ⁸⁶Department of Non-Communicable Disease Epidemiology, London School of Hygiene & Tropical Medicine, London, UK. ⁸⁷The Department of Medical Research, Bærum Hospital, Vestre Viken Hospital Trust, Gjøtting, Norway. ⁸⁸Division of Pharmacoepidemiology and Clinical Pharmacology, Utrecht Institute of Pharmaceutical Sciences, Faculty of Science, Utrecht University, Utrecht, The Netherlands. ⁸⁹Department of Respiratory Medicine, Academic Medical Centre, University of Amsterdam, Amsterdam, the Netherlands. ⁹⁰Robertson Centre for Biostatistics, University of Glasgow, Glasgow, UK. ⁹¹Institute of Genetics and Molecular Medicine, University of Edinburgh, Edinburgh, UK. ⁹²Institute of Cardiovascular and Medical Sciences, University of Glasgow, Glasgow G12 8TA, UK. ⁹³CNRS UMR 8199, European Genomic Institute for Diabetes (EGID), Institut Pasteur de Lille, University of Lille, 59000 Lille, France. ⁹⁴Department of Genomics of Common Disease, Imperial College London, W12 0NN London, United Kingdom. ⁹⁵INSERM, U-1138, Centre de Recherche des Cordeliers, Paris, France. ⁹⁶UFR de Médecine, Université Paris Diderot, Sorbonne Paris Cité, Paris, France. ⁹⁷Département de Diabétologie, Endocrinologie et Nutrition, Assistance Publique Hôpitaux de Paris, Hôpital Bichat, Paris, France. ⁹⁸Institut du Thorax, INSERM, CNRS, UNIV Nantes, CHU Nantes, Nantes, France. ⁹⁹Institute for Social and Economic Research, University of Essex, Essex, UK. ¹⁰⁰Wellcome Trust Centre for Human Genetics, University of Oxford, Oxford, England. ¹⁰¹Boston University School of Medicine, Boston, MA, USA. ¹⁰²Harvard Medical School Center for Cardiovascular Disease Prevention Brigham and Women's Hospital, Boston, USA. ¹⁰³UWash, Seattle, USA. ¹⁰⁴Penn State, State College, USA. ¹⁰⁵University of Colorado Denver, Denver, USA. ¹⁰⁶Division of Genetics and Epidemiology, The Institute of Cancer Research, London, UK. ¹⁰⁷Div. Molecular Genetic Epidemiology German Cancer Research Center, Im Neuenheimer Feld 580, 69120 Heidelberg, Germany. ¹⁰⁸Deutsches Krebsforschungszentrum, Heidelberg, Germany. ¹⁰⁹The Institut du Cerveau et de la Moelle épinière – ICM, Paris, France. ¹¹⁰Sorbonne Universités, UPMC Université Paris 06, UMR S 1127, F-75013 Paris, France. ¹¹¹Department of Neurosurgery, Bethel Clinic, Kantensiek 11, 33617 Bielefeld, Germany. ¹¹²Division of Hematology-Oncology, Department of Pediatrics, Dan L. Duncan Cancer Center, Baylor College of Medicine, Houston, Texas 77030, USA. ¹¹³Northern Institute for Cancer Research, Newcastle University, Newcastle upon Tyne, UK. ¹¹⁴Myeloma Institute for Research and Therapy, University of Arkansas for Medical Sciences, Little Rock, USA. ¹¹⁵Department of Hematology, Erasmus MC Cancer Institute, 3075 EA Rotterdam, the Netherlands. ¹¹⁶Hematology and Transfusion Medicine, Department of Laboratory Medicine, BMC B13, SE-221 84 Lund, Sweden. ¹¹⁷University Clinic Heidelberg, Internal Medicine V and National Center for Tumor Diseases (NCT), Heidelberg, Germany. ¹¹⁸Department of Internal Medicine, University Hospital of Cologne, Cologne, Germany. ¹¹⁹Hematology Clinic, Skåne University Hospital, Skåne, Sweden. ¹²⁰Wallenberg Center for Molecular Medicine, Lund University, Lund, Sweden. ¹²¹The National Institute for Health Research University College London Hospitals Biomedical Research Centre, University College London, 222 Euston Road, London NW1 2DA, UK. ¹²²Health Data Research UK and Institute of Health Informatics, University College London, London, United Kingdom. ¹²³The Barts Heart Centre, St Bartholomew's Hospital, London, UK. ¹²⁴UPenn, Philadelphia, USA. ¹²⁵Massachusetts Veterans Epidemiology and Research Information Center (MAVERIC) Veterans Affairs Boston Healthcare System, Boston, USA.

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