Supplemental Materials

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SUPPLEMENTARY METHODS

Study design and data sources

We performed two-sample Mendelian randomization analyses, taking genetic associations with lipid concentrations from one dataset, and genetic associations with cardiovascular disease outcomes from an independent dataset. Two-sample Mendelian randomization was performed in preference to one-sample Mendelian randomization using lipid measurements in UK Biobank to avoid the possibility of false positive findings arising due to weak-instrument bias, which can occur in a one-sample analysis. 1

We obtained genetic associations with lipid concentrations (LDL-cholesterol, HDLcholesterol, and triglycerides) from the Global Lipids Genetics Consortium (GLGC) on up to 188,577 participants of European ancestries. $2,3$ $2,3$ We decided to use variants and associations from the 2013 version of GLGC results, and not the later exome release,⁴ to make sure that both exonic and intronic variants were treated consistently in the analysis. Genetic associations were estimated with adjustment for age, sex, and genomic principal components within each participating study after inverse rank quantile normalization of lipid concentrations, and then meta-analysed across studies. Inverse rank quantile normalization means that associations are not affected by the skewed distribution of variables, in particular for triglycerides. The triglyceride measurement mostly represents the triglyceride content of triglyceride-rich lipoproteins.

We estimated genetic associations with disease outcomes from UK Biobank, a cohort of 502,682 participants (94% of self-reported European ancestries) recruited between 2006-2010 in 22 assessment centres throughout the UK and followed-up until 31st

March 2017 or their date of death (recorded until 14th February 2018).⁵ In addition to standard genetic and individual-level quality control procedures (call rate $\geq 99\%$, info score > 0.9, Hardy-Weinberg equilibrium p-value $\geq 10^{-5}$, removal of genetic sex mismatches), ⁶ we excluded participants having non-European ancestries (self-report or inferred by genetics) or excess heterozygosity (>3 standard deviations from the mean). We included only one of each set of related participants (third-degree relatives or closer). The sample for our analyses comprised 367,703 unrelated participants of European ancestries.

We defined 19 outcomes based on electronic health records (ICD-9 or ICD-10 diagnosis and hospital procedure codes) from hospital episode statistics and death certificates, and self-reported information validated by interview at baseline with a trained nurse. With respect to study baseline, we included both prevalent and incident events in our outcome definitions. However, with respect to the genetic variants, all events are incident. For presentation, we divide outcomes into five categories: 1) ischaemic cerebrovascular diseases (ischaemic stroke, transient ischaemic attack, combined ischaemic cerebrovascular event [i.e. ischaemic stroke or transient ischaemic attack]), 2) haemorrhagic stroke (intracerebral haemorrhage, subarachnoid haemorrhage, combined haemorrhagic stroke); 3) aneurysms (any aortic aneurysm, abdominal aortic aneurysm, thoracic aortic aneurysm); 4) thromboembolic diseases (any venous thromboembolism, pulmonary embolism, deep vein thrombosis); and 5) other cardiovascular diseases (hypertension, peripheral vascular disease, aortic valve stenosis, atrial fibrillation, and heart failure). Additionally, we included CAD as a positive control outcome and chronic kidney disease as a negative control outcome as it is thought to be influenced by some cardiovascular risk factors (such as hypertension and diabetes) but not dyslipidaemia.⁷ Precise outcome definitions are given in

Supplementary Table S14. For composite outcomes (such as any venous thromboembolism), if an individual had both the constituent outcomes (here, deep vein thrombosis and pulmonary embolism), then they were included as a case in each analysis. To obtain genetic association estimates for each outcome, we conducted logistic regression with adjustment for age at recruitment, sex and 10 genomic principal components using the *snptest* program.⁸ These summarized data are required to combine evidence across datasets in a two-sample Mendelian randomization analysis.

Exploratory analyses

To validate unexpected results in the primary Mendelian randomization analyses, we performed additional exploratory analyses for relevant continuous traits. To provide further evidence on the link between triglycerides and thromboembolism, we performed multivariable Mendelian randomization analyses with tissue-type plasminogen activator and platelet count as outcomes. Tissue-type plasminogen activator was chosen as it was associated with triglycerides in a previous observational study among 1227 men free of coronary heart disease.⁹ Genetic associations with tissue-type plasminogen activator were obtained from a genome-wide association study (GWAS) of 83 proteins in 3,394 participants of European ancestries.¹⁰ Platelet count was chosen as excess thrombocytopenia events were observed in the treatment arm of the APPROACH trial of the APOC3 inhibitor volanesorsen. ¹¹ Genetic associations with platelet count were obtained from a GWAS of blood cell traits in up to 173,480 participants of European ancestries. ⁶ To provide further evidence on the links between HDL-cholesterol and triglycerides with hypertension, we performed multivariable Mendelian randomization analyses with systolic and diastolic blood

pressure as outcomes. Genetic associations with blood pressure were obtained from up to 757,601 European-ancestry participants.¹²

Statistical analysis

We carried out analyses based on 184 genetic variants previously demonstrated to be associated with at least one of LDL-cholesterol, HDL-cholesterol or triglycerides at a genome-wide level of significance ($p < 5 \times 10^{-8}$) in the GLGC. Only one variant per gene region was included in the analysis, except for a small number of regions where a conditionally independent signal was found. We accounted for correlations between variants in the analysis methods, except for the weighted median method, where we further pruned to only include one variant per gene region, and correlations were derived in 502 European descent participants from the 1000 Genomes Project Phase 3. These variants explained 13.7% of the variance in HDL-cholesterol, 14.6% in LDLcholesterol, and 11.7% in triglycerides in GLGC. One variant (rs2954022) was omitted from analyses as it was triallelic in UK Biobank but all other variants were well-imputed (info score >0.9) and were retained.

To obtain the associations of genetically-predicted values of these lipids with each cardiovascular outcome while accounting for measured genetic pleiotropy via other major lipids, we performed multivariable Mendelian randomization analyses.¹³ We implemented this method by weighted regression of the genetic associations with the outcome on the genetic associations with the three predictors, while fixing the intercept to be zero. For sensitivity analysis, we also performed (i) univariable Mendelian randomization using the inverse-variance weighted method for each lipid fraction based on all variants associated with that lipid fraction at a genome-wide level of significance,¹⁴ (ii) MR-Egger regression to account for unmeasured pleiotropy,¹⁵ and

(iii) weighted median regression to assess robustness to invalid genetic instruments.¹⁶ To account for between-variant heterogeneity, we used multiplicative random-effects models in all analyses. Heterogeneity statistics for the study outcomes are reported in Supplementary Tables S2 and S3. All Mendelian randomization estimates are expressed per 1 standard deviation increase in the corresponding lipid fraction in GLGC (1 standard deviation was 39.0 mg/dL for LDL-cholesterol, 15.8 mg/dL for HDLcholesterol, and 90.5 mg/dL for triglycerides).

For cardiovascular diseases for which it is possible that effect of lipid fractions on that outcome may act via other conditions (e.g. via CAD for heart failure), we also performed multivariable Mendelian randomization analyses adjusted for the hypothesised mediating condition. This analysis estimates the direct effects of the lipid fractions on the outcome that are not mediated via CAD. The coefficient for CAD represents the odds ratio for the outcome per unit increase in the beta-coefficient for CAD (i.e. per unit increase in log odds ratio for CAD) Additionally, we performed multivariable Mendelian randomization analyses for these outcomes excluding participants with a CAD diagnosis, to assess whether associations persisted in participants without comorbid CAD. To assess the effect of triglycerides on molecular traits related to thrombosis, we searched publicly-available genetic associations from PhenoScanner [\(http://www.phenoscanner.medschl.cam.ac.uk/\)](http://www.phenoscanner.medschl.cam.ac.uk/),¹⁷ a curated database of over 60 billion genetic associations, and performed multivariable Mendelian randomization using these traits as outcomes. Finally, we performed replication and meta-analysis of the most surprising finding of this study through collaborating with the INVENT consortium. ¹⁸ In this replication, we requested exclusion of non-European-ancestry participants and of participants from the UK

Biobank, and performed the same multivariable Mendelian randomization methods applied in the main analyses.

We also performed gene-specific analyses for variants in specific gene regions that can be considered as proxies for existing or proposed lipid-lowering therapies (since the protein products of these genes are the pharmacological targets of these drugs) and that explain enough variance in lipid concentration to result in adequately powered analyses (at least 0.4% for at least one lipid measure). We considered the following gene regions: *HMGCR* (proxy for statin treatment), *PCSK9* (proxy for PCSK9 inhibitors), *LDLR* (proxy for inhibition of the LDL receptor), *APOC3* (proxy for APOC3 inhibitors), and *LPL* (proxy for lipoprotein lipase inhibition). For each gene region, we selected genetic variants that were not strongly correlated $(R^2<0.8)$ with one another and that were previously associated with LDL-cholesterol in a conditional analysis (Supplementary Table S15). All variants are within 500kb of the gene. We obtained univariable Mendelian randomization estimates using the inverse-variance weighted method while accounting for correlations between variants.¹⁴ Variants in each gene region explained 0.4% (*HMGCR*), 1.2% (*PCSK9*), 1.0% (*LDLR*), 0.1% (*APOC3*) and less than 0.1% (*LPL*) of the variance in LDL-cholesterol in GLGC. The *APOC3* and *LPL* variants also explained 1.0% and 0.9% of the variance in triglycerides respectively. The power of the gene-specific analyses differs considerably owing to the different proportion of variance explained. Gene-specific estimates are expressed per 1 standard deviation increase in LDL-cholesterol for the *HMGCR*, *PCSK9*, and *LDLR* regions, and per 1 standard deviation increase in triglycerides for the *APOC3* and *LPL* regions.

We carried out all analyses using R (version 3.4.4) unless otherwise stated. All pvalues presented are two-sided. To account for multiple testing, we use a Bonferroni-

corrected threshold of $p < 0.05/19 = 0.003$ for statistical significance. P-values between 0.003 and 0.05 are described as "nominally significant".

SUPPLEMENTARY RESULTS

Supplementary Table S1: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) from polygenic multivariable Mendelian randomization analyses for all lipid-related variants. Estimates with *p* < 0.05 are reported in **bold**.

Supplementary Table S2: Heterogeneity statistics for the main multivariable analyses.

Supplementary Table S3: Heterogeneity statistics and p-values for the main multivariable analyses from calculated standard inverse-variance weighted method with unmodified weights, and calculated using MR-PRESSO package (bootstrapped p-values).

Supplementary Figure S1: Scatter plots of genetic associations with lipid fraction (risk factor) and cardiovascular disease (outcome) for selected risk factor/outcome pairs. Genetic associations with the risk factor are in standard deviation units. Genetic associations with the outcome are log odds ratios. The line through the origin represents the estimate from the relevant univariable inverse-variance weighted method.

Supplementary Table S4: Associations of genetically-predicted triglyceride levels with traits relating to thrombosis.

Beta coefficients are scaled to a standard deviation (SD) change in the trait per 1 SD increase in genetically-predicted triglyceride levels. Estimates were obtained from multivariable Mendelian randomization accounting for genetic associations with LDLcholesterol and HDL-cholesterol.

Supplementary Table S5: Replication of the associations between lipids and hypertension, using publicly available data on systolic and diastolic blood pressure in up to 757,601 participants of European ancestries¹²

*: $p < 0.05$ †: $p < 0.003$ ‡: $p < 10^{-8}$

Supplementary Figure S2: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) for HDL-cholesterol from genome-wide Mendelian randomization analyses for all lipid-related variants associated with the target lipid fraction.

Supplementary Table S6: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) for HDL-cholesterol from polygenic univariable Mendelian randomization analyses for all lipid-related variants associated with the target lipid fraction. Estimates with *p* < 0.05 are reported in **bold**.

Supplementary Figure S3: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) for LDL-cholesterol from genome-wide Mendelian randomization analyses for all lipid-related variants associated with the target lipid fraction.

Supplementary Table S7: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) for LDL-cholesterol from polygenic univariable Mendelian randomization analyses for all lipid-related variants associated with the target lipid fraction. Estimates with *p* < 0.05 are reported in **bold**.

Supplementary Figure S4: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) for triglycerides from genome-wide Mendelian randomization analyses for all lipid-related variants associated with the target lipid fraction.

Supplementary Table S8: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) for triglycerides from polygenic univariable Mendelian randomization analyses for all lipid-related variants associated with the target lipid fraction. Estimates with *p* < 0.05 are reported in **bold**.

Supplementary Table S9: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) from polygenic multivariable Mendelian randomization analyses for all lipid-related variants for heart failure in participants without a CAD diagnosis.

Supplementary Table S10: Multivariable Mendelian randomization estimates (odds ratio with 95% confidence interval) for hypertension, abdominal aortic aneurysm and haemorrhagic stroke without and with adjustment for body mass index (BMI) and type 2 diabetes (T2D), and for venous thromboembolism with and without adjustment for heart failure (HF).

*: $p < 0.05$ \uparrow : $p < 0.003$ \downarrow : $p < 10^{-8}$

The BMI/T2D column indicates the odds ratio representing the association between genetically-predicted body mass index (per SD increase) or type 2 diabetes (per unit increase in log odds ratio) and the outcomes of interest (i.e. hypertension, abdominal aortic aneurysm, or haemorrhagic stroke) while accounting for the genetic associations of the lipid fractions in a multivariable analysis. If the regression coefficients for the lipid measurements remain unchanged on adjustment for BMI/T2D, then the effects of lipids on the outcome do not operate via BMI/T2D. If the regression coefficients attenuate to the null, then the effects of lipids on the outcome are entirely mediated via BMI/T2D.

A similar interpretation applies for the HF column for venous thromboembolism, which indicates the odds ratio between heart failure (per unit increase in log odds ratio) and venous thromboembolism while accounting for the genetic effects of circulating lipids on venous thromboembolism. This analysis shows that heart failure is confirmed as an independent risk factor for venous thromboembolism, but does not appear to mediate the effect of triglycerides on venous thromboembolism.

Supplementary Table S11: Estimates (odds ratio per 1 standard deviation increase in lipid fraction and 95% confidence interval) from gene-specific Mendelian randomization analyses for four specified gene regions. Estimates with *p* < 0.05 are reported in **bold**. Estimates are calibrated per standard deviation increase in LDL-cholesterol for *HMGCR*, *PCSK9*, and *LDLR* regions, and per standard deviation increase in triglycerides for *APOC3* and *LPL* regions.

Supplementary Table S12: P-values from gene-specific Mendelian randomization analyses for five specified gene regions.

Supplementary Table S13: Comparison of main findings from this study with

evidence from external genetic studies and clinical trials

PMID, Pubmed study identifier. LDL-C, low-density lipoprotein. TG, triglycerides.

PCSK9i, proprotein convertase subtilisin-kexin type 9 ihibitor.

EPA, eicosapentaenoic acid. NA, not available.

In the interest of clarity, this table includes only findings that were at least nominally significant in the main multivariable Mendelian randomization analyses and were confirmed in subsequent sensitivity analyses.

Supplementary Table S14: Summary of disease outcomes considered: sources of information

Note that .X means that all sub-codes are matched. Abbreviations: ICD = International Classification of Disease; OPCS = Office of Population Censuses and Surveys Classification of Surgical Operations and Procedures; TIA = transient ischaemic attack

* Health condition diagnosed by doctor (6150/6152) and Medication for health condition (6177) were self-reported from touchscreen; Non-cancer illness code (20002) and Surgical operation code (20004) were self-reported from interview with trained nurse.

Supplementary Table S15: Genetic variants included in gene-specific analyses for each

region.

Genetic associations (beta and standard error [SE]) are with LDL-cholesterol for the *HMGCR*, *PCSK9*, and *LDLR* gene regions, and with triglycerides for the *APOC3* and *LPL* gene regions. They are given in standard deviation units and are taken from the Global Lipids Genetics Consortium (see [http://csg.sph.umich.edu/willer/public/lipids2013/\)](http://csg.sph.umich.edu/willer/public/lipids2013/).

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