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Determination of genetic correlation between tobacco smoking and coronary artery disease

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Backgrounds: Tobacco smoking is an important risk factor for coronary artery disease (CAD), but the genetic mechanisms linking smoking to CAD remain largely unknown.

Methods: We analyzed summary data from the genome-wide association study (GWAS) of the UK Biobank for CAD, plasma lipid concentrations (n = 184,305), and smoking (n = 337,030) using different biostatistical methods, which included LD score regression and Mendelian randomization (MR).

Results: We identified SNPs shared by CAD and at least one smoking behavior, the genes where these SNPs are located were found to be significantly enriched in the processes related to lipoprotein metabolic, chylomicron-mediated lipid transport, lipid digestion, mobilization, and transport. The MR analysis revealed a positive correlation between smoking cessation and decreased risk for CAD when smoking cessation was considered as exposure (p = 0.001), and a negative correlation between the increased risk for CAD and smoking cessation when CAD was considered as exposure (p = 2.95E-08). This analysis further indicated that genetic liability for smoking cessation increased the risk of CAD.

Conclusion: These findings inform the concomitant conditions of CAD and smoking and support the idea that genetic liabilities for smoking behaviors are strongly associated with the risk of CAD.

KEYWORDS

smoking, CAD, coronary artery disease, genome-wide association study, mendelian randomization analysis

Introduction

Tobacco smoking is one of the most important public health problems world-wide, accounting for 9% of deaths (1). Several epidemiologic studies have proved that tobacco smoking is a major risk factor for many diseases such as lung cancer and cardiovascular and respiratory diseases (1). Lung cancer is the most prevalent smoking-associated cause of death, followed by ischemic heart disease and chronic airway obstruction. Tobacco smoking and its impact on the

respiratory system caused an estimated 8 million deaths per year, with more than 10% of these deaths related to second-hand smoke (2). In the United States, tobacco smoking is associated with 30% of all CAD-related deaths each year (3) and also doubles the risk of premature cardiovascular deaths (4).

Several epidemiologic studies have revealed that tobacco smoking increases the incidence of fatal CAD and associates with various cardiovascular diseases (5–8). Extensive clinical evidence has supported the idea that tobacco smoking causes multiple genetic and epigenetic abnormalities in the respiratory epithelium (9, 10). In a previous study, Sabater-Ileal and colleagues identified a genetic locus that influences both lung function and CAD (11), although the findings were not genome-wide in scale and were underpowered due to a small sample size.

Tobacco contains more than 4,000 chemicals (12), and the exact toxic components and the mechanisms involved in tobacco-related CAD and cardiovascular dysfunction are still unknown. Recently, a genetic predisposition to the development of atherogenesis in individuals exposed to cigarette smoke has been reported. The commonly documented examples are *CYP1A1* MSP polymorphism and certain endothelial NO synthase intron 4 polymorphisms. Both increase the susceptibility to cigarette smoke exposure-related atherosclerotic diseases including multi-vessel CAD and myocardial infarction (MI) (13, 14). Given that much of the available data were derived from observational studies, which are unable to account fully for confounding and reverse causation, the genetic correlation and causal relations between smoking and cardiovascular diseases remain to be determined.

The principle of Mendelian randomization (MR) relies on the basic laws of Mendelian genetics, segregation, and independent association. When these principles hold at a population level, the influence of confounding factors can be reduced because religion, growth, environment, and other confounding factors are considered to be random (15). Given that alleles are randomly allocated and become fixed at conception, MR studies are less susceptible to reverse causality than are observational studies.

In this study, we examined the pleiotropic effect of tobacco smoking and CAD using publicly available GWAS summary statistics. Then, we used bidirectional MR method to reveal the nature of the causal relations between CAD and tobacco smoking. Finally, we determined the biological processes or pathways involved in the comorbidity of these two diseases.

Materials and methods

GWAS summary data sets

The GWAS summary data for CAD and plasma lipid concentrations have been described in a previous report (16). Briefly, the summary statistics of a large GWAS meta-analysis comprising more than 120,000 CAD cases and 339,115 controls were obtained from CARDIoGRAMplusC4D Consortium website (http://www.cardiogramplusc4d.org/data-downloads/). A total of 9,149,595 variants were included either in the CARDIoGRAMplusC4D 1,000 Genomes-imputed GWAS or the MIGen/CARDIoGRAM Exome chip study. The smoking data were obtained from the large GSCAN summary statistics (17). The GSCAN investigated four

smoking-related phenotypes, including age at initiation of regular smoking (AgeSmk; n = 341,427), whether an individual had ever smoked regularly (SmkInit; n = 1,232,091), cigarettes smoked per day (CPD; n = 337,334), and smoking cessation (SmkCes; n = 547,219). The GWAS summary statistics for different smoking phenotypes can be found at https://conservancy.umn.edu/handle/11299/201564.

Further, we obtained published GWAS meta-analysis association data for lipid concentrations from the Center for Statistical Genetics, which was a joint analysis that examined 188,577 individuals whose genomic DNA samples were genotyped with two platforms from multiple studies (18). Complete GWAS summary statistics were downloaded from webpage http://csg.sph.umich.edu/willer/public/lipids2013/.

Estimation of genetic correlation by LD score regression (LDSC)

The genetic correlations (r_s) between CAD and smoking behaviors were estimated by LDSC (19). Pairwise LD r^2 among SNPs was conducted using pre-computed LD scores with the 1,000 Genomes Project reference panel of subjects of European ancestry. Quality control steps were adopted from LD scores default procedures, including imputation quality >0.9 and minor allele frequency > 0.1. Moreover, all SNPs retained for further analysis were merged with SNPs in the HapMap 3 reference panel. Correlation was considered significant at a corrected p value of <0.05 by Bonferroni correction.

Mendelian randomization (MR)

We extracted the effect estimates and standard errors (Ses) from relevant GWAS and employed TwoSampleMR (v. 0.4.22) R package to clarify the potential causal effect for both smoking and CAD (20). The following strategies were used to identify genetic instruments. First, we filtered GWAS summary datasets to require shared susceptible loci in both smoking and CAD. The variants showing genome-wide significance ($p < 5 \times 10^{-8}$) in GWAS for CAD were considered to be candidate variants, and then we checked the significance of these genetic loci separately in four other smoking behaviors: smoking initiation age, smoking cessation, CPD, and age of initiation in GSCAN studies. The common SNPs were harmonized using default parameters within the built-in "harmonize data" function and then trimmed by PLINK (v. 1.07) to obtain independent risk variants for each disease (21).

To start, the MR analysis was performed by generating instrumental variable estimates for each SNP. The averaged causal estimate of each SNP was calculated using the inverse-varianceweighted (IVW) method, i.e., specifically defined as the beta coefficient associated with SNP-CAD divided by the beta coefficient associated with SNP-smoking behaviors (22). In addition, we used a series of sensitivity analyses, which included weighted median and MR Egger, to evaluate the reliability of our results.

Gene and pathway analysis

The gene-based analysis that links SNPs to genes was conducted using MAGMA with default settings. To gain biological insights into shared genes, we used the WebGestalt tool (23) to assess enrichment of the identified shared gene set in the Gene Ontology (GO) biological processes with redundant GO terms been removed. Both analyses were based on shared genes that were identified from cross-trait metaanalysis. Pathways with a false discovery rate (FDR) < 0.05 were considered significant.

Results

Susceptible loci shared by CAD and smoking behaviors

To investigate the genetic overlap between CAD and smoking behaviors, we used GWAS summary data from large-scale genomewide studies (Supplementary Table S1). We detected a great number of significantly associated SNPs overlapped between CAD risk loci and at least one smoking phenotype (Supplementary Table S2). Of the CAD GWAS loci, 2091 SNPs (35.01%) showed nominal significance in CPD, 526 SNPs (25.16%) showed nominal significance for smoking initiation, 317 SNPs (15.16%) for smoking cession, and 85 SNPs (4.07%) for age at smoking initiation. Notably, 24 SNPs reached genome-wide significance (Figure 1).

Genetic correlations between CAD and smoking behaviors

We used the LDSC method to test for the genome-wide correlations between CAD and smoking behaviors. Significant genetic correlations were found between CAD and all the smoking-related traits with the smallest *p* values $<1 \times 10^{-18}$ (Figure 2). We observed significant positive genetic correlations between CAD and CPD, smoking initiation, and smoking cessation ($r_g > 0.2$), but a



FIGURE 1

Genetic susceptibility map for CAD and smoking behaviors. Outer ring defines location of 22 human autosomes. Scatter plots in second and fourth rings demonstrate analogy of Manhattan plot for association results from CAD and smoking behaviors, respectively. Altitude of each dot represents statistical significance as $-\log_{10}$ (P). SNPs that reached genome-wide significance are colored red for CAD and green for smoking behaviors. Yellow bars in third ring mark 24 CAD risk loci at least nominally associated with smoking behaviors, and tag SNPs in these loci are labeled.



FIGURE 2

Heatmap showing patterns of genetic correlations across CAD and smoking. Red color indicates a positive correlation and yellow color indicates a negative correlation. CAD, coronary artery disease; HDL, high-density lipoprotein; AgeofInitiation, age of Initiation; CigarettesPerDay, cigarettes per day; SmokingCessation, smoking cessation; SmokingInitiation, smoking Initiation; LDL, low-density lipoprotein; TG, triglycerides; TC, total cholesterol.

TABLE 1	Effect	of smoking	behaviors	on CAD	using	two-sample	MR analysis.
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Exposure	Outcome	No. SNPs	Beta	<i>p</i> -value	Method
Age of initiation	CAD	15	0.295	0.039	Inverse-variance weighted
		15	0.295	0.026	Weighted median
		15	-0.443	0.609	MR Egger
CPD	CAD	63	-0.039	0.406	Inverse-variance weighted
		63	-0.072	0.097	Weighted median
		63	-0.175	0.025	MR Egger
Smoking cessation	CAD	18	0.234	0.001	Inverse-variance weighted
		18	0.234	0.001	Weighted median
		18	0.091	0.559	MR Egger
Smoking initiation	CAD	152	0.191	2.59E-06	Inverse-variance weighted
		152	0.193	3.35E-06	Weighted median
		152	0.419	0.052	MR Egger

negative genetic correlation between CAD and age at initiation ($r_{\rm g}$ < -0.2).

Mendelian randomization analysis

Considering the presence of potential LD relations among those significant SNPs of interest, we performed p value-informed LD pruning with the goal of obtaining independent GWAS SNPs. This led to the identification of 15, 63, 18, and 152 independent SNPs for Age of Initiation, CPD, Smoking Cessation and Smoking Initiation,

respectively. Bi-directional MR analysis provided strong evidence that smoking initiation increased the risk of CAD (IVW: $\beta = 0.191$; $p = 2.59 \times 10^{-6}$) with a consistent direction of effect in all three MR methods (Table 1). There also was evidence for a consistent but weaker genetic liability for smoking cessation on CAD (IVW: $\beta = 0.234$; p = 0.001). The same findings were observed for age at initiation as the instrument on CAD (IVW: $\beta = 0.295$; p = 0.039).

On the other hand, we obtained negative correlations between CPD and CAD, although the statistic was less significant. Only the result of MR Egger was significant. When treating CAD as an instrument, strong evidence of decreased risk of smoking cessation by

Exposure	Outcome	No. SNPs	Beta	<i>p</i> -value	Method
	Age of initiation	68	0.012	0.013	Inverse-variance weighted
CAD		68	0.012	0.03	Weighted median
		68	-0.001	0.932	MR Egger
		68	-0.047	0.075	Inverse-variance weighted
CAD	CPD	68	-0.063	0.018	Weighted median
		68	-0.055	0.366	MR Egger
	Smoking cessation	75	-0.046	2.95E-08	Inverse-variance weighted
CAD		75	-0.046	4.29E-06	Weighted median
	_	75	-0.079	6.10E-04	MR Egger
	Smoking initiation	68	0.004	0.584	Inverse-variance weighted
CAD		68	0.003	0.637	Weighted median
		68	-0.019	0.299	MR Egger

TABLE 2 Effect of CAD on smoking behaviors using two-sample MR analysis.

CAD was observed (IVW: $\beta = -0.046$; $p = 2.95 \times 10^{-8}$; Table 2). The effects of CAD on other smoking behaviors became even weaker and inconsistent across the different methods.

Biological pathway and enrichment analysis

We performed pathway analyses to identify biological pathways enriched for shared genetic loci related to smoking and CAD based on significant cross-trait meta-analysis results. For a detailed list of the overlapped genes and SNPs between CAD and smoking-related phenotypes, please refer to Table 3. Pathway analysis showed that the SNP-related genes were significantly enriched in lipoprotein metabolic, chylomicron-mediated lipid transport, lipid digestion, mobilization, and transport (Table 4). The GO analysis suggested that shared genes in CAD and smoking behaviors were enriched in triglyceride-rich lipoprotein particle clearance, blood vessel development, and very-low-density lipoprotein particle clearance (Table 5).

Discussion

In this study, we revealed the genetic correlation and causal relations between smoking and CAD, providing a comprehensive evaluation of the shared genetic etiology of tobacco smoking and cardiovascular diseases. Our findings have highlighted the discovery that different smoking behaviors have strong associations with CAD, specifically, the correlation between smoking initiation, smoking cessation, and CAD.

The approach to MR is based on the assumptions that: (1) the genetic marker is associated with the exposure; (2) the genetic marker is independent of any confounding factors; and (3) there is no association between the genetic marker and outcome except through confounding factors. However, it should be acknowledged that these assumptions generally are not easy to evaluate. Results from the present MR study were based on data from the GWAS, which has corroborated the results obtained from conventional prospective

observational studies that confirmed that tobacco smoking is a risk factor for CAD (5-8).

To our knowledge, this study represents one of a few large-scale genome-wide analysis to investigate the genetic overlap between smoking and CAD (24-26). Similar to the findings from these reports, our analyses also revealed strong associations between smoking initiation, smoking cessation, and CAD. Further, we found a significant positive association between smoking initiation and CAD when smoking initiation was considered as exposure (inversevariance-weighted: $\beta = 0.191$; p = 2.59E-06; weighted median: $\beta = 0.193$; p = 3.35E-06), suggesting that smokers are more susceptible to CAD. In addition, we found a negative correlation between CAD and smoking cessation when CAD was considered as exposure (inverse-variance-weighted: $\beta = -0.046$; p = 2.95E-08; weighted median: $\beta = 0.193$; p=3.35E-06). This indicates that patients with CAD are less likely to quit smoking, possibly because of tobacco addiction. Together, these findings demonstrate the presence of shared genetic etiologies between tobacco smoking and CAD.

Moreover, we found strong evidence of a genetic correlation between CAD and serum high-density lipoprotein (HDL) and low-density lipoprotein (LDL) as well, which is consistent with the results reported previously (27–29). We observed a positive correlation between LDL and CAD when LDL was considered as the exposure (inverse–variance-weighted: $\beta = 0.387$; p = 4.96E-43; weighted median: $\beta = 0.406$; p = 7.41E-39), and a negative correlation between HDL and CAD when HDL was considered as the exposure (inverse–varianceweighted: $\beta = -0.245$; p = 1.19E-14; weighted median: $\beta = -0.188$; p = 2.66E-10). As is well documented in the literature, LDL is a strong risk factor for CAD (27–29) whereas HDL is an anti-atherosclerotic plasma lipoprotein and a protective factor against CAD (30–32).

Further, we found that smoking has significant associations with HDL and LDL. A positive correlation between CPD and LDL was observed when CPD was considered as exposure (inverse-variance-weighted: $\beta = 0.06$; p = 0.01; weighted median: $\beta = 0.06$; p = 0.02), and a negative correlation between CPD and HDL when CPD was considered as exposure (inverse-variance-weighted: $\beta = -0.06$; p = 0.005; weighted median: $\beta = -0.06$; p = 0.02). There also is a positive correlation between LDL and smoking cessation when LDL was considered as exposure (inverse-variance-weighted:

Gene	Gene function	SNPs				
ABI2	ABL Interactor 2	rs116426890				
ADAMTS7	ADAM metallopeptidase with Thrombospondin type 1 motif 7	rs7182642; rs4887096; rs55834964; rs7174367; rs12916326; rs12916648; rs1809419; rs1809409; rs11635870; rs11635931; rs1807006; rs12286; rs1807007; rs8043119; rs11633351; rs6495267; rs1809420; rs4886591; rs2904223; rs4887102; rs35934157				
ALDH2	Aldehyde dehydrogenase 2 family member	rs11513729				
APOB	Apolipoprotein B	rs503105				
APOC1	Apolipoprotein C1	rs12721051; rs56131196; rs4420638; rs12721046; rs111789331; rs66626994; rs10414043; rs7256200				
APOC1P1	Apolipoprotein C1 pseudogene 1	rs111789331; rs66626994				
APOE	Apolipoprotein E	rs429358; rs10414043; rs7256200; rs769449				
ATXN2	Ataxin 2	rs4766578;rs10774625;rs653178;rs597808;rs7137828;rs11065979; rs11065987				
BRAP	BRCA1 associated protein	rs11065979; rs11065987; rs11065991				
C19orf38	Chromosome 19 open reading frame 38	rs11881156; rs12973042; rs12979495; rs67987899; rs11881438; rs7260254; rs35443547				
CARF	Calcium responsive transcription factor	rs140244541; rs72928613; rs72936866; rs72932559; rs114702158; rs72928620; rs72932566; rs72928610; rs114123510; rs72928609; rs72932560; rs72926798; rs141002954; rs72932554;rs72932561; rs9749722; rs74421437; rs72936872; rs72936873				
CARM1	Coactivator associated arginine methyltransferase 1	rs12971616; rs4804547; rs17616105; rs12976693; rs35991287; rs74179956; rs12977506; rs73009538; rs4804142; rs11880628; rs35734575; rs11879571; rs36045835; rs7256879; rs34857893; rs2053064; rs1971038; rs8105092; rs2053065; rs1541595; rs1971039; rs2304089				
CDKN2B-AS1	CDKN2B antisense RNA 1	rs568447; rs545226; rs1333045; rs10217586; rs10738604; rs10811641; rs3217992; rs3218020; rs2891168; rs4977574; rs10757274; rs504318; rs10757264; rs7028268; rs567453; rs496892; rs490005; rs10757272; rs10738606; rs10120688; rs10738607; rs10811645; rs10811643; rs10738610; rs7035484; rs7049105; rs1591136; rs10811647; rs3218012; rs10115049; rs10811656; rs10757266; rs10757279; rs10811650				
CELSR2	Cadherin EGF LAG seven-pass G-type receptor 2	rs11102967; rs611917				
CHRNB4	Cholinergic receptor nicotinic beta 4 subunit	rs56354501; rs8031513; rs56195905; rs1809412; rs1814880; rs11072799; rs1809414; rs11856441; rs12901228; rs2219939; rs1825086; rs1383635; rs1383634; rs58717592; rs59671175; rs1809415				
CNNM2	Cyclin and CBS domain divalent metal cation transport mediator 2	rs12414777				
COL4A1	Collagen type IV alpha 1 chain	rs12866570				
COL4A2	Collagen type IV alpha 2 chain	rs55940034; rs9515201				
CYP20A1	Cytochrome P450 family 20 subfamily A member 1	rs72938351; rs72938315; rs115600411; rs114863726; rs114407963; rs116443099				
DMRTA1	DMRT like family A1	rs10738610; rs10811656; rs10757279; rs10757277; rs1333046; rs1333049; rs10757278; rs1333048; rs4977575; rs7857118; rs1333047				
DNM2	Dynamin 2	rs117159625; rs11880613; rs117786851; rs8111962; rs1109375; rs2287029				
EDNRA	Endothelin receptor type A	rs7668383				
FAM117B	Family with sequence similarity 117 member B	rs146289328				
FES	FES proto-oncogene, tyrosine kinase	rs2521501				
GGCX	Gamma-glutamyl carboxylase	rs7568458; rs35215812; rs10198569; rs10187424; rs12473819; rs6547624; rs6757263				
HECTD4	HECT domain E3 ubiquitin protein ligase 4	rs11066188; rs7953257				
ICA1L IL6R	Islet cell autoantigen 1 like Interleukin 6 receptor	rs11675462; rs4675290; rs2036927; rs12693975; rs10207567; rs1971819; rs934287; rs72932720; rs72932716; rs72932711; rs72932709; rs72932707; rs76122535; rs72932722; rs72932723; rs72932731; rs72932737; rs72934734; rs72932725; rs143911965; rs6435168; rs6732078; rs72932746; rs80087860; rs6705330; rs72932752; rs78128841 rs6694258; rs6689393; rs4845625; rs12129500; rs6686750; rs4845619; rs7549338; rs6667434; rs7553796; rs4553185; rs59632925; rs7549250; rs4845371; rs11265612; rs6687726; rs4845618; rs6694817; rs11265611; rs6427558; rs12118721; rs10008239; rs13118770; rs13117822;				
		rs6689306				

TABLE 3 The overlapped genes and SNPs between CAD and smoking-related phenotypes.

TABLE 3 (Continued)

Gene	Gene function	SNPs				
KCNE2	Potassium voltage-gated channel subfamily E regulatory	rs1018757; rs2211693; rs2211695; rs2211694; rs8128536; rs7279974; rs8129119; rs4817636;				
	subunit 2	rs2096469; rs4817637; rs2096467; rs2834439; rs7280276; rs9982672; rs1023354; rs28451064;				
		rs762158; rs9980618; rs743339; rs7277800; rs9982601; rs4817639; rs9974878; rs7281592;				
WO WE		rs6068/229; rs2834431; rs28591415; rs16991453; rs99/8142; rs/283231; rs8132042; rs8131284				
KCNK5	Potassium two pore domain channel subtamily K member 5	rs1155347; rs10456468; rs56015508; rs55902013; rs9394577; rs12211281; rs55856036; rs1544935				
LDLR	Low density lipoprotein receptor	rs8103309; rs11666925; rs876794; rs10409001; rs10415811; rs11085757; rs8102273; rs11879026; rs73013202; rs73013198				
LINC00310	Long intergenic non-protein coding RNA 310	rs1018757; rs2211693; rs2211695; rs2211694; rs8128536; rs7279974; rs8129119; rs4817636; rs2096469; rs4817637; rs2096467; rs2834439; rs7280276; rs9982672; rs1023354; rs28451064				
LINC00841	Long intergenic non-protein coding RNA 841	rs2051120; rs6593388; rs2624688; rs6593393; rs4948800; rs2085797; rs7478408; rs898549; rs10899956; rs12359058; rs10899955; rs7924201; rs7088951; rs10899963; rs10899954; rs898551; rs10793515; rs66887775				
LOC646938	TBC1 domain family member 2B pseudogene	rs56354501; rs8031513; rs55834964; rs56195905; rs1809412; rs1814880; rs11072799; rs1809414; rs11856441; rs12901228; rs2219939; rs1825086				
MAP3K4	Mitogen-activated protein kinase kinase kinase 4	rs117340856				
MAPKAPK5- AS1	MAPKAPK5 antisense RNA 1	rs11513729				
MC4R	Melanocortin 4 receptor	rs523288; rs538656; rs571312; rs663129; rs6567160; rs34633411; rs11152213; rs66723169; rs1942872; rs12958167; rs12955983; rs12954782; rs11663816				
MIA3	MIA SH3 domain ER export factor 3	rs1909196; rs34679168; rs34767248; rs2378584; rs17163301; rs4618978; rs4846769; rs4846384;				
	<u>^</u>	rs4846770; rs17163345; rs17163313				
MIR4422	Microrna 4,422	rs55694910; rs72664304; rs72664303				
MORF4L1	Mortality factor 4 like 1	rs11634042; rs8034274; rs7168915; rs11852830; rs11857877; rs4439728; rs12232282;				
		rs4420501; rs4438276; rs8037171; rs11637783				
MRAS	Muscle RAS oncogene homolog	rs13324341; rs1199337; rs1199338; rs185244				
MTAP	Methylthioadenosine phosphorylase	rs7041637				
NAA25	N-alpha-acetyltransferase 25, NatB auxiliary subunit	rs17696736				
NBEAL1	Neurobeachin like 1	rs72932566; rs76461893; rs145299755; rs72932573; rs148707292; rs72932572; rs72932574;				
		rs72932583; rs6728861; rs72932575; rs4675310; rs148812085; rs2351524; rs140750546;				
NOS3	Nitric ovide synthese 3	re3018226				
DHACTD1	Phosphatase and actin regulator 1	re10207333; re13200003; re0321401; re12202801; re0305224; re0472700; re1571007; re0321463;				
THACTRI	Thosphatase and actin regulator 1	rs6916397; rs2026458; rs6916421; rs7776079				
PLG	Plasminogen	rs117340856				
PMAIP1	Phorbol-12-myristate-13-acetate-induced protein 1	rs523288; rs538656; rs571312; rs663129; rs6567160; rs34633411; rs11152213; rs66723169;				
		rs1942872; rs12958167; rs12955983; rs12954782; rs11663816; rs11664883; rs2045438; rs35476226; rs17175602; rs2045439; rs11660069; rs12957325; rs12970134				
RAPH1	RAS association (RalGDS/AF-6) and pleckstrin homology domains 1	rs139644567; rs140274075				
SAYSD1	SAYSVFN motif domain containing 1	rs1155347; rs10456468; rs56015508; rs55902013; rs9394577; rs12211281; rs55856036; rs1544935				
SH2B3	SH2B adaptor protein 3	rs7310615; rs3184504; rs10774624				
SH2D6	SH2 domain containing 6	rs6750847; rs7591175; rs6722691; rs11895399; rs11895401; rs1446669; rs2166529; rs6547620; rs6721924; rs6739015; rs2044474; rs13394343; rs6719046; rs6733913; rs17026396				
SMARCA4	SWI/SNF related, matrix associated, actin dependent regulator of chromatin, subfamily a, member 4	rs8103309; rs60314748; rs60448955; rs73013159; rs7275; rs55677033; rs3786727; rs11666925; rs11670205; rs3786725; rs10417578; rs68010235; rs6511718; rs876794				
SPECC1L-	SPECC1L-ADORA2A readthrough (NMD candidate)	rs5760347; rs62233136; rs5760359; rs5751841; rs5760368; rs62233133; rs5760350; rs2298379				
ADORA2A						

(Continued)

TABLE 3 (Continued)

Gene	Gene function	SNPs		
SVIL	Supervillin	rs1774241; rs1774240; rs1832864; rs12259037; rs12779954; rs1418276; rs4749520; rs11007851;		
		rs10826749		
TAF1A-AS1	TAF1A antisense RNA 1	rs1909196; rs34679168; rs4618978		
TMED1	Transmembrane p24 trafficking protein 1	rs73007593; rs11881156		
TRAFD1	TRAF-type zinc finger domain containing 1	rs17630235		
TTC29	Tetratricopeptide repeat domain 29	rs7668383		
VAMP8	Vesicle associated membrane protein 8	rs35215812; rs10198569; rs10187424; rs6547624; rs6757263		
WDR12	WD repeat domain 12	rs72934734; rs35212307; rs114395475; rs7582720; rs72936856; rs72934751; rs72934763;		
		rs72934749; rs6738618; rs150788469; rs6725887; rs6435169; rs7560547; rs77931721;		
		rs142603618; rs72936846; rs72936852; rs72936830; rs143035655; rs77268589		
YIPF2	Yip1 domain family member 2	rs17850995		
ZC3HC1	Zinc finger C3HC-type Containing 1	rs11556924; rs56179563		

TABLE 4 Detected shared pathway between smoking and CAD based on pathway analysis.

Pathway ID	Pathway name	<i>p</i> -value	q-value (FDR B&H)	Genes included
PW:0000482	Lipoprotein metabolic	5.15E-06	9.66E-04	APOB, APOC1, APOE
1,270,005	Lipoprotein metabolism	4.57E-05	3.86E-03	APOB, APOC1, APOE, LDLR
1,270,006	Chylomicron-mediated lipid transport	5.74E-05	4.22E-03	APOB, APOE, LDLR
1,270,002	Lipid digestion, mobilization, and transport	2.54E-04	1.18E-02	APOB, APOC1, APOE, LDLR
M12950	Angiotensin-converting enzyme 2 regulates heart function	5.49E-04	1.62E-02	COL4A1, COL4A2
1,270,115	Metabolism of nitric oxide	1.60E-03	3.41E-02	NOS3, DNM2
1,427,851	VLDL interactions	1.75E-03	3.41E-02	APOB, APOC1
172,847	Protein digestion and absorption	1.83E-03	3.41E-02	KCNK5, COL4A1, COL4A2
1,270,008	LDL-mediated lipid transport	1.91E-03	3.41E-02	APOB, LDLR

TABLE 5 Detected shared pathways between CAD and smoking based on GO analysis.

Gene Ontology (GO) ID	Gene Ontology (GO) term	p-value	q-value (FDR B&H)	Genes included
GO:0071830	Triglyceride-rich lipoprotein particle clearance	5.75E-09	7.17E-06	APOB, APOC1, APOE, LDLR
GO:0001568	Blood vessel development	1.94E-06	1.34E-03	APOB, APOE, NOS3, PLG, COL4A1, COL4A2, DNM2, EDNRA, MIA3, SMARCA4, IL6R, LDLR
GO:0034447	Very-low-density lipoprotein particle clearance	2.16E-06	1.34E-03	APOB, APOC1, APOE
GO:0001944	Vasculature development	2.95E-06	1.43E-03	APOB, APOE, NOS3, PLG, COL4A1, COL4A2, DNM2, EDNRA, MIA3, SMARCA4, IL6R, LDLR
GO:0072358	Cardiovascular system development	3.43E-06	1.43E-03	APOB, APOE, NOS3, PLG, COL4A1, COL4A2, DNM2, EDNRA, MIA3, SMARCA4, IL6R, LDLR
GO:0042159	Lipoprotein catabolic process	8.10E-06	2.53E-03	APOB, APOE, LDLR
GO:0030195	Negative regulation of blood coagulation	1.53E-05	3.72E-03	APOE, NOS3, PLG, SH2B3
GO:0048514	Blood vessel morphogenesis	2.76E-05	4.31E-03	APOB, APOE, NOS3, PLG, COL4A1, COL4A2, EDNRA, MIA3, SMARCA4, LDLR
GO:1990777	Lipoprotein particle	2.51E-06	2.30E-04	APOB, APOC1, APOE, LDLR
GO:0034358	Plasma lipoprotein particle	2.51E-06	2.30E-04	APOB, APOC1, APOE, LDLR
GO:0032994	Protein-lipid complex	3.09E-06	2.30E-04	APOB, APOC1, APOE, LDLR
GO:0034362	Low-density lipoprotein particle	6.77E-06	3.00E-04	APOB, APOE, LDLR
GO:0034364	High-density lipoprotein particle	4.26E-05	1.13E-03	APOB, APOC1, APOE

 $\beta = -0.03$; p = 0.002; weighted median: $\beta = -0.04$; p = 2.43E-5). Taken together, these findings provided a clear indication that smoking increases the risk of CAD by affecting the regulation of LDL and HDL, which needs to be further investigated.

We also performed GO enrichment and KEGG pathway analyses based on the genes where the SNPs overlapped between CAD and smoking-related phenotypes are located. We found several functions and pathways to be related to the lipoprotein metabolic and blood vessel development, which are all closely associated with CAD. It has been reported that the APOE–APOC1–APOC2–APOC4 cluster was significantly related to lipoprotein-associated phospholipase A2 mass and activity and CAD (33). Interestingly, the SNP rs4420638 located downstream of the APOC1 gene was found to be significantly related to smoking cessation (p=7.4E-6) (34). Moreover, as a brain eQTL based on the information from BRAINEAC, rs4420638, this SNP has been linked to Alzheimer's disease (35–40) and cognitive decline (41).

We used non-overlapping data sources in the context of summary-level MR analysis of exposure and outcome, which greatly improved the confidence in the causal effect estimates. In addition, through a range of sensitivity analysis methods, similar causal estimates and consistent causal inferences could be drawn. However, this study had limitations as well. First, there was a stark difference in sample sizes among different phenotypes, which might contribute to discrepancies in statistical power. Second, the information available on the summary-level GWAS data had limited us to divide samples into subgroups, which prevented us from studying the age-related heterogeneity.

In conclusion, this was a systematic analysis of the shared etiology and possible causal relations of smoking and CAD by employing large-scale GWASs. Genetic methods represent another option for assessing causality when there are too many confounding factors in randomized controlled traits, and our findings strongly support the hypothesis that smoking behavior is causally related to CAD risk. We found significant genetic overlap and correlations between CAD and smoking at the SNP level. Taken together, the data from this study enhance the understanding of the genetic etiology of the relations between CAD and smoking and might help to dissect smoking behaviors and develop preventive strategies to reduce the burden of cardiovascular disease in public health.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary material, further inquiries can be directed to the corresponding authors.

Author contributions

ZZ: Writing – original draft, Writing – review & editing, Data curation, Formal analysis. QL: Data curation, Formal analysis, Writing – original draft, Methodology. MeL: Data curation, Formal analysis, Methodology, Writing – review & editing. YY: Data curation, Formal analysis, Methodology, Software, Writing – original draft. FQ: Data curation, Formal analysis, Software, Writing – original draft. YX: Data curation, Formal analysis, Methodology, Software, Writing – original draft. SL: Data curation, Formal analysis, Methodology, Writing – review & editing. ZY: Data curation, Project administration, Writing – review & editing. YG: Project administration, Writing – review & editing, Investigation, Resources, Supervision. MiL: Funding acquisition, Investigation, Resources, Writing – review & editing, Conceptualization, Supervision, Writing – original draft. JY: Funding acquisition, Investigation, Project administration, Resources, Writing – review & editing.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fpsyt.2023.1279962/ full#supplementary-material

References

1. CDC. Current cigarette smoking among adults-United States. MMWR Morb Mortal Wkly Rep. (2012) 61:889-94.

2. World Health Organization. *WHO report on the global tobacco epidemic, 2011: warning about the dangers of tobacco*. Geneva, Switzerland: World Health Organization (2011).

3. National Center for Chronic Disease Prevention and Health Promotion (US) Office on Smoking and Health In: *The health consequences of Smoking-50 years of progress: a report of the surgeon general.* Atlanta, GA: Centers for Disease Control and Prevention (US) (2014).

4. Mons U, Muezzinler A, Gellert C, Schottker B, Abnet CC, Bobak M, et al. Impact of smoking and smoking cessation on cardiovascular events and mortality among older adults: meta-analysis of individual participant data from prospective cohort studies of the CHANCES consortium. *BMJ*. (2015) 350:h1551. doi: 10.1136/bmj.h1551

5. Doll R, Hill AB. Lung cancer and other causes of death in relation to smoking; a second report on the mortality of British doctors. *Br Med J*. (1956) 2:1071–81. doi: 10.1136/bmj.2.5001.1071

6. Jonas MA, Oates JA, Ockene JK, Hennekens CH. Statement on smoking and cardiovascular disease for health care professionals. *Am Heart Assoc Circul.* (1992) 86:1664–9. doi: 10.1161/01.cir.86.5.1664

7. Price JF, Mowbray PI, Lee AJ, Rumley A, Lowe GD, Fowkes FG. Relationship between smoking and cardiovascular risk factors in the development of peripheral arterial disease and coronary artery disease: Edinburgh artery study. *Eur Heart J.* (1999) 20:344–53. doi: 10.1053/euhj.1998.1194

8. Willett WC, Green A, Stampfer MJ, Speizer FE, Colditz GA, Rosner B, et al. Relative and absolute excess risks of coronary heart disease among women who smoke cigarettes. *N Engl J Med.* (1987) 317:1303–9. doi: 10.1056/NEJM198711193172102

9. Franklin WA, Gazdar AF, Haney J, Wistuba II, La Rosa FG, Kennedy T, et al. Widely dispersed p53 mutation in respiratory epithelium. A novel mechanism for field carcinogenesis. J Clin Invest. (1997) 100:2133–7. doi: 10.1172/JCI119748

10. Wistuba II, Lam S., Behrens C., Virmani, A.K., Fong, K.M., LeRiche J., et al. (1997). Molecular damage in the bronchial epithelium of current and former smokers. *J Natl Cancer Inst* 89, 1366–1373. doi: 10.1093/jnci/89.18.1366

11. Sabater-Lleal M, Malarstig A, Folkersen L, Soler Artigas M, Baldassarre D, Kavousi M, et al. Common genetic determinants of lung function, subclinical atherosclerosis and risk of coronary artery disease. *PLoS One*. (2014) 9:e104082. doi: 10.1371/journal.pone.0104082

12. Burns DM. Cigarettes and cigarette smoking. *Clin Chest Med.* (1991) 12:631–42. doi: 10.1016/S0272-5231(21)00813-3

13. Wang XL, Greco M, Sim AS, Duarte N, Wang J, Wilcken DE. Effect of CYP1A1 MspI polymorphism on cigarette smoking related coronary artery disease and diabetes. *Atherosclerosis.* (2002) 162:391–7. doi: 10.1016/s0021-9150(01)00723-7

14. Wang XL, Sim AS, Wang MX, Murrell GA, Trudinger B, Wang J. Genotype dependent and cigarette specific effects on endothelial nitric oxide synthase gene expression and enzyme activity. *FEBS Lett.* (2000) 471:45–50. doi: 10.1016/s0014-5793(00)01356-9

15. Smith GD, Ebrahim S. 'Mendelian randomization': can genetic epidemiology contribute to understanding environmental determinants of disease? *Int J Epidemiol.* (2003) 32:1–22. doi: 10.1093/ije/dyg070

16. Chen L, Yao Y, Jin C, Wu S, Liu Q, Li J, et al. Integrative genomic analysis identified common regulatory networks underlying the correlation between coronary artery disease and plasma lipid levels. *BMC Cardiovasc Disord*. (2019) 19:310. doi: 10.1186/ s12872-019-01271-9

17. Liu M, Jiang Y, Wedow R, Li Y, Brazel DM, Chen F, et al. Association studies of up to 1.2 million individuals yield new insights into the genetic etiology of tobacco and alcohol use. *Nat Genet.* (2019) 51:237–44. doi: 10.1038/s41588-018-0307-5

18. Willer CJ, Schmidt EM, Sengupta S, Peloso GM, Gustafsson S, Kanoni S, et al. Discovery and refinement of loci associated with lipid levels. *Nat Genet.* (2013) 45:1274–83. doi: 10.1038/ng.2797

19. Bulik-Sullivan B, Finucane HK, Anttila V, Gusev A, Day FR, Loh PR, et al. An atlas of genetic correlations across human diseases and traits. *Nat Genet.* (2015) 47:1236–41. doi: 10.1038/ng.3406

20. Hemani G, Zheng J, Elsworth B, Wade KH, Haberland V, Baird D, et al. The MRbase platform supports systematic causal inference across the human phenome. *elife*. (2018) 7. doi: 10.7554/eLife.34408 21. Purcell S, Neale B, Todd-Brown K, Thomas L, Ferreira MA, Bender D, et al. PLINK: a tool set for whole-genome association and population-based linkage analyses. *Am J Hum Genet.* (2007) 81:559–75. doi: 10.1086/519795

22. Burgess S, Butterworth A, Thompson SG. Mendelian randomization analysis with multiple genetic variants using summarized data. *Genet Epidemiol.* (2013) 37:658–65. doi: 10.1002/gepi.21758

23. Zhang B, Kirov S, Snoddy J. WebGestalt: an integrated system for exploring gene sets in various biological contexts. *Nucleic Acids Res.* (2005) 33:W741–8. doi: 10.1093/nar/gki475

24. Chen S, Yang F, Xu T, Wang Y, Zhang K, Fu G, et al. Smoking and coronary artery disease risk in patients with diabetes: a mendelian randomization study. *Front Immunol.* (2023) 14:891947. doi: 10.3389/fimmu.2023.891947

25. Jia Y, Wang R, Guo D, Sun L, Shi M, Zhang K, et al. Contribution of metabolic risk factors and lifestyle behaviors to cardiovascular disease: a mendelian randomization study. *Nutr Metab Cardiovasc Dis.* (2022) 32:1972–81. doi: 10.1016/j.numecd.2022.04.019

26. Levin MG, Klarin D, Assimes TL, Freiberg MS, Ingelsson E, Lynch J, et al. Genetics of smoking and risk of atherosclerotic cardiovascular diseases: a mendelian randomization study. *JAMA Netw Open.* (2021) 4:e2034461. doi: 10.1001/jamanetworkopen.2020.34461

27. Moon JY, Kwon HM, Kwon SW, Yoon SJ, Kim JS, Lee SJ, et al. Lipoprotein(a) and LDL particle size are related to the severity of coronary artery disease. *Cardiology*. (2007) 108:282–9. doi: 10.1159/000099097

28. van den Berg VJ, Vroegindewey MM, Kardys I, Boersma E, Haskard D, Hartley A, et al. Anti-oxidized LDL antibodies and coronary artery disease: a systematic review. *Antioxidants.* (2019) 8. doi: 10.3390/antiox8100484

29. Zhao X, Zhang HW, Xu RX, Guo YL, Zhu CG, Wu NQ, et al. Oxidized-LDL is a useful marker for predicting the very early coronary artery disease and cardiovascular outcomes. *Per Med.* (2018) 15:521–9. doi: 10.2217/pme-2018-0046

30. Heinecke JW. The HDL proteome: a marker--and perhaps mediator--of coronary artery disease. J Lipid Res. (2009) 50:S167–71. doi: 10.1194/jlr.R800097-JLR200

31. Natarajan P, Collier TS, Jin Z, Lyass A, Li Y, Ibrahim NE, et al. Association of an HDL Apolipoproteomic score with coronary atherosclerosis and cardiovascular death. *J Am Coll Cardiol*. (2019) 73:2135–45. doi: 10.1016/j.jacc.2019.01.073

32. Weissglas-Volkov D, Pajukanta P. Genetic causes of high and low serum HDLcholesterol. *J Lipid Res.* (2010) 51:2032–57. doi: 10.1194/jlr.R004739

33. Grallert H, Dupuis J, Bis JC, Dehghan A, Barbalic M, Baumert J, et al. Eight genetic loci associated with variation in lipoprotein-associated phospholipase A2 mass and activity and coronary heart disease: meta-analysis of genome-wide association studies from five community-based studies. *Eur Heart J.* (2012) 33:238–51. doi: 10.1093/eurheartj/ehr372

34. Horwitz T, Lam K, Chen Y, Xia Y, Liu C. A decade in psychiatric GWAS research. *Mol Psychiatry*. (2019) 24:378–89. doi: 10.1038/s41380-018-0055-z

35. Coon KD, Dunckley TL, Stephan DA. A generic research paradigm for identification and validation of early molecular diagnostics and new therapeutics in common disorders. *Mol Diagn Ther*. (2007) 11:1–14. doi: 10.1007/BF03256218

36. Herold C, Hooli BV, Mullin K, Liu T, Roehr JT, Mattheisen M, et al. Family-based association analyses of imputed genotypes reveal genome-wide significant association of Alzheimer's disease with OSBPL6, PTPRG, and PDCL3. *Mol Psychiatry*. (2016) 21:1608–12. doi: 10.1038/mp.2015.218

37. Kamboh MI, Barmada MM, Demirci FY, Minster RL, Carrasquillo MM, Pankratz VS, et al. Genome-wide association analysis of age-at-onset in Alzheimer's disease. *Mol Psychiatry*. (2012) 17:1340–6. doi: 10.1038/mp.2011.135

38. Kamboh MI, Demirci FY, Wang X, Minster RL, Carrasquillo MM, Pankratz VS, et al. Genome-wide association study of Alzheimer's disease. *Transl Psychiatry*. (2012) 2:e117. doi: 10.1038/tp.2012.45

39. Li H, Wetten S, Li L, St Jean PL, Upmanyu R, Surh L, et al. Candidate singlenucleotide polymorphisms from a genomewide association study of Alzheimer disease. *Arch Neurol.* (2008) 65:45–53. doi: 10.1001/archneurol.2007.3

40. Webster JA, Myers AJ, Pearson JV, Craig DW, Hu-Lince D, Coon KD, et al. Sorl1 as an Alzheimer's disease predisposition gene? *Neurodegener Dis.* (2008) 5:60–4. doi: 10.1159/000110789

41. De Jager PL, Shulman JM, Chibnik LB, Keenan BT, Raj T, Wilson RS, et al. A genomewide scan for common variants affecting the rate of age-related cognitive decline. *Neurobiol Aging*. (2012) 33:1017 e1011–5. doi: 10.1016/j.neurobiolaging.2011.09.033