Family history and polygenic risk of cardiovascular disease:

independent factors associated with secondary cardiovascular

events in patients undergoing carotid endarterectomy

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Short title: Family history and polygenic risk are independently associated with sCVE

Category: Original article

3 tables, 3 figures (supplemental: 8 tables, 1 figure)

Word count abstract: 245

Total word count manuscript: 4991

Declarations of interest: none

Financial support

Dr. van der Laan was funded through grants from the Netherlands CardioVascular Research Initiative of

the Netherlands Heart Foundation (CVON 2011/B019 and CVON 2017-20: Generating the best evidence-

based pharmaceutical targets for atherosclerosis [GENIUS I&II]). Dr. Timmerman, Dr. De Borst, dr. De

Kleijn, and dr. Pasterkamp are funded through the EU 755320 Taxinomisis grant. We acknowledge the

European Research Area Network on Cardiovascular diseases (ERA-CVD, grant number 01KL1802).

There are no competing interests declared.

Acknowledgements

We thank Sara van Laar, Sara Bregman, Bram Vermeulen, Evelyn Velema and Joëlle van Bennekom for

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their technical support in the Athero-Express Biobank.

Abstract

Background

Family history (FHx) of cardiovascular disease (CVD) is a risk factor for CVD and a proxy for

cardiovascular heritability. Polygenic risk scores (PRS) summarizing >1 million variants for

coronary artery disease (CAD) are associated with incident and recurrent CAD events. However,

little is known about the influence of FHx or PRS on secondary cardiovascular events (sCVE) in

patients undergoing carotid endarterectomy (CEA).

Methods

We included 1,788 CEA patients from the Athero-Express Biobank. A weighted PRS for CAD

including 1.7 million variants was calculated (MetaGRS). The composite endpoint of sCVE

during three years follow-up included coronary, cerebrovascular and peripheral events and

cardiovascular death. We assessed the impact of FHx and MetaGRS on sCVE and carotid plaque

composition.

Results

Positive FHx was associated with a higher 3-year risk of sCVE independent of cardiovascular

risk factors and MetaGRS (adjusted HR 1.40, 95% CI 1.07-1.82, p=0.013). Patients in the highest

MetaGRS quintile had a higher 3-year risk of sCVE compared to the rest of the cohort

independent of cardiovascular risk factors including FHx (adjusted HR 1.35, 95%CI 1.01-1.79,

p=0.043), and their atherosclerotic plaques contained more fat (adjusted OR 1.59, 95%CI, 1.11-

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2.29, p=0.013) and more macrophages (OR 1.49, 95% CI 1.12-1.99, p=0.006).

Conclusion

In CEA patients, both positive FHx and higher MetaGRS were independently associated with

increased risk of sCVE. Moreover, higher MetaGRS was associated with vulnerable plaque

characteristics. Future studies should unravel underlying mechanisms and focus on the added

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value of PRS and FHx in individual risk prediction for sCVE.

Key words

Atherosclerosis

Polygenic risk score

Family history

Vascular biology

Carotid artery stenosis

Carotid endarterectomy

Abbreviations

PRS polygenic risk score

CVD cardiovascular disease

CAD coronary artery disease

SNP single-nucleotide polymorphism

FHx family history of premature CVD

MetaGRS polygenic risk score for CAD including 1.7 million SNPs associated with CAD

CEA carotid endarterectomy

CVE cardiovascular event

sCVE secondary cardiovascular event

MI myocardial infarction

AAA abdominal aortic aneurysm

ACS acute coronary syndrome defined as a composite of myocardial infarction and

unstable angina pectoris

PAD peripheral artery disease

GWAS large-scale genome wide association studies

Introduction

Family history of cardiovascular disease (FHx) is a major risk factor for primary cardiovascular disease (CVD) and serves as a surrogate for genetic predisposition.(1, 2) Risk prediction for secondary cardiovascular events remains challenging as traditional risk factors have limited discriminative performance.(3) The main underlying mechanism of CVD is atherosclerosis, and atherosclerotic plaque composition, exemplified by intraplaque haemorrhage (IPH), was shown to be associated to adverse secondary cardiovascular events (sCVE).(4) Yet, the relevance of FHx for secondary outcome of cardiovascular events is still unclear.(5–9)

Large-scale genome-wide association studies (GWAS) have identified hundreds of common genetic variants (single-nucleotide polymorphisms or SNPs) robustly associated with coronary artery disease (CAD)(10–14) and ischemic stroke(15–17) predisposition, albeit with small individual effects. Exact pathobiological mechanisms leading to cardiovascular symptoms are still poorly understood, but CAD- and ischemic stroke genetic variants were previously associated to atherosclerotic plaque composition.(18)

Polygenic risk scores (PRS) summarize the small individual genetic effects into a quantitative measure of genetic disease susceptibility. Recent studies showed that PRS were strongly correlated with prevalent and incident CAD independent of traditional risk factors including family history in the UK Biobank population.(19, 20) For example, individuals with higher scores of the MetaGRS (a PRS for CAD including 1.7 million SNPs) were at 1.7-4.2 fold higher risk for a first coronary event compared to individuals with lower MetaGRS scores. In

addition, a very recent study in a French-Canadian population with established CAD showed that MetaGRS was also associated with increased risk of recurrent CAD events.(21)

As atherosclerosis is known for its complex nature with genetic overlap in different diseases such as stroke, CAD, abdominal aortic aneurysm (AAA) and peripheral artery disease (PAD)(22), we aimed to investigate the association between MetaGRS and long-term secondary cardiovascular events in a different cohort of severe atherosclerotic patients with carotid artery stenosis undergoing carotid endarterectomy. Given that FHx is used in clinical practice as a derivative of genetic background, we also examined the association between FHx and secondary cardiovascular events. Moreover, to explore possible underlying pathophysiological mechanisms, we studied the impact of MetaGRS and FHx on carotid histological plaque characteristics.

Methods

Athero-Express Biobank

All patients in this study were included in the Athero-Express Biobank, a prospective cohort study that included consecutive patients with severe carotid artery stenosis undergoing CEA in two large tertiary referral hospitals in The Netherlands, the University Medical Centre Utrecht (inclusion is ongoing) and the St. Antonius Hospital Nieuwegein (inclusion until 2014).(23) The study design has been published before.(23) In short, patient characteristics, such as demographics, cardiovascular risk factors including medical history, medication use, and FHx for cardiovascular disease, were obtained through standardized questionnaires and checked in electronic health records. Preoperative blood samples were drawn. The atherosclerotic plaque obtained during surgery was collected and immunohistochemically analysed for plaque characteristics. Patients were followed up for three years after surgery for the occurrence of secondary cardiovascular events through standardized questionnaires and by checking electronic health records. General practitioners were consulted in case of no response to questionnaires or in order to obtain further information regarding reported cardiovascular events. Patients operated for restenosis (6% of 2,044 eligible patients for this study) were excluded because these differ in future cardiovascular event risk.(24) Thus for the current study, a total of 1,788 patients operated from March 2002 until July 2016 had available 3-year follow-up data and FHx data, and were included for analysis. Of these, 1,551/1,788 (87%) patients had available histological carotid plaque data. A total of 1,319/1,788 (74%) patients had available genotype data of whom 1,301 (98%) also had histological carotid plaque data. This study was performed according to the Declaration of Helsinki and was approved by the local ethics committee of both hospitals. Patients provided written informed consent before study participation.

Definitions

A positive FHx was defined as having a first-degree relative (either a parent or sibling) with onset of cardiovascular disease (myocardial infarction (MI), coronary artery stenosis, stroke, AAA, or cardiovascular death including sudden death) before the age of 60 years. The primary outcome of this study was defined as a composite secondary cardiovascular event (sCVE) within three years of follow-up including fatal or non-fatal MI, fatal or non-fatal stroke, ruptured AAA, fatal cardiac failure, coronary or peripheral interventions (either percutaneous or bypass surgery), leg amputation due to cardiovascular causes and cardiovascular death. Secondary outcomes were histological atherosclerotic carotid plaque characteristics.

Genotyping

Methods for genotyping, quality control and imputation in the Athero-Express biobank have been published elsewhere (25, 26). Briefly, DNA was extracted from EDTA whole blood samples or if not present from atherosclerotic plaque tissue according to validated protocols. Genotyping was performed with two commercially available chips: the first batch by Affymetrix Genome-Wide Human SNP array 5.0 (previously used in the Athero-Express Genomics Study 1 (AEGS1), covering samples obtained in 2002-2007) and the second batch by Affymetrix Axiom GW CEU 1 array (previously used in Athero-Express Genomics Study 2 (AEGS2), covering samples obtained in 2002-2013). Procedures for data quality control and data cleaning were in accordance with global standards.(27) After genotype calling according to Affymetrix' specification, data was filtered on 1) individual call rate > 97%, 2) genotype call rate > 97%, 3) minor allele frequencies (MAF) > 3%, 4) average heterozygosity rate ± 3.0 standard deviations,

5) relatedness (pi-hat > 0.20), 6) Hardy–Weinberg Equilibrium (HWE p $< 1.0 \times 10^{-6}$), and 7) population stratification (based on HapMap 2, release 22, b36) by excluding samples deviating more than 6 standard deviations from the average in 5 iterations during principal component analysis and by visual inspection.(25) After pre-phasing using SHAPEIT2 v2.644, a combined dataset of 1000 Genome (phase 3, version 5) and The Genome of the Netherlands Project release

5 was used as a reference for imputation with IMPUTE2 v2.3.0.(28)

Polygenic risk score (MetaGRS)

To estimate the polygenic cardiovascular disease susceptibility for included patients in our cohort, we used the previously published polygenic risk score for CAD (MetaGRS).(19) Its construction was described elsewhere.(19) Briefly, the MetaGRS comprises 1.7 million genetic variants associated with CAD and was constructed through meta-analysis of three genomic risk scores: GRS46K (comprising 46,000 cardiometabolic genetic variants), FDR202 (including 202 genetic variants associated with CAD at false discovery rate p<0.05 in the recent GWAS CARDIoGRAMplusC4D), and the 1000Genomes genetic score also created with CARDIoGRAMplusC4D. The MetaGRS was internally and externally validated for the primary risk of prevalent and incident CAD in the UK Biobank.(19) We calculated the MetaGRS for each included patient in this study and standardized it to mean-zero and unit-variance for each genotyping batch separately, i.e. AEGS1 and AEGS2, respectively.

Sample handling

After CEA, the atherosclerotic plaque was directly processed in the laboratory following standardized protocols.(4, 23, 29, 30) The plaque was cut in cross-sectional segments of 5mm.

The segment with largest plaque burden was chosen as the culprit lesion and immunohistochemically analysed for macrophages, smooth muscle cells (SMC), lipid core, calcification, collagen, intraplaque haemorrhage (IPH) and microvessel content. Extensive description of the standardized protocol for atherosclerotic plaque processing and analysis of plaque characteristics has been previously reported and is added to the Supplemental. (4, 23, 29, 30) To assess the overall vulnerability of the atherosclerotic plaque, a vulnerability score was created ranging from 0-5 with 1 point for plaque characteristics that are considered hallmarks of a vulnerable plaque (moderate/heavy macrophages, no/minor collagen, no/minor SMC, lipid core>10% and presence of IPH), based on a previous publication.(31)

Statistical analysis

Baseline characteristics were compared between patient groups (FHx and MetaGRS) by chisquare test for categorical variables and Student's t-test for continuous variables (lipid levels were log-transformed). We analysed the association between FHx and MetaGRS and sCVE by Cox-proportional hazard regression models and the associations with plaque characteristics through logistic or linear regression models. To fully unravel the genetic association, the association of MetaGRS with sCVE and plaque characteristics was analysed in three ways: (1) MetaGRS as a continuous quantification of genetic CAD susceptibility (2) patients in the top 20% of the MetaGRS distribution compared to the remaining 80%, and (3) patients in the top 20% of the MetaGRS distribution compared to those in the bottom 20% of the distribution. Kaplan-Meier curves were constructed to graphically illustrate univariate associations. Confounders for multivariable analyses were selected based on literature(19, 31, 32) (for sCVE these were age, sex, diabetes, BMI, smoking and hypercholesterolemia and for plaque

characteristics these were age, sex, surgery year and type of cerebrovascular symptoms). Additional confounders were added when showing an association of p<0.20 with the determinant (FHx or MetaGRS) and outcome of interest (sCVE or plaque characteristics). For MetaGRS models genotype array and principal components 1-4 were also added. Full model description is displayed in the Supplemental Tables S1 and S2. Because a previous study in our biobank showed that IPH is associated with sCVE(4), IPH was added to multivariable models of FHx, MetaGRS and sCVE to explore whether IPH could be one possible underlying mechanism. Sexstratified analyses were performed to unravel sex-dependent differences in associations. Values with p<0.05 were considered statistically significant. All analyses were performed in IBM SPSS Statistics version 25.0.

Results

Patient selection from the Athero-Express Biobank and characteristics of the study population are displayed in Figure 1 and Table 1. Patients had a mean age 69 years and 70% were men. The cohort represented a typically severe atherosclerotic cohort with high prevalence of traditional risk factors and atherosclerotic manifestations in other vascular beds (coronary or peripheral arteries, respectively 30% and 20%). Baseline characteristics were similar between the total

Patients with positive FHx have a higher risk of sCVE

cohort with FHx data and the sub-cohort with genotyped data (Table 1).

Patients with a positive FHx (744/1,788, 41.6%) were younger and had on average more cardiovascular risk factors (Table 1). During a median follow-up of 2.9 years, 418 patients (23.4%) reached the composite endpoint of sCVE (Figure 2A) of whom 105 (5.9%) had stroke or fatal stroke, 119 (6.7%) had MI or fatal MI, 29 (1.6%) had cardiovascular death due to other causes (fatal cardiac failure, AAA rupture or sudden death) and 165 (9.2%) had a peripheral intervention or leg amputation.

Patients with positive FHx had an increased risk of sCVE compared to those without (absolute 3-year risks of 26.5% versus 21.2% respectively, hazard ratio (HR) 1.292, 95% confidence interval (CI), 1.066-1.566, p=0.009 (Figure 2A, Figure 3; Supplemental Table S3). This association remained significant after correction for confounders with adjusted HR 1.287, 95%CI 1.033-1.604, p=0.024 (Figure 3; Supplemental Table S3) and was independent of genetic predisposition as measured by MetaGRS (adjusted HR 1.397, 95%CI 1.074-1.819, p=0.013, Figure 3; Supplemental Table S3). Sex-stratified analyses confirmed results in men (with adjusted HR after correction for confounders of 1.380, 95%CI 1.068-1.783, p=0.014; adjusted HR after

correction for confounders including MetaGRS was 1.513, 95%CI 1.115-2.052, p=0.008).

However, in women the univariate association between FHx and sCVE was not significant

(unadjusted HR 1.187, 95%CI 0.822-1.171, p=0.360) but multivariable analyses could not be

performed because of limited power (Supplemental Table S3).

Patients with higher MetaGRS have a higher risk of sCVE

The MetaGRS, standardized to mean-zero and unit-variance, approximated a normal distribution

in the study population (Supplemental Figure S1). Patients in the top 20% of MetaGRS were

relatively more often females, younger of age and had less often diabetes compared to the

remaining 80% of the cohort (Table 1). Differences in baseline characteristics between high

(highest quintile of MetaGRS) and low genetic risk patients (lowest quintile of MetGRS) can be

found in the supplemental (Table S4). In the 3-year follow-up, a total of 326/1319 (24.7%)

patients reached the composite endpoint of sCVE of whom 96 (7.3%) had stroke or fatal stroke,

85 (6.4%) had MI or fatal MI, 21 (1.6%) died of other cardiovascular causes (fatal cardiac

failure, AAA rupture or sudden death) and 124 (9.4%) had a peripheral intervention or leg

amputation.

Patients in the top 20% of MetaGRS had 1.4 fold increased risk of developing sCVE within the

three years of follow-up when compared to the remaining 80% of the cohort (Figure 2B, absolute

3-year risks of 29.5% versus 23.5% respectively, HR 1.353, 95% CI 1.047-1.749, p=0.021). After

adjustment for possible confounders including FHx this association remained statistically

significant (HR for top 20% 1.345, 95%CI 1.009-1.792, p=0.043). We found similar results

when we compared patients in top 20% to the bottom 20% of MetaGRS (in univariate analysis

with HR 1.539, 95% CI,1.086-2.181, p=0.015 (Figure 2C) and for multivariable analysis adjusted

HR including FHx 1.583, 95%CI, 1.066-2.351, p=0.023), and when analysing MetaGRS as a

continuous quantity (adjusted for confounders with HR 1.150 per one SD increase in MetaGRS,

95%CI 1.022-1.293, p=0.021, adjusted HR including FHx 1.112 per one SD increase in

MetaGRS, 95% CI, 0.983-1.259, p=0.091). Results are illustrated in Figures 2B+C and 3, and

displayed in Supplemental Table S3. Confounders added to multivariable models are displayed

in Supplemental Tables S1 and S2.

Similar results were found in men (adjusted for confounders including FHx showed a HR 1.219

per one SD increase in MetaGRS, 95%CI 1.056-1.408, p=0.006, Supplemental Table S3). In

women univariate analyses showed no significant associations between MetaGRS and sCVE

(HR 0.916 per one SD increase in MetaGRS, 95% CI 0.743-1.129, p=0.413), yet multivariable

analysis was not possible due to lack of power (Supplemental Table S3).

MetaGRS is associated with vulnerable carotid plaque characteristics

To unravel possible underlying pathophysiological mechanisms of the associations between

MetaGRS, FHx and CVD, we explored the impact of FHx and the MetaGRS on atherosclerotic

plaque characteristics. We found no associations between histological plaque characteristics and

FHx in the total cohort or in women although not all multivariable analyses could be performed

(Supplemental Table S7 and S8). However carotid plaques from men with a positive FHx

contained less collagen and less SMC content compared to men with negative FHx

(Supplemental table S8). MetaGRS was associated with significantly higher overall plaque

vulnerability score (regression coefficient β of 0.070 per SD increase in MetaGRS, 95%CI,

0.003-0.137, p=0.040). To determine the plaque characteristics on which this association was

based, plaque characteristics were analysed separately. High genetic risk patients (in top 20% of

MetaGRS) more frequently had a lipid core>10% of total plaque area (adjusted odds ratio (OR), 1.591, 95% CI 1.105-2.291, p=0.013) and more macrophage infiltration (adjusted OR 1.490, 95% CI 1.118-1.986, p=0.006) compared to patients with MetaGRS in the remaining 80% (Table 2). For lipid core, we found the same association when comparing the top 20% with the bottom 20% of the MetaGRS (adjusted HR 1.887, 95% CI, 1.188-2.997, p=0.007, Supplemental Table S5). Analyses of MetaGRS as a continuous quantity confirmed the association with lipid core>10% (adjusted OR 1.171 per SD increase in MetaGRS, 95% CI 1.026-1.337, p=0.019, Table 3). Sex-stratified analyses revealed a significant association of MetaGRS with macrophages in women (adjusted OR per SD increase in MetaGRS 1.238, 95% CI, 1.007-1.521, p=0.043) whereas a significant association of MetaGRS with IPH was found in men (adjusted OR per SD increase 1.220, 95% CI 1.050-1.418, p=0.010 Supplemental Table S6).

Because IPH has been associated with increased risk of sCVE(4), we added IPH to the multivariable models of FHx, MetaGRS and sCVE to unravel whether the association between FHx, MetaGRS and sCVE could be explained by IPH. We found that all associations of FHx and MetaGRS were independent of IPH given that adding IPH to multivariable models of sCVE did not alter the effect sizes (Supplemental Table S3).

Discussion

We validated the polygenic risk score for CAD (MetaGRS) for sCVE in a severe atherosclerotic cohort of carotid artery stenosis patients undergoing carotid endarterectomy. We report four key findings. First, FHx was independently associated with an increased risk of sCVE. Second, higher MetaGRS score was independently associated with an increased risk of sCVE. Third, the association of MetaGRS with sCVE was independent of FHx and the association of FHx with sCVE was independent of MetaGRS. Fourth, high MetaGRS was associated with more vulnerable carotid atherosclerotic plaque characteristics suggesting possible underlying pathobiological mechanisms through which genetic variants could affect CVD.

Although positive FHx is a well-known risk factor for primary CVE(1, 2), previous studies assessing FHx and secondary outcome are inconsistent.(5, 7–9, 33) In patients with first-MI(7, 8, 33), studies have reported a protective effect of FHx on all-cause mortality, whereas others showed an increased risk of CVE.(5, 9) One can assume that patients with positive FHx are identified earlier as at-risk individuals through screening programs resulting in more intensive surveillance and preventive strategies leading to the benefit in overall survival.(7) Indeed, in our cohort patients with positive FHx were also younger at timing of CEA. Of note, it is known that the sensitivity of self-reported FHx can be low (50%-70%) and might therefore be an unreliable estimate.(34)

Our results are an independent validation of and consistent with a recent French Canadian study that revealed the association of MetaGRS in patients with recurrent CAD.(21) We now confirm this association in a different patient population with high prevalence of other CVD

comorbidities either in coronary or peripheral vascular beds having severe carotid artery stenosis requiring intervention. Our results therefore underscore the concept of atherosclerosis as a complex and systemic disease underlying CVD. We extend the results from the French Canadian study as we corrected for FHx and possible confounding factors such as baseline medication, medical history and cardiovascular risk factors, highlighting the independent association of MetaGRS with sCVE. Furthermore, we provide mechanistic insights by showing associations of MetaGRS with vulnerable plaque characteristics (lipid core and macrophages content) indicative of an unstable overall plaque morphology. Previous studies in CAD patients assessing other CAD-GRS and sCVE were inconsistent, but these GRSs incorporated a limited number of SNPs (varying from 19 to 47 SNPs).(35–40) The current MetaGRS is the result of a meta-analysis and therefore captures information from the full genome for CAD, possibly explaining the clear reproducibility of this MetaGRS in two CAD cohorts and our CEA cohort.(19)

The independent association of FHx and PRS has been reported frequently for primary CAD events(19, 20, 41), however only one study showed this for sCVE.(40) We confirm the independent association of MetaGRS with sCVE from FHx. Indeed, we also found that positive FHx was associated with increased risk of sCVE independent of genetic risk (MetaGRS). Several reasons could be hypothesized for the non-overlapping associations of FHx and MetaGRS with CVE. MetaGRS includes common genetic variants associated to an increased risk for CAD in the general population, whilst CVD in families may arise in part from more rare genetic mutational events, and thus a positive FHx captures individual yet family specific rare variation. Another explanation could be that FHx reflects not only genetic factors but also non-genetic factors. Although we corrected for traditional risk factors, other environmental factors that were

not taken into account such as social economic status or nutrition patterns could still be attributable to the risk of secondary events.

Exactly how genetic variants affect CVD is currently unknown. CAD-variants have been linked to pathways involved in atherosclerosis (e.g. lipid metabolism, inflammation, blood pressure and vascular remodelling), but most CAD-variants are situated outside protein-coding regions with unknown functions, making them hard to map to pathophysiological mechanisms.(13, 42) We found that MetaGRS was associated with a more rupture-prone atherosclerotic plaque displayed by a higher plaque vulnerability score caused by more fat, IPH (in men) and macrophages (predominantly in women). Moreover, carotid plaques from men with positive FHx were associated with less SMC and less collagen, whereas the association in women remains unclear. Similarly, a previous AE study showed that a CAD-PRS was correlated with more fat, whereas a large-artery stroke-PRS was correlated with more IPH and SMC.(18) Our results support the view that genetic variants could mediate their effect on CVD by influencing atherosclerotic plaque composition and morphology. Interestingly, IPH has been associated with sCVE in men independent of other clinical risk factors in our biobank (4, 43), yet we found that the associations of MetaGRS, FHx and sCVE were independent of IPH. It therefore remains unclear what the underlying mechanism is through which MetaGRS and FHx exert their increased sCVE risk. Moreover, it is shown that only small proportions (<8%) of the effect of genetic risk as well as of FHx for CAD were mediated through known metabolic pathways such as blood lipids and hypertension, whereas the majority (>80%) was not.(44) Albeit, these findings were based on a GRS consisting of only 50-CAD-SNPs. Future studies should explore exact pathophysiological mechanisms how genetic variants captured by the

MetaGRS influence plaque destabilization, primary and secondary CVE risk, for example through deep-phenotyping of atherosclerotic plaque characteristics by quantitative computerized analysis(45), mapping MetaGRS loci to specific CVE (MI, stroke or PAD), and further investigating the potential sex-differences.

Although PRS could identify high-risk patients for CVE, one could conclude that due to their unfavourable genetic risk the CVE risk is unchangeable. However, previous studies have suggested that high genetic risk is modifiable by lifestyle interventions or medication, thus not deterministic per se.(46) Among individuals with high genetic risk (top 20% of CAD-GRS), those adhering to a healthy lifestyle had lower relative risk of a first-CAD event compared to those with an unfavourable lifestyle. (46) Moreover, both in primary or secondary prevention trials, patients with high genetic risk (those in top 20% of CAD-GRS) had a greater absolute and relative risk reduction of (recurrent) MI or coronary death than those with lower genetic risk despite equal LDL level reductions. (40, 47) In the current era of evolving new expensive drugs (e.g. PCSK9-inhibitors or new anti-inflammatory drugs such as IL1b-blockers(48)) and the increasing burden of CVD on the sustainability of the healthcare system, the cost-effectiveness ratio becomes important. In the future, it may be useful to investigate whether the efficacy of such medications varies across genetic risk categories for selecting those benefit most and qualify for these add-on therapies.

Admittedly, our study has several limitations. First, although our results may suggest interesting sex-differences as we only observed the independent association of MetaGRS and FHx for sCVE in men, the association in women remains unclear because we were

underpowered for multivariable analysis. Second, we did not have data regarding medication use during follow-up nor therapy compliance. These factors could have interfered with observed sCVE rates. Second, owing to limited power we were unable to assess associations with separate CVE or determine the predictive value of MetaGRS above clinical risk factors. Last, most included patients are of European ancestries and generalizability to other ethnicities needs further attention. Yet, the Athero-Express Biobank is unique in its scope and its major strengths are 1) the unique population that is relatively unexplored in the field of FHx and PRS, 2) the prospective study design with detailed data on patient characteristics (including FHx), and 3) the extensive data on plaque composition and morphology that enable us to identify putative pathological mechanisms.

In the future, PRS may be a useful tool for personalized risk prediction for primary or secondary CVE. Adding MetaGRS to a model with traditional risk factors improved prediction of first-CAD events, although model improvement was modest.(19) The clinical utility of PRS for sCVE is still unknown because the incremental value of PRS above clinical factors still needs to be established. One study suggested an additive predictive value of PRS above clinical factors but did not include FHx (38), whereas other studies failed to demonstrate this.(35, 36, 39) The power of these studies may have been limited due to limited number of CVE. Pooling data of several cohorts including detailed data on preventive strategies and medications during follow-up together with use of uniform outcome definitions for sCVE and uniform PRS composition could help elucidate the clinical value of PRS, for example within international collaborations such as the GENIUS-CHD Consortium.(49) Furthermore, possible sex-differences in the role of risk prediction with PRS need to be further elucidated.

In conclusion, both higher MetaGRS and positive FHx were independently associated

with increased risk of sCVE in carotid artery stenosis patients undergoing CEA. Higher

MetaGRS was also associated with more vulnerable carotid plaque characteristics indicating

possible underlying mechanism how genetic variants influence CVD. PRS could identify high-

risk individuals and may help selecting future study populations when investigating new

therapeutic CVD prevention strategies.

Conflict of interest statement:

none

Author contributions

Study design: NT, SH, SWvdL

Data collection: NT, SH, SWvdL, GP

Data analysis and interpretation: NT, SH, SWvdL

Drafting the article and figures: NT, SH, SWvdL

Critical revision and final approval: NT, SH, SWvdL, DPVdK, GJB, HMdR, FWA, GP

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Figure legends

Central illustration. Family history and genetic risk score (MetaGRS) are independently

associated with secondary cardiovascular events (sCVE) within 3-years after carotid

endarterectomy (CEA).

In the figure the hazard ratio's (HR) of the Cox-regression analyses are displayed for different

multivariable models for the associaton with sCVE.

Figure 1. Flowchart.

Figure 2. Kaplan-Meier graphs for 3-year risk of sCVE after CEA.

2A. Patients with positive FHx compared to those without.

2B. Patients in the top 20% of MetaGRS compared to the rest of the patients (remaining 80%).

2C. Patients in the top 20% of MetaGRS compared to those in the bottom 20% of MetaGRS.

Figure 3. Cox-regression analyses of FHx and MetaGRs for sCVE after CEA.

Hazard ratio's (HR) and 95% confidence intervals (95% CI) for the different univariate and

multivariable Cox-regression models of FHx or MetaGRS for sCVE. The HR for MetaGRS as a

continuous quantity indicates the HR per one SD increase in MetaGRS.

UV, univariate model. The univariate model for MetaGRS included age, sex, PC1-4 and

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genotype array.

MV, multivariable model.

The multivariable model for FHx was corrected for traditional risk factors (age, sex, hypercholesterolemia, diabetes, hypertension, BMI and smoking) and additional confounders (history of CAD, history of PAD, cerebrovascular symptoms and eGFR).

The univariate model for MetaGRS included age, sex, PC1-4 and genotype array. For multivariable analyses of MetaGRS traditional risk factors (hypercholesterolemia, diabetes, hypertension, BMI and smoking) were added.

Tables

| | Cohort with data on FHx | | | | Sub cohort with genotype data | | | |
|-----------------------------------|-------------------------|-----------------|-----------------|------------------------------|-------------------------------|--------------------------|-------------|---------------------|
| | Total | Negative FHx | Positive FHx | p-value | Total | Remaining 809 MetaGRS | MetaGRS | p-value |
| | (n=1,788) | (n=1,044) | (n=744) | 12 1 | (n=1,319) | (n=1,055) | (n=264) | |
| Age (mean, SD) | 68.9 [9.1] | 70.1 [9.0] | 67.1 [8.9] | 2.754 x10 ⁻¹² * b | 68.7 [9.3] | 69.1 [9.1] | 67.1 [9.7] | 0.002* b |
| Male | 1244 (69.6) | 745 (71.4) | 499 (67.1) | 0.052 ^b | 914 (69.3) | 748 (70.9) | 166 (62.9) | 0.012* b |
| BMI (mean, SD) | 26.4 [3.9] | 26.2 [3.7] | 26.8 [4.2] | 0.002* b | 26.3 [3.8] | 26.2 [3.6] | 26.7 [4.3] | 0.103 ^b |
| GFR (mean, SD) | 73 [20.8] | 72.6 [21.0] | 74.1 [20.5] | 0.146 ^b | 72.3 [20.1] | 72.2 [20.1] | 72.7 [20.2] | 0.744 |
| Smoking | 604 (34.0) | 341 (33.0) | 263 (35.4) | 0.286 | 456 (35.2) | 359 (34.6) | 97 (37.5) | 0.387 |
| Diabetes | 390 (21.8) | 220 (21.1) | 170 (22.8) | 0.370 | 303 (23.0) | 255 (24.2) | 48 (18.2) | 0.039* ^b |
| Hypertension | 1,306 (74.4) | 728 (71.4) | 578 (78.4) | 0.001 * ^b | 921 (72.2) | 729 (71.8) | 192 (74.1) | 0.445 |
| Hypercholesterolemia | 1,162 (65.0) | 637(64.0) | 525 (73.9) | 1.500 x10 ⁻⁵ * b | 806 (65.9) | 634 (65.1) | 172 (69.1) | 0.237 |
| History of CAD | 535 (29.9) | 248 (23.8) | 287 (38.6) | 1.327 x10 ⁻¹¹ * b | 389 (29.5) | 306 (29.0) | 83 (31.6) | 0.422 |
| History of stroke | 574 (32.1) | 365 (35.0) | 209 (28.1) | 0.002* b | 437 (33.1) | 348 (33.0) | 89 (33.7) | 0.823 |
| History of PAD | 355 (19.9) | 185 (17.7) | 170 (22.9) | 0.006* b | 270 (20.5) | 216 (20.5) | 54 (20.5) | 0.989 |
| Stenosis ipsilateral | | | | 0.083 ^b | | | | 0.990 |
| Ф 50-70% | 132 (7.5) | 70 (6.9) | 62 (8.5) | | 83 (6.4) | 66 6.4) | 17 (6.6) | |
| Ф 70-99% | 1609 (91.9) | 941 (92.3) | 668 (91.4) | | 1200 (93.2) | 962 (93.2) | 238 (93.0) | |
| Contralateral stenosis of 50-100% | 714 (39.9) | 406 (43.1) | 308 (45.4) | 0.362 | 543 (45.4) | 425 (44.4) | 118 (49.4) | 0.168 ^b |
| Presenting symptoms | | | | 0.010* b | | | | 0.497 |
| Φ Asymptomatic | 228 (12.8) | 120 (11.6) | 108 (14.6) | | 178 (13.5) | 135 (12.8) | 43 (16.3) | |
| Φ Ocular | 302 (16.9) | 174 (16.8) | 128 (17.3) | | 203 (15.4) | 163 (15.5) | 40 (15.2) | |
| Ф ТІА | 774 (43.3) | 437 (42.2) | 337(45.5) | | 580 (44.1) | 466 (44.3) | 114 (43.3) | |
| Φ Stroke | 472(26.4) | 304 (29.4) | 168 (22.7) | | 354 (26.9) | 288 (27.4) | 66 (25.1) | |
| Total cholesterol (median, IQR) | 4.4 [1.6] | 4.4 [1.6] | 4.4 [1.7] | 0.140 ^b | 4.4 [1.6] | 4.4 [1.6] | 4.4 [1.7] | 0.716 |
| LDL cholesterol (median, IQR) | 2.4 [1.3] | 2.4 [1.3] | 2.5 [1.4] | 0.101 ^b | 2.4 [1.3] | 2.4 [1.3] | 2.5 [1.4] | 0.330 |
| HDL cholesterol (median, IQR) | 1.1 [0.4] | 1.1 [0.4] | 1.1 [0.4] | 0.785 | 1.1 [0.4] | 1.1 [0.4] | 1.1 [0.4] | 0.384 |

| Triglycerides (median, IQR) | 1.5 [1.0] | 1.4 [0.9] | 1.6 [1.0] | 0.001* b | 1.5 [1.0] | 1.5 [1.0] | 1.5 [0.9] | 0.491 |
|-----------------------------|--------------|------------|------------|----------|-------------|------------|------------|-------|
| Lipid lowering drug use | 1,379 (77.1) | 779 (74.6) | 600 (80.8) | 0.002* b | 999 (75.9) | 795 (75.5) | 204 (77.3) | 0.547 |
| Antiplatelet drug use | 1,583 (88.5) | 926 (88.8) | 657(88.5) | 0.876 | 1164 (88.5) | 928 (88.3) | 236 (89.4) | 0.617 |
| Surgery year | | | | 0.612 | | | | 0.820 |
| Ф 2002-2003 | 215 (12) | 129 (11.6) | 86 (11.6) | | 192 (14.6) | 151 (14.3) | 41 (15.5) | |
| Ф 2004-2005 | 327 (18.3) | 197 (18.9) | 130 (17.5) | | 307 (23.3) | 243 (23.0) | 64 (24.2) | |
| Ф 2006-2007 | 288 (16.1) | 156 (14.9) | 132 (17.7) | | 270 (20.5) | 214 (20.3) | 56 (21.2) | |
| Ф 2008-2009 | 241 (13.5) | 146 (14.0) | 95 (12.8) | | 232 (17.6) | 185 (17.5) | 47 (17.8) | |
| Ф 2010-2011 | 265 (14.8) | 155 (14.8) | 110 (14.8) | | 224 (17.0) | 182 (17.3) | 42 (15.9) | |
| Ф 2012-2013 | 260 (14.5) | 157 (15.0) | 103 (13.8) | | 94 (7.1) | 80 (7.6) | 14 (5.3) | |
| Ф 2014-2015 | 154 (8.6) | 82 (7.9) | 72 (9.7) | | - | - | | |
| Ф 2016 | 38 (2.1) | 22 (2.1) | 16 (2.2) | | - | - | | |

Table 1. Baseline characteristics.

Values are displayed as n (%), unless otherwise specified. Values indicated as bold* are statistically significant. Values below p<0.20 are indicated by ^b. *BMI*, Body mass index; *CAD*, coronary artery disease; *PAD*, peripheral artery disease; *eGFR*, estimated glomerular filtration rate; *TIA*, transient ischemic attack; *SD*, standard deviation. When multiple cerebrovascular symptoms occurred in the six months prior to the operation, the most serious symptom counts in the following order: stroke> TIA> ocular. Ocular symptoms include transient or permanent retinal ischemia.

History of any stroke includes ipsilateral or contralateral stroke. Antiplatelet drug use includes the use of aspirin, dipyridamole or any ADP-inhibitor. Lipid lowering drug use includes any lipid-lowering drug. Diabetes, hypercholesterolemia and hypertension were defined as diagnosed by a medical doctor or medication use for the specific comorbidity.

| Categorical plaque characteristics | 80% of MetaGRS (n=1,040) | Top 20% of MetaGRS (n=261) | Unadjusted OR (95% CI) | Unadjusted p-value | Adjusted OR (95% CI) | Adjusted p-value |
|---|--------------------------------|----------------------------------|-----------------------------|-----------------------|---------------------------|---------------------|
| Moderate/heavy calcifications | 525/1,038 (50.6) | 135/261 (51.7) | 1.047 (0.798-1.374) | 0.741 | 1.001 (0.754-1.330) | 0.995 |
| Moderate/heavy collagen | 826/1,039 (79.5) | 211/261 (80.8) | 1.088 (0.772-1.534) | 0.629 | 1.091 (0.755-1.577) | 0.644 |
| Presence of lipid core >10% | 752/1,040 (72.3) | 209/261 (80.1) | 1.539 (1.104-2.147) | 0.011* | 1.591 (1.105-2.291) | 0.013* |
| Presence of IPH | 624/1,039 (60.1) | 160/261 (61.3) | 1.054 (0.798-1.392) | 0.713 | 1.112 (0.821-1.506) | 0.493 |
| Moderate/heavy smooth muscle cells | 736/1,037 (71.0) | 185/260 (71.2) | 1.009 (0.747-1.362) | 0.954 | 0.908 (0.652-1.265) | 0.570 |
| Moderate/heavy macrophages | 545/1,036 (52.6) | 161/260 (61.9) | 1.465 (1.109-1.936) | 0.007* | 1.490 (1.118-1.986) | 0.006* |
| Continuous plaque characteristics | | | Unadjusted beta (95% CI) | Unadjusted p-value | Adjusted beta (95% CI) | Adjusted p-value |
| Number of macrophages (median, IQR) | 0.3 [1.0] | 0.5 [1.1] | 0.058 [-0.006-0.123] | 0.077 | 0.055 (-0.012-0.121) | 0.107 |
| Number of smooth muscle cells (median, IQR) | 1.4 [2.3] | 1.3 [2.3] | -0.010 [-0.095-0.076] | 0.822 | -0.056 (-0.143-0.031) | 0.209 |
| Microvessels (median, IQR) | 7.0 [7.0] | 7.3 [7.0] | 0.088 [-0.021-0.197] | 0.114 | 0.072 (-0.037-0.182) | 0.196 |
| Plaque vulnerability score | 2.0 [2.0] | 3 [1] | 0.216 (0.048-0.385) | 0.012* | 0.198 (0.003-0.364) | 0.004* |

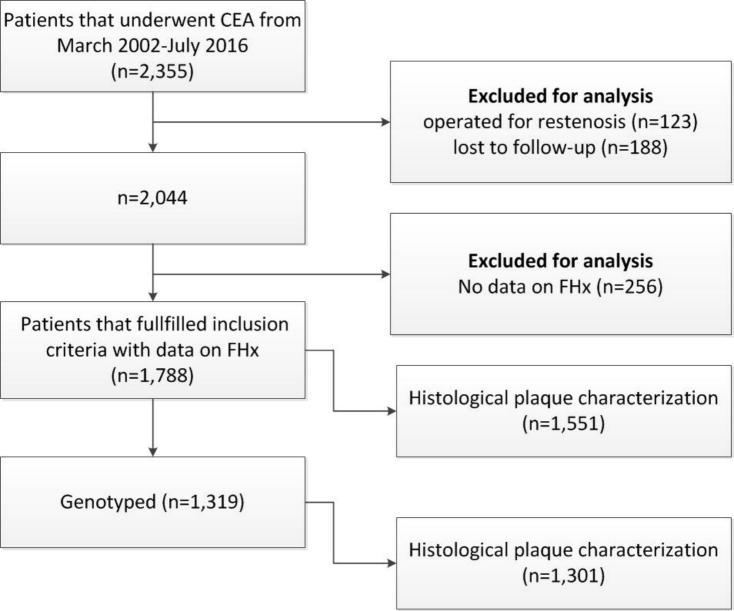
Table 2. Associations of carotid plaque characteristics from patients in top 20% of MetaGRS compared to the rest of the patients (80% of MetaGRS).

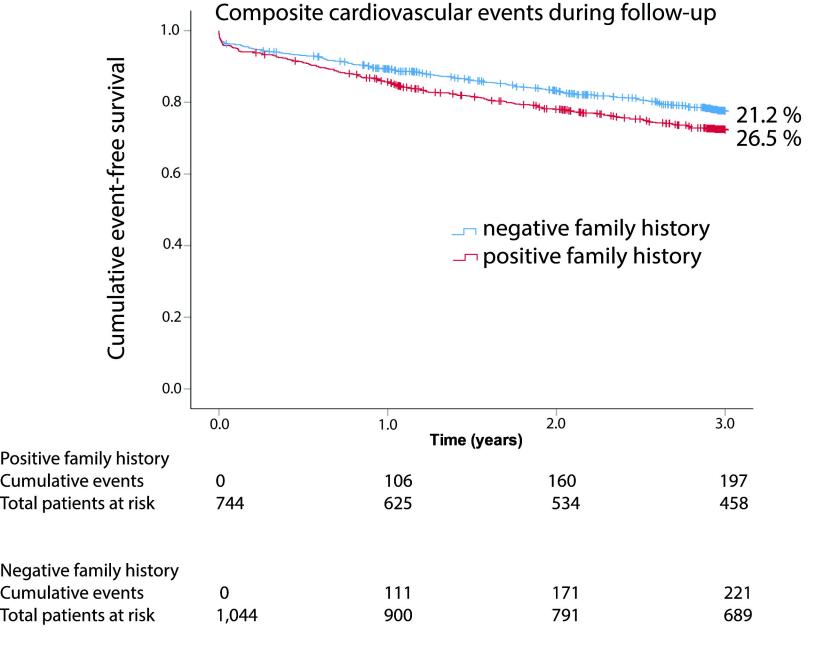
Values are displayed as n/total (%), unless otherwise specified. Values indicated as bold* are statistically significant. *OR*, odds ratio; *IPH*, intraplaque haemorrhage. Unadjusted OR indicates univariate analysis, adjusted OR indicates multivariable analysis. Multivariable analyses were corrected for age, sex, surgery year, cerebrovascular symptoms and additional confounders showing p<0.20 association with MetaGRS and the plaque characteristics of interest. Confounders included in various multivariable models can be found in Supplemental Table S2.

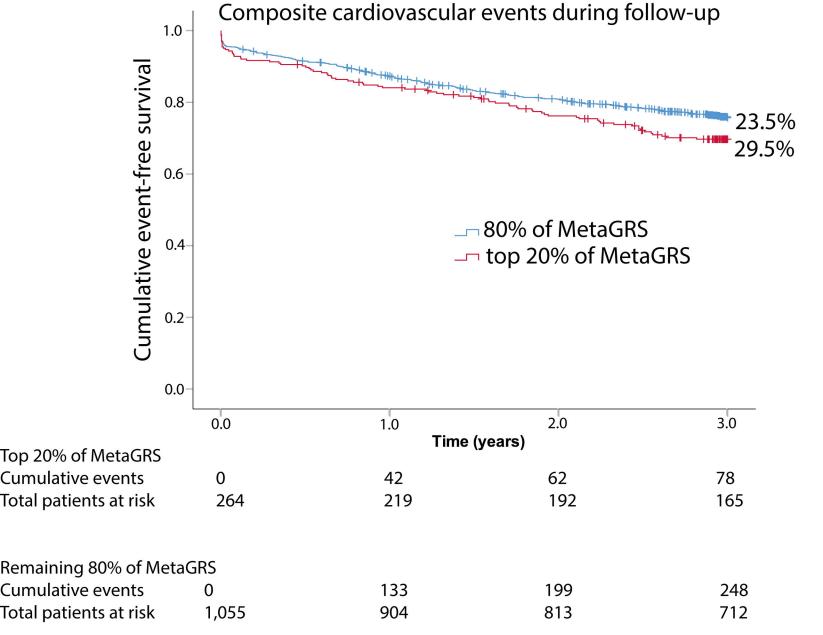
| Categorical plaque characteristics | Unadjusted OR (95% CI) | Unadjusted p-value | Adjusted OR (95% CI) | Adjusted p-value |
|---|-----------------------------|-----------------------|---------------------------|---------------------|
| Moderate/heavy calcifications | 0.977 (0.874-1.092) | 0.677 | 0.940 (0.826-1.070) | 0.348 |
| Moderate/heavy collagen | 1.035 (0.900-1.190) | 0.632 | 1.064 (0.919-1.231) | 0.408 |
| Presence of lipid core >10% | 1.148 (1.011-1.305) | 0.034* | 1.171 (1.026-1.337) | 0.019* |
| Presence of IPH | 1.101 (0.981-1.235) | 0.101 | 1.126 (0.999-1.270) | 0.052 |
| Moderate/heavy smooth muscle cells | 1.008 (0.889-1.143) | 0.899 | 1.004 (0.880-1.145) | 0.952 |
| Moderate/heavy macrophages | 1.095 (0.978-1.225) | 0.115 | 1.103 (0.983-1.237) | 0.095 |
| Continuous plaque characteristics | Unadjusted beta (95% CI) | Unadjusted p-value | Adjusted beta (95% CI) | Adjusted p-value |
| Number of macrophages (median, IQR) | 0.011 (-0.016-0.037) | 0.419 | 0.010 (-0.015-0.036) | 0.436 |
| Number of smooth muscle cells (median, IQR) | -0.007 (-0.041-0.027) | 0.647 | -0.004 (-0.038-0.031) | 0.837 |
| Microvessels (median, IQR) | 0.035 (-0.010-0.080) | 0.124 | 0.037 (-0.006-0.080) | 0.091 |
| Plaque vulnerability score | 0.072 (0.004-0.140) | 0.039* | 0.070 (0.003-0.137) | 0.040* |

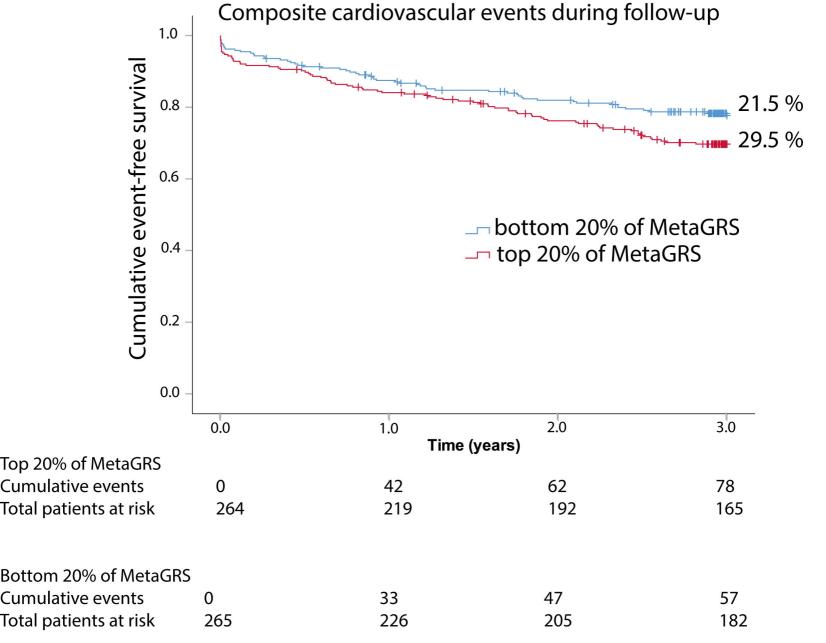
Table 3. Associations of MetaGRS (as continuous measure) and histological carotid plaque characteristics.

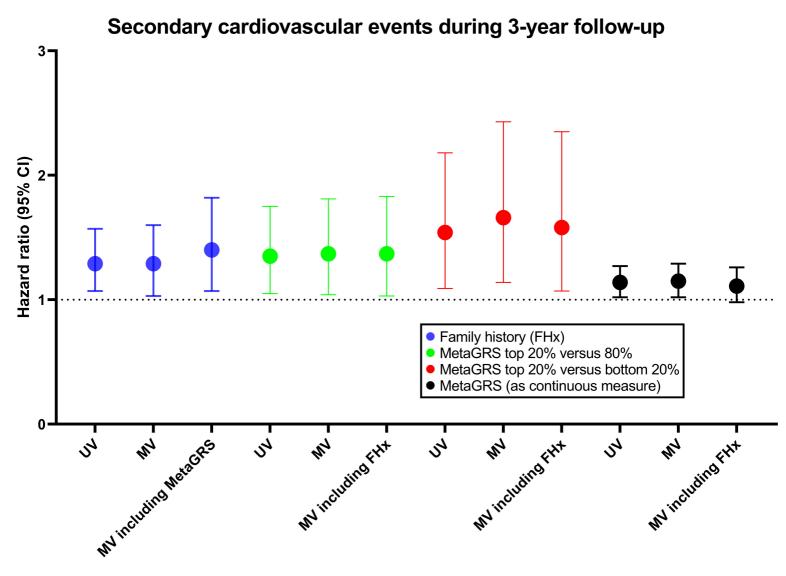
Values are displayed as n/total (%), unless otherwise specified. Values indicated as bold* are statistically significant. MetaGRS was standardized to mean-zero and unit-variance. The odds ratio (OR) indicates OR per one SD increase in MetaGRS. *IPH*, intraplaque haemorrhage. Unadjusted OR indicates univariate analysis, adjusted OR indicates multivariable analysis. Multivariable analyses were corrected for age, sex, surgery year, cerebrovascular symptoms and additional confounders showing p<0.20 association with MetaGRS and the plaque characteristics of interest. Confounders included in various multivariable models can be found in Supplemental Table S2.











Distribution of MetaGRS

