



Air pollutants, genetic susceptibility, and the risk of age-related macular degeneration: a large prospective cohort study

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Abstract

Age-related macular degeneration (AMD) is a leading cause of irreversible vision loss worldwide. However, evidence regarding the relationship between air pollution and AMD is limited, and the modifying effect of genetic susceptibility on this association remains unknown. A total of 445,237 participants without AMD at baseline were included from the UK Biobank. The concentrations of nitrogen dioxide (NO₂), nitrogen oxides (NO_x), particulate matter (PM_{2.5}, PM₁₀, PM_{2.5–10}) were collected by using land-use regression models. Air pollution score (APS) was constructed through summing each pollutant weighted by the regression coefficients with AMD from single-pollutant model. Cox proportional hazard models were used to evaluate hazard ratios (HRs) and 95% confidence intervals (95% CIs) of associations between air pollutants and polygenic risk score (PRS) with incident AMD. During a median follow-up of 13.83 years, we observed 9,635 incident AMD events. The HR (95%CI) of incident AMD for each standard deviation increase in NO₂, NO_x, PM_{2.5}, PM₁₀, and APS were 1.04(1.02, 1.06), 1.03(1.01, 1.05), 1.04(1.02, 1.07), 1.02(1.00, 1.04), and 1.04(1.02, 1.06), respectively. Significant additive interaction effects of NO₂, NO_x, PM_{2.5–10}, APS and PRS with incident risk of AMD were observed, with the relative excess risk due to the interaction (RERI), attributable proportion (AP), and their 95% CIs of 0.10(0.01, 0.18) and 0.05(0.01, 0.11) for NO₂, 0.11(0.01, 0.19) and 0.05(0.02, 0.10) for NO_x, 0.15(0.06, 0.23) and 0.08(0.03, 0.13) for PM_{2.5–10}, and 0.12(0.03, 0.20) and 0.06(0.01, 0.11) for APS, respectively. Compared with participants exposed to low level of above air pollutants and low PRS, those exposed to high air pollution and high PRS had almost double incident risk of AMD [HR(95%CI) ranged from 1.83(1.68, 1.99) to 2.03(1.86, 2.21)]. Long-term exposure to air pollutants NO₂, NO_x, PM_{2.5}, and PM₁₀ showed positive associations with increased risk of AMD, which could be further enhanced by genetic susceptibility.

Keywords Air pollution · Genetic susceptibility · Age-related macular degeneration · Interaction · Cohort study

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Abbreviations

NO ₂	Nitrogen dioxide
NO _x	Nitrogen oxides
PM _{2.5}	Fine particulate matter with diameter < 2.5 µm
PM ₁₀	Particulate matter with diameter < 10 µm
PM _{2.5–10}	Particulate matter with aerodynamic diameter between 2.5 and 10 µm

Introduction

Age-related macular degeneration (AMD) is a complex, multifactorial disease characterized by progressive degeneration of the macula, the central region of the retina, and

represents as a leading cause of permanent vision loss worldwide [1, 2]. The pooled global prevalence of AMD is estimated up 8.7% [3] and the number of affected individuals is expected to increase to 288 million by 2040 [4]. It is of great concern that AMD will pose an increasing threat to global public health due to the population aging. Therefore, identifying the potential risk factors of AMD is crucial to reduce the burden of this disease.

Ambient air pollution has emerged as a major environmental threat to the global disease burden [5]. The human ocular system may be susceptible to environmental damage due to its direct exposure to various external pollutants [6]. Although several epidemiological studies have explored the association between air pollution exposure and AMD, the findings in this field were limited and conflicting [7–12]. For instance, a national cross-sectional study showed a positive correlation between particulate matter (PM) with diameters $\leq 2.5 \mu\text{m}$ ($\text{PM}_{2.5}$) exposure and AMD prevalence across 10 provinces in rural China [12]. Another longitudinal cohort study conducted in Taiwan also showed that $\text{PM}_{2.5}$ is a risk factor for the development of AMD [10]. In contrast, a large cross-sectional study involving 115,954 European participants did not observe significant associations of exposure to nitrogen dioxide (NO_2), nitrogen oxides (NO_x), $\text{PM} \leq 10 \mu\text{m}$ (PM_{10}) and PM with diameters of $2.5\text{--}10 \mu\text{m}$ ($\text{PM}_{2.5\text{--}10}$) with AMD [11]. Actually, in the real world, people are usually exposure to various air pollutants simultaneously [13, 14], but most previous studies have focused on individual pollutant. Hence, more prospective studies are required to elucidate the long-term effects of various air pollutants on AMD incidence.

Growing evidence has revealed that genetic factors also play a key role in the etiology of AMD [2, 15]. Comprehensive genome-wide association studies (GWAS) have identified certain genetic variants associated with AMD risk [16]. Although individual variants account for only a relatively modest proportion on the heritability, using polygenic risk score (PRS) has proven to be an effective method for quantifying the cumulative effect of multiple risk-associated variants [17]. Moreover, given that the effects of environment factors may be modified by genetic predispositions [18], exploring gene-environment interactions may provide novel insights into the precise prediction and targeted interventions for AMD [19]. However, whether and how the genetic susceptibility modifies the associations between ambient air pollutions and AMD remains largely unknown.

Taking advantage of the national information from the UK Biobank, the present study aimed to investigate the effects of individual and combined air pollutants, including NO_2 , NO_x , $\text{PM}_{2.5}$, PM_{10} and $\text{PM}_{2.5\text{--}10}$, on the incident risk of AMD. We further assessed the potential gene-air pollution

interactions and explored their joint effects of air pollution and genetic susceptibility on AMD incidence.

Methods

Study population

The UK Biobank is a national prospective study of over half a million participants aged 40–69 years recruited from 22 assessment centers across the UK between 2006 and 2010. It was designed to support in-depth investigations of the genetic and non-genetic determinants of disease in middle and older age. The UK Biobank seeks to advance understanding and promote innovative research by integrating detailed exposure assessments with long-term follow-up and broad phenotypic characterisation, and by providing open access to its data and samples. Extensive information was collected at baseline, including questionnaires, biological samples and clinical measurements [20]. All participants provided informed written consent, and the study protocol was approved by the North West Multi-Centre Research Ethics Committee.

In the present study, we first excluded participants with pre-existing AMD at baseline ($n=689$). Subsequently, we excluded individuals with missing information on air pollution ($n=41,248$), and genetic data ($n=14,977$). Ultimately, a total of 445,237 participants were included for the final analysis (Figure S1).

Air pollution assessment

The annual average concentrations of ambient air pollutants, including NO_2 , NO_x , $\text{PM}_{2.5}$, PM_{10} and $\text{PM}_{2.5\text{--}10}$ were estimated using Land Use Regression (LUR) models developed by the European Study of Cohorts for Air Pollution Effects (ESCAPE) project (<https://biobank.ndph.ox.ac.uk/ukb/refer.cgi?id=2010>). The LUR models were based on monitoring data collected by the ESCAPE project between 26 January 2010 and 18 January 2011, with air pollution estimates representative of the year 2010. Generally, LUR is a widely adopted method for capturing spatial variation in ambient pollution, incorporating geospatial predictor variables such as traffic intensity, topography, and land use derived from the Geographic Information System (GIS). Leave-one-out cross-validation showed good model performance for air pollutants [21, 22]. Then, air pollution concentrations were assigned to each UK Biobank participant according to their baseline residential address, geocoded to $100 \text{ m} \times 100 \text{ m}$ grid cells.

To capture the combined effects of various air pollutants, we constructed an air pollution score (APS) through

summing up concentrations of five ambient air pollutants (NO_2 , NO_x , $\text{PM}_{2.5}$, PM_{10} and $\text{PM}_{2.5-10}$), weighted by their respective risk estimates (β coefficients) for AMD obtained from the fully-adjusted model (Table S1). This method of calculating APS has been well accepted in previous UK Biobank studies to reflect the combined effects of multiple air pollutants on health outcomes [23]. The equation was: $\text{APS} = [\beta_{(\text{NO}_2)} \times \text{NO}_2 + \beta_{(\text{NO}_x)} \times \text{NO}_x + \beta_{(\text{PM}_{2.5})} \times \text{PM}_{2.5} + \beta_{(\text{PM}_{10})} \times \text{PM}_{10} + \beta_{(\text{PM}_{2.5-10})} \times \text{PM}_{2.5-10}] \times (5 / \sum \beta)$. The APS ranges from 43.76 to 154.45, with higher score indicating greater exposure to air pollution. In sensitivity analysis, we also calculated a weighted APS excluding $\text{PM}_{2.5-10}$, based on the remaining four air pollutants.

Definition of the genetic risk

Genetic risk was assessed using the standard AMD PRS set provided by the UK Biobank, which was calculated on all individuals by meta-analysis of multiple external GWAS data [24]. Detailed procedures about genotyping, imputation, and quality control have been described previously [25]. In brief, the UK Biobank employed standardized subgroups and uniform definitions for diseases and quantitative traits to ensure consistent PRS evaluation. The AMD PRS values ranged from -4.32 to 4.94 , with a higher score indicating greater genetic predisposition. Participants were then categorized into low and high genetic risk based on the median of the PRS.

Assessment of outcomes

Incident cases of AMD were identified by linking to the national hospital inpatient admissions, where AMD was recorded as either the primary or secondary diagnosis according to the 9th Revision International Classification of Diseases (ICD-9: 3625) or 10th Revision (ICD-10: H35.3). Person-years of follow-up was calculated as the interval from the enrollment date to the time of the first AMD diagnosis, death, withdrawal from the study, or the end of the follow-up (31 December 2022), whichever came first.

Assessment of covariates

Several potential covariates were included based on the prior literatures. These covariates included sociodemographic factors (age, sex, ethnicity, education level, household income and paid employment status), lifestyles (body mass index [BMI], smoking status, healthy diet score, physical activity and alcohol intake), and self-reported history of chronic diseases (e.g., hyperlipidemia, hypertension and diabetes). The healthy diet score was computed based on the

following dietary components: fresh fruit, fish, vegetables, processed meat and unprocessed red meat [26]. The total diet score ranged from 0 to 5, with one point awarded for each favorable diet factor. More detailed information on covariate definitions is provided in supplementary methods. Additionally, missing data for categorical variables were assigned as a missing indicator category, while missing value for continuous variables were imputed with the median value.

Statistical analysis

Baseline characteristics of the participants were described using mean $\pm$ standard deviation (SD) for continuous variables and frequency (proportion) for categorical variables, while the concentrations of air pollution were presented as the median (interquartile range, IQR). Differences in basic information between groups stratified by AMD status were examined using Student's *t*-test, Mann-Whitney U-test, or Chi-square test where appropriate.

Cox proportional hazards model was applied to evaluate the hazard ratio (HR) and 95% confidence interval (CI) for AMD associated with the individual air pollutant, APS, and PRS. We incorporated several potential confounders for adjustment. In the basic model (Model 1), we only adjusted for age and sex. In the fully-adjusted model (Model 2), we further adjusted for educational level, ethnicity, BMI, household income, paid employment status, smoking status, alcohol intake, physical activity, healthy diet score, self-reported hypertension, diabetes and hyperlipidemia. Ambient air pollutants were analyzed as continuous variable (per SD increment) and categorical variable (quartiles). We conducted subgroup analyses by age at recruitment (< 60 , and ≥ 60 years) and tested the interaction effects between age group and air pollution using likelihood ratio tests. In genetic analysis, we further adjusted the genetic batches and the first ten genetic principal components. Schoenfeld residuals were used to verify the proportional hazard assumption. In addition, we explored the dose-response relationships between air pollution and risk of AMD by restricted cubic spline regression with three knots placed at the 10th, 50th, and 90th percentiles, and restricted pollutant concentrations to the top 99% percentile to avoid the possible influences of extreme values. Furthermore, we imputed population attributable fractions (PAFs) to quantify the proportion of AMD cases that could be theoretically prevented if air pollutant concentrations were reduced to the target level [27]. According to the World Health Organization (WHO) interim target guidelines, the cut-offs for NO_2 , $\text{PM}_{2.5}$, PM_{10} were 20, 10, and 20 $\mu\text{g}/\text{m}^3$, respectively [28]; while cutoffs for NO_x (42.15 $\mu\text{g}/\text{m}^3$) and $\text{PM}_{2.5-10}$ (6.11 $\mu\text{g}/\text{m}^3$) were their median concentrations (the two pollutants not included in

the WHO global air quality guidelines) [18]. The 95% CIs of PAFs were estimated by bootstrapping using 1000 independent replications.

To evaluate the joint effects of air pollution exposure and genetic factors on AMD risk, we divided participants into eight groups based on their PRS and air pollution quartiles. The reference group was defined as individuals with low genetic risk and exposure to the lowest quartile of air pollution. We further quantified the gene-environment interactions on additive scales using two indexes: the relative excess risk due to the interaction (RERI) and the attributable proportion due to the interaction (AP) [29]. The 95% CIs of RERI and AP did not include zero indicated the presence of additive interaction.

We conducted several sensitivity analyses to examine the robustness and reliability of the results. First, we reconstructed a new APS that only incorporated NO₂, NO_x, PM_{2.5}, PM₁₀. Second, we excluded incident AMD cases within the first two year of follow-up to minimize the potential reverse causality. Third, to minimize potential misclassification of long-term air pollution exposure, we restricted our analysis to participants who had resided at their current address for at least five years. Fourth, we re-examined the additive interactions between air pollutants and PRS on the risk of AMD in the sample weighted by inverse probability weighting (IPW) to minimize potential selection bias due to the missing data. Fifth, we further performed quantile-based g computation (Q-g comp) to estimate the effect of air pollutants mixture on incident AMD.

All analyses were performed using R software (version of 4.2.2). A two-tailed *P*-value < 0.05 was indicated statistical significance.

Results

Demographic characteristics of study population

Table 1 showed baseline characteristics of participants. Among the included 445,237 participants, the mean age was 56.56 ± 8.09 and 45.9% were men. During a median follow-up of 13.83 years, we observed 9,635 (2.16%) incident AMD events. Compared to participants without AMD, people who had incident AMD were more likely to be older, women, less educated, lower household income, unemployment, higher BMI, former smokers, physically inactive, more frequent alcohol intake and healthier diet, and have a history of hypertension, diabetes and hyperlipidemia. The media (IQR) concentrations of NO₂, NO_x, PM_{2.5}, PM₁₀, PM_{2.5-10} and APS was 26.06 (9.84), 42.15 (16.56), 9.93 (1.28), 16.02 (1.76), 6.11 (0.79), and 60.55 (9.85) µg/m³, respectively.

Association of air pollution with incident AMD

The associations of air pollutants with AMD were presented in Table 2. Schoenfeld residual test indicated that suggesting that the conditional proportional hazards assumption was satisfied (Fig. S2). Specifically, after multivariable adjustment, per SD increase in NO₂, NO_x, PM_{2.5}, and PM₁₀, the HR (95%CI) for incident risk of AMD was 1.04 (1.02, 1.06), 1.03 (1.01, 1.05), 1.04 (1.02, 1.07), and 1.02 (1.00, 1.04), respectively. However, we did not observe a significant linear association between per SD increase in PM_{2.5-10} and AMD (HR, 1.02; 95%CI, 0.99–1.04; *P*=0.117). When compared to participants in the lowest quartile (Q1) subgroup of NO₂, NO_x, PM_{2.5}, PM₁₀, and PM_{2.5-10}, those in the highest quartile had a separate HR (95%CI) of 1.12 (1.06, 1.19), 1.14 (1.08, 1.21), 1.15 (1.09, 1.22), 1.07 (1.01, 1.13), and 1.08 (1.02, 1.14).

APS was significantly associated with incident AMD in both continuous and quartile forms (Table 2). For each SD increase in the APS, the HR (95%CI) for AMD risk was 1.04 (1.02, 1.06). Participants in the highest quartile (Q4) of APS had a 16% higher risk of AMD than those in the lowest quartile [HR (95%CI)=1.16 (1.09–1.23)]. Subgroup analysis indicated that all five pollutants and APS demonstrated significant associations with AMD among elderly participants (all *P* for interaction < 0.001) (Table S2). The sensitive analysis demonstrated that the results were generally robust when excluding PM_{2.5-10} in the APS (Table S3), excluding participants with less than 2 years of follow-up (Table S4), limiting participants who had resided at their current address for at least five years (Table S5), using the Q-g comp to estimating the effect of air pollutants mixture (Figure S3).

The dose-response curves depicting the relationships between NO₂, NO_x, PM_{2.5}, and APS with incident AMD were almost non-linear (Fig. 1). Notably, the curves exhibited a significantly steeper slope at relatively lower concentrations of air pollutants, whereas the association plateaued or showed a modest decline at higher exposure levels. Moreover, reducing NO₂, NO_x, and PM_{2.5} to the interim target levels could lead to a significant reduction in AMD incidence, ranging from 3.40% to 8.20% (all *P*<0.05) (Table S6).

Association of PRS with incident AMD

After adjusting for various covariates, the incident risk of AMD associated with each per-SD increase in PRS of AMD was 1.46 (95% CI, 1.43–1.49). Similarly, when stratified by PRS, participants in the high genetic risk category had a 1.75-fold (HR, 1.75; 95% CI, 1.68–1.82) risk of AMD than those with the low genetic risk (Table S7).

Table 1 Baseline characteristics of participants

Characteristics	Total	Age-related macular degeneration	
		No	Yes
Number of participants	445,237	435,602	9635
Age (mean $\pm$ SD), years	56.56 $\pm$ 8.09	56.42 $\pm$ 8.08	62.89 $\pm$ 5.34
<i>Sex</i>			
Women	241,045 (54.1)	235,173 (54.0)	5872 (60.9)
Men	204,192 (45.9)	200,429 (46.0)	3763 (39.1)
<i>Ethnicity</i>			
White	418,514 (94.0)	409,379 (94.0)	9135 (94.8)
Asian	10,397 (2.3)	10,171 (2.3)	226 (2.3)
Black	7401 (1.7)	7285 (1.7)	116 (1.2)
Others	6762 (1.5)	6660 (1.5)	102 (1.1)
Missing	2163 (0.5)	2107 (0.5)	56 (0.6)
<i>Education</i>			
Low	75,248 (16.9)	72,722 (16.7)	2526 (26.2)
Intermediate	222,945 (50.1)	218,368 (50.1)	4577 (47.5)
High	141,247 (31.7)	138,853 (31.9)	2394 (24.8)
Missing	5797 (1.3)	5659 (1.3)	138 (1.4)
<i>Household income</i>			
< £18,000	86,487 (19.4)	83,694 (19.2)	2793 (29.0)
£18,000–30,999	96,985 (21.8)	94,627 (21.7)	2358 (24.5)
£31,000–51,999	98,778 (22.2)	97,149 (22.3)	1629 (16.9)
£52,000–100,000	76,705 (17.2)	75,834 (17.4)	871 (9.0)
> £100,000	20,318 (4.6)	20,161 (4.6)	157 (1.6)
Missing	65,964 (14.8)	64,137 (14.7)	1827 (19.0)
<i>Paid employment</i>			
No	187,441 (42.1)	180,869 (41.5)	6572 (68.2)
Yes	255,463 (57.4)	252,442 (58.0)	3021 (31.4)
Missing	2333 (0.5)	2291 (0.5)	42 (0.4)
<i>BMI, n (%)</i>			
<25 kg/m ²	146,405 (32.9)	143,583 (33.0)	2822 (29.3)
25–30 kg/m ²	190,625 (42.8)	186,458 (42.8)	4167 (43.2)
$\geq$ 30 kg/m ²	108,207 (24.3)	105,561 (24.2)	2646 (27.5)
<i>Smoking status</i>			
Never	242,468 (54.5)	237,752 (54.6)	4716 (48.9)
Former	154,558 (34.7)	150,540 (34.6)	4018 (41.7)
Current	45,903 (10.3)	45,076 (10.3)	827 (8.6)
Missing	2308 (0.5)	2234 (0.5)	74 (0.8)
<i>Physical activity</i>			
Low	68,186 (15.3)	66,723 (15.3)	1463 (15.2)
Intermediate	145,600 (32.7)	142,525 (32.7)	3075 (31.9)
High	145,405 (32.7)	142,456 (32.7)	2949 (30.6)
Missing	86,046 (19.3)	83,898 (19.3)	2148 (22.3)
<i>Alcohol intake</i>			
Daily or almost daily	91,270 (20.5)	89,129 (20.5)	2141 (22.2)
Three or four times a week	102,602 (23.0)	100,601 (23.1)	2001 (20.8)
Once or twice a week	114,166 (25.6)	111,946 (25.7)	2220 (23.0)
One to three times a month	49,343 (11.1)	48,373 (11.1)	970 (10.1)
Special occasions only	51,235 (11.5)	49,926 (11.5)	1309 (13.6)
Never	35,614 (8.0)	34,646 (8.0)	968 (10.0)
Missing	1007 (0.2)	981 (0.2)	26 (0.3)
<i>Healthy diet score</i>			
0–1	95,682 (21.5)	93,897 (21.6)	1785 (18.5)
2–3	278,328 (62.5)	272,336 (62.5)	5992 (65.2)
4–5	57,803 (13.0)	56,266 (12.9)	1537 (16.0)
Missing	13,424 (3.0)	13,103 (3.0)	321 (3.3)

Table 1 (continued)

Characteristics	Total	Age-related macular degeneration	
		No	Yes
<i>Self-reported hypertension</i>			
No	333,877 (75.0)	327,518 (75.2)	6359 (66.0)
Yes	111,360 (25.0)	108,084 (24.8)	3276 (34.0)
<i>Self-reported diabetes</i>			
No	421,651 (94.7)	412,965 (94.8)	8686 (90.2)
Yes	23,586 (5.3)	22,637 (5.2)	949 (9.8)
<i>Self-reported hyperlipidemia</i>			
No	365,801 (82.2)	358,927 (82.4)	6874 (71.3)
Yes	79,436 (17.8)	76,675 (17.6)	2761 (28.7)
NO ₂ level [$\mu\text{g}/\text{m}^3$, median (IQR)]	26.06 (9.84)	26.06 (9.85)	26.08 (9.54)
NO _x level [$\mu\text{g}/\text{m}^3$, median (IQR)]	42.15 (16.56)	42.15 (16.57)	42.15 (16.08)
PM _{2.5} level [$\mu\text{g}/\text{m}^3$, median (IQR)]	9.93 (1.28)	9.93 (1.28)	9.94 (1.24)
PM ₁₀ level [$\mu\text{g}/\text{m}^3$, median (IQR)]	16.02 (1.76)	16.02 (1.76)	16.03 (1.72)
PM _{2.5-10} level [$\mu\text{g}/\text{m}^3$, median (IQR)]	6.11 (0.79)	6.11 (0.79)	6.10 (0.81)
Air pollution score [$\mu\text{g}/\text{m}^3$, median (IQR)]	60.55 (9.85)	60.55 (9.85)	60.60 (9.45)

Educational levels were classified into three categories: low (no relevant qualifications); intermediate (A levels, O levels/GCSEs, CSEs, NVQ/HND/HNC, other professional qualifications); high (college or university degree). Physical activity status was classified into three categories: low [<600 metabolic equivalent (MET) minutes/week], moderate (600 to <3000 MET) or high (3000+ MET)

Interaction and joint association of air pollution and PRS with incident AMD

We then explored the relationship between air pollution and AMD among participants with different levels of PRS (Table 3). Specifically, the adverse effects of all five air pollutants and APS on AMD were generally higher among participants in the high PRS group than those with low PRS. Remarkably, significant associations between PM_{2.5} and APS with AMD risk was observed even in the low PRS group (Table 3). There were significant additive interactions between PRS and exposure to NO₂, NO_x, PM_{2.5-10}, and APS on incident risk of AMD (Table 3). For instance, among participants with the highest quartile (Q4) of APS and with high PRS, the RERI was 0.12 (95% CI: 0.03, 0.20) and the AP was 0.06 (95% CI: 0.01, 0.11), indicating that the additive interaction contributed a relative excess risk of 0.12 and accounted for 6% of the risk of AMD in this subgroup. Similar significant additive interactions were also noted for NO₂, NO_x, and PM_{2.5-10}, with RERI (95% CI) and AP (95% CI) of 0.10 (0.01, 0.18) and 0.05 (0.01, 0.11) for NO₂, 0.11 (0.01, 0.19) and 0.05 (0.02, 0.10) for NO_x, and 0.15 (0.06, 0.23) and 0.08 (0.03, 0.13) for PM_{2.5-10}, respectively. The sensitive analysis showed that these results remained consistent when the inverse probability weighting model was used (Table S8).

The joint effects of air pollution and genetic susceptibility on AMD risk were further tested and shown in Fig. 2. Specifically, compared with those with low PRS and exposed to lowest (Q1) quartile of air pollutants NO₂, NO_x, PM_{2.5}, PM₁₀, PM_{2.5-10}, and APS, those with high PRS and highest (Q4) quartile of these pollutants had the highest risk

of incident AMD, with the HR (95%CI)=1.90 (1.75, 2.07), 1.95 (1.79, 2.12), 2.03 (1.86, 2.21), 1.88 (1.73, 2.04), 1.83 (1.68, 1.99), 1.98 (1.82, 2.16), respectively.

Discussion

In this large-scale prospective study, we observed the significant positive associations of long-term exposure to NO₂, NO_x, PM_{2.5}, PM₁₀ and APS with increase incident risk of AMD. Non-linear exposure-response curves indicated that the risk of incident AMD increased more steeply at lower pollution levels and plateaued at higher exposures. Our results also demonstrated the additive interactions of exposure to NO₂, NO_x, PM_{2.5-10} and APS with genetic risk score on incident AMD, with enhanced adverse effects of air pollution among those with higher genetic susceptibility. In particularly, compared with participants with low PRS and lowest air pollution exposure, those with highest quartile air pollution exposure and high PRS had almost double the risk of AMD.

Ambient air pollution has emerged as a major environmental issue with substantial health implications, and some studies have linked the air pollution to age-related eye disease [30]. To date, few epidemiological studies have examined the associations between air pollution and AMD incidence, with inconsistent results. In line with our findings, a prospective cohort study included 39,819 participants in Taiwan reported that exposure to NO₂ was associated with an increased risk of AMD [9]. Similarly, a meta-analysis included eight studies indicated that exposure to NO₂ and PM_{2.5} was associated with an increased risk of AMD

Table 2 Association between air pollution and risk of age-related macular degeneration

Table 2 Association between air pollution and risk of age-related macular degeneration	Air pollution	Events/person year	Model 1		Model 2	
			HR (95%CI)	P	HR (95%CI)	P
HR, hazard ratio; CI, confidence interval; NO ₂ , nitrogen dioxide; NO _x , nitrogen oxides; PM _{2.5} , fine particulate matter with diameter <2.5 μm; PM ₁₀ , particulate matter with diameter <10 μm; PM _{2.5–10} , particulate matter with aerodynamic diameter between 2.5 and 10 μm ^a Air pollution score ranges: quartile 1: (43.96–55.67); quartile 2: (55.68–60.54); quartile 3: (60.55–65.52); and quartile 4: (65.53–156.80). Media (IQR) of the air pollution score is 60.55(9.85). Model 1 was adjusted for age and sex. Model 2 was further adjusted education level, ethnicity, BMI, household income, paid employment status, smoking status, alcohol intake, physical activity, healthy diet score, self-reported hypertension, diabetes and hyperlipidemia. Bold values indicate statistical significance at P<0.05.	<i>NO₂ level</i>					
	Q1	2358/1,536,239	Ref			
	Q2	2445/1,535,989	1.06 (1.00, 1.12)	0.059	1.03 (0.98, 1.09)	0.263
	Q3	2511/1,530,005	1.13 (1.07, 1.20)	<0.001	1.10 (1.04, 1.16)	0.001
	Q4	2321/1,523,509	1.17 (1.11, 1.24)	<0.001	1.12 (1.06, 1.19)	<0.001
	Per SD		1.06 (1.04, 1.08)	<0.001	1.04 (1.02, 1.06)	<0.001
	<i>NO_x level</i>					
	Q1	2313/1,535,876	Ref			
	Q2	2504/1,530,509	1.11 (1.05, 1.18)	<0.001	1.09 (1.03, 1.15)	0.003
	Q3	2437/1,525,519	1.14 (1.07, 1.20)	<0.001	1.10 (1.04, 1.16)	0.001
	Q4	2381/1,533,837	1.20 (1.13, 1.27)	<0.001	1.14 (1.08, 1.21)	<0.001
	Per SD		1.05 (1.03, 1.07)	<0.001	1.03 (1.01, 1.05)	0.002
	<i>PM_{2.5} level</i>					
	Q1	2265/1,515,558	Ref		Ref	
	Q2	2462/1,531,845	1.10 (1.04, 1.17)	<0.001	1.08 (1.02, 1.15)	0.007
	Q3	2443/1,528,060	1.14 (1.08, 1.21)	<0.001	1.10 (1.04, 1.17)	<0.001
	Q4	2465/1,550,279	1.21 (1.14, 1.28)	<0.001	1.15 (1.09, 1.22)	<0.001
	Per SD		1.06 (1.04, 1.08)	<0.001	1.04 (1.02, 1.07)	<0.001
	<i>PM₁₀ level</i>					
	Q1	2348/1,527,696	Ref		Ref	
	Q2	2402/1,531,463	1.04 (0.99, 1.10)	0.225	1.02 (0.96, 1.08)	0.478
	Q3	2481/1,531,703	1.11 (1.05, 1.18)	<0.001	1.09 (1.03, 1.15)	0.005
	Q4	2404/1,534,879	1.09 (1.03, 1.16)	0.002	1.07 (1.01, 1.13)	0.029
	Per SD		1.03 (1.01, 1.06)	<0.001	1.02 (1.00, 1.04)	0.020
	<i>PM_{2.5–10} level</i>					
	Q1	2291/1,487,454	Ref		Ref	
	Q2	2554/1,576,453	1.07 (1.01, 1.13)	0.016	1.06 (1.00, 1.12)	0.043
	Q3	2310/1,515,073	1.04 (0.98, 1.10)	0.220	1.02 (0.96, 1.08)	0.508
Q4	2480/1,546,761	1.10 (1.04, 1.17)	<0.001	1.08 (1.02, 1.14)	0.007	
Per SD		1.02 (1.00, 1.04)	0.043	1.02 (0.99, 1.04)	0.117	
<i>Air pollution score^a</i>						
Q1	2271/1,535,623	Ref		Ref		
Q2	2510/1,529,620	1.13 (1.07, 1.20)	<0.001	1.11 (1.05, 1.17)	<0.001	
Q3	2472/1,526,312	1.18 (1.11, 1.25)	<0.001	1.14 (1.08, 1.21)	<0.001	
Q4	2382/1,534,186	1.21 (1.14, 1.28)	<0.001	1.16 (1.09, 1.23)	<0.001	
Per SD		1.06 (1.04, 1.08)	<0.001	1.04 (1.02, 1.06)	<0.001	

HR, hazard ratio; CI, confidence interval; NO₂, nitrogen dioxide; NO_x, nitrogen oxides; PM_{2.5}, fine particulate matter with diameter < 2.5 µm; PM₁₀, particulate matter with diameter < 10 µm; PM_{2.5-10}, particulate matter with aerodynamic diameter between 2.5 and 10 µm

^aAir pollution score ranges: quartile 1: (43.96–55.67); quartile 2: (55.68–60.54); quartile 3: (60.55–65.52); and quartile 4: (65.53–156.80). Media (IQR) of the air pollution score is 60.55(9.85). Model 1 was adjusted for age and sex. Model 2 was further adjusted education level, ethnicity, BMI, household income, paid employment status, smoking status, alcohol intake, physical activity, healthy diet score, self-reported hypertension, diabetes and hyperlipidemia. Bold values indicate statistical significance at *P* < 0.05.

[31]. A recent review further highlighted that exposure to air pollutants can accelerate or exacerbate the prevalence and morbidity of AMD [32]. In our study, we did not find a significant linear association of PM_{2.5-10} and incident AMD, which is similar to prior findings [33]. This null result is biologically reasonable because PM_{2.5-10} originates mainly from mechanical resuspension of crustal and road dust, resulting in significant compositional and source heterogeneity that reduces its toxicological potency and increases exposure-measurement error [34]. These factors may collectively reduce the observable epidemiological relationship relative to PM_{2.5}. Chua et al. only found a significant association between PM_{2.5} exposure and self-reported AMD in a cross-sectional study based on European participants, but not for NO₂, NO_x, PM₁₀, and PM_{2.5-10} [11], which may

be due to the cross-sectional design that overlooks long-term air pollution effects and the reliance on self-reported AMD data. In addition, a national cohort study in Korea observed the positive association between NO₂ exposure and AMD risk but did not find a significant association between PM₁₀ exposure and AMD [35]. Overall, the heterogeneity among studies may be attributable to the differences of study designs, population characteristics, geographical conditions, and exposure assessment methods.

Our research comprehensively demonstrated the significantly increased risk of incident AMD associated with the air pollution score, which reflects the combined effects of various air pollutants co-exposure. Similar integrated scores have been applied in the field of environment health [23, 36] and epidemiological studies [37, 38], and previous studies

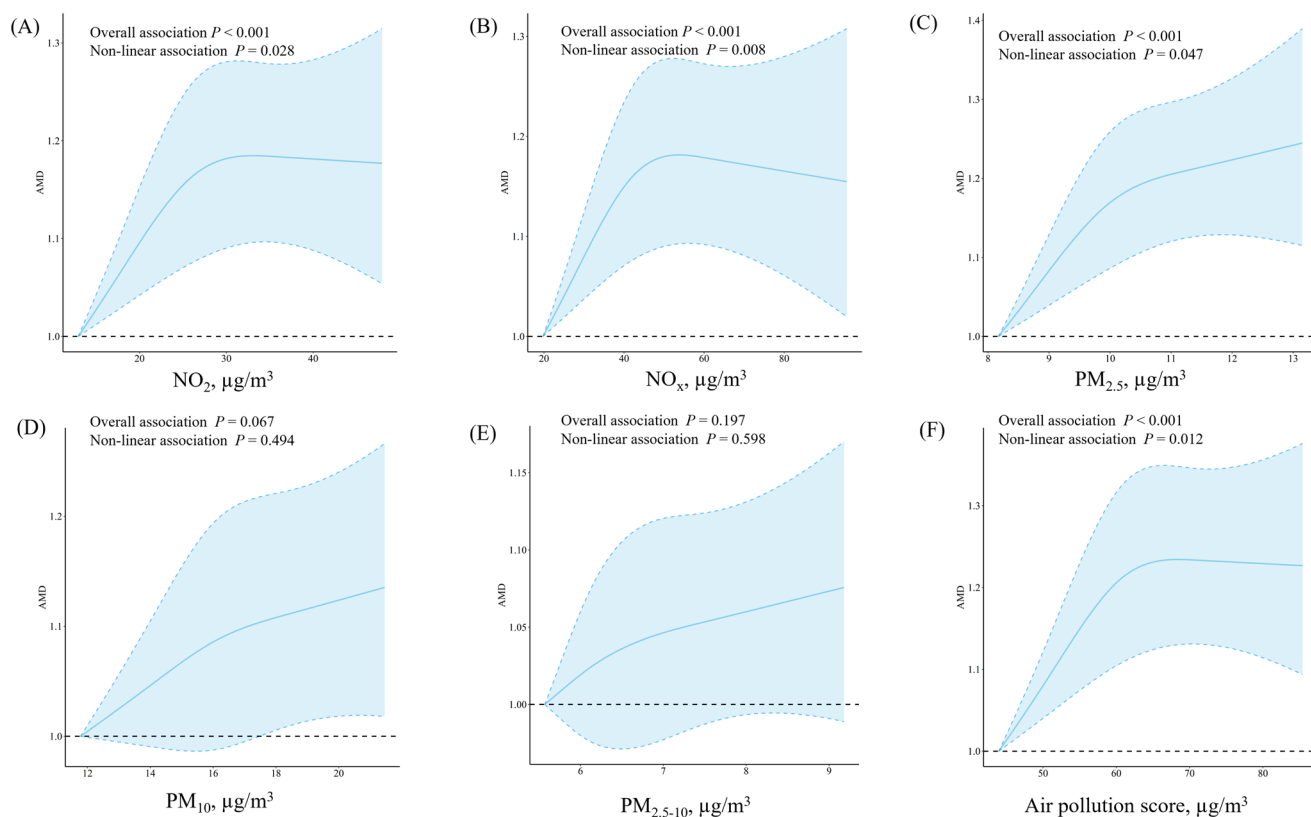


Fig. 1 Dose-response associations between air pollution and age-related macular degeneration. A restricted cubic spline regression model with three knots (at the 10th, 50th and 90th percentiles) was used to estimate the dose-response relations between air pollutants and the risk of age-related macular degeneration. We restricted pollutant concentrations to the top 99% percentile to mitigate the influ-

ence of extreme values. Hazard ratios (HRs) solid lines) and 95% CIs (shaded areas) were adjusted for adjusted for age, sex, ethnicity, education level, household income, paid employment status, BMI, smoking status, alcohol intake, physical activity, healthy diet score, self-reported hypertension, diabetes and hyperlipidemias

have validated that this method is relatively reliable and accurate compared to other common methods for estimating exposure to mixture pollutants within the UK Biobank [39]. Nonlinear exposure-response curves were observed between NO_2 , NO_x , $\text{PM}_{2.5}$, and APS and incident AMD, characterized by a steeper slope at lower exposure levels and a plateau or slight decline at higher concentrations of air pollution. These findings suggest that even exposure to air pollution levels below the UK air quality standards may still increase AMD risk. More importantly, in alignment with WHO recommendations [28], reducing air pollutants to the interim target levels could potentially decrease the incidence of AMD by 3.40%–8.20%. Our results highlight the imperative to formulate stricter air pollution control policies, and reducing exposure to multiple air pollutants may lead to a reduction of incident AMD.

The precise mechanisms linking AMD to ambient air pollution remain incompletely understood. However, emerging evidence suggests that chronic inflammation and oxidative stress may represent critical mechanistic pathways in AMD pathogenesis [15, 40]. The retina's high oxygen demand and

abundance of polyunsaturated fatty acids render it highly susceptible to oxidative damage [41]. Mechanistic studies have shown that $\text{PM}_{2.5}$ exposure triggers reactive oxygen species (ROS)-driven epithelial-mesenchymal transition (EMT) in human retinal pigment epithelium (RPE) cells, resulting in a fibroblast-like phenotypic shift and enhanced cell migration [42]. EMT is implicated in RPE disorders and the development of intraocular fibrotic conditions, including AMD [43]. Additionally, animal studies have revealed that $\text{PM}_{2.5}$ impairs retinal microvascular function [44], which is a major contributor to AMD pathogenesis [45]. Prior epidemiological studies have demonstrated that both long-term and short-term exposure to $\text{PM}_{2.5}$ and NO_x increases levels of inflammatory markers [46, 47]. Other population-based studies have also reported that $\text{PM}_{2.5}$, PM_{10} , NO_2 and NO_x were all associated with adverse retinal structure [11, 48].

Our study is the first to explore the interactive and joint effects between genetic risk and air pollution on AMD risk. We observed the significant additive interactions, indicating that the additive effects of high air pollution exposure and high genetic risk exceeded the sum of their individual

Table 3 Additive interactions between air pollutants and PRS on the risk of age-related macular degeneration

Air pollution	Low genetic risk (<i>n</i> =222618)		High genetic risk (<i>n</i> =222619)		RERI (95%CI)	AP (95%CI)
	HR (95%CI)	<i>P</i>	HR (95%CI)	<i>P</i>		
<i>NO₂ level</i>						
Q1	Ref					
Q2	1.00 (0.91, 1.10)	0.945	1.04 (0.97, 1.12)	0.294	0.06 (−0.03, 0.15)	0.04 (−0.02, 0.09)
Q3	1.07 (0.97, 1.17)	0.183	1.10 (1.02, 1.18)	0.01	0.09 (0.00, 0.17)	0.05 (−0.01, 0.10)
Q4	1.06 (0.96, 1.17)	0.259	1.12 (1.04, 1.21)	0.002	0.10 (0.01, 0.18)	0.05 (0.01, 0.11)
<i>NO_x level</i>						
Q1	Ref		Ref			
Q2	1.06 (0.97, 1.17)	0.201	1.10 (1.02, 1.17)	0.012	0.10 (0.01, 0.18)	0.05 (0.00, 0.10)
Q3	1.07 (0.97, 1.18)	0.157	1.10 (1.02, 1.18)	0.013	0.08 (−0.01, 0.17)	0.04 (−0.01, 0.10)
Q4	1.10 (1.00, 1.21)	0.057	1.14 (1.06, 1.22)	<0.001	0.11 (0.01, 0.19)	0.05 (0.02, 0.10)
<i>PM_{2.5} level</i>						
Q1	Ref		Ref			
Q2	1.13 (1.03, 1.24)	0.013	1.05 (0.98, 1.13)	0.158	−0.04 (−0.14, 0.06)	−0.02 (−0.07, 0.03)
Q3	1.09 (0.99, 1.20)	0.072	1.10 (1.02, 1.18)	0.014	0.07 (−0.03, 0.15)	0.03 (−0.02, 0.09)
Q4	1.15 (1.04, 1.27)	0.005	1.13 (1.05, 1.22)	<0.001	0.07 (−0.03, 0.16)	0.03 (−0.01, 0.08)
<i>PM₁₀ level</i>						
Q1	Ref		Ref			
Q2	1.05 (0.96, 1.15)	0.298	1.01 (0.94, 1.08)	0.875	−0.05 (−0.10, 0.05)	−0.03 (−0.08, 0.03)
Q3	1.07 (0.97, 1.17)	0.173	1.08 (1.01, 1.16)	0.028	0.06 (−0.04, 0.14)	0.03 (−0.02, 0.08)
Q4	1.03 (0.94, 1.14)	0.497	1.08 (1.00, 1.16)	0.050	0.07 (−0.03, 0.15)	0.04 (−0.01, 0.09)
<i>PM_{2.5–10} level</i>						
Q1	Ref		Ref			
Q2	0.99 (0.90, 1.09)	0.861	1.09 (1.02, 1.17)	0.015	0.16 (0.08, 0.24)	0.09 (0.04, 0.15)
Q3	0.99 (0.90, 1.08)	0.785	1.03 (0.96, 1.11)	0.422	0.05 (−0.04, 0.14)	0.03 (−0.02, 0.09)
Q4	1.01 (0.92, 1.11)	0.821	1.11 (1.03, 1.19)	0.005	0.15 (0.06, 0.23)	0.08 (0.03, 0.13)
<i>Air pollution score</i>						
Q1	Ref		Ref			
Q2	1.09 (0.99, 1.20)	0.067	1.11 (1.03, 1.19)	0.005	0.09 (−0.01, 0.18)	0.05 (−0.01, 0.10)
Q3	1.11 (1.01, 1.22)	0.032	1.14 (1.06, 1.23)	<0.001	0.12 (0.02, 0.20)	0.06 (0.01, 0.11)
Q4	1.11 (1.01, 1.22)	0.037	1.15 (1.07, 1.24)	<0.001	0.12 (0.03, 0.20)	0.06 (0.01, 0.11)

Model was adjusted for age, sex, education level, ethnicity, BMI, household income, paid employment status, smoking status, alcohol intake, physical activity, healthy diet score, self-reported hypertension, diabetes and hyperlipidemia, genetic batch and 10 genetic principal components. Bold values indicate statistical significance at *P*<0.05.

HR, hazard ratio; CI, confidence interval; NO₂, nitrogen dioxide; NO_x, nitrogen oxides; PM_{2.5}, fine particulate matter with diameter<2.5 μm; PM₁₀, particulate matter with diameter<10 μm; PM_{2.5–10}, particulate matter with aerodynamic diameter between 2.5 and 10 μm

effects. Furthermore, 5–9% of AMD risk could be attributed to the additive interactions. Our results also revealed that higher levels of air pollution and greater genetic susceptibility were associated with a higher risk of AMD. Consistent with our results, a twin study revealed the significant role of genetic factors and unique environmental influences, highlighting the gene-environment interplay in AMD etiology [19]. Our findings provide novel insights into the precise prevention and control of AMD, as they can be used to identify individuals who may benefit most from targeted interventions to reduce air pollution exposure, particularly among those with high genetic risk. Future research should take a more novel approach to exploring the impact of air pollution and genetic interactions on AMD development [49].

Strengths of the present study included that we comprehensively assessed the associations of individual and combined exposure to various air pollutants with AMD risk based on the prospective design and large sample size. In addition, we firstly explored the interactive and joint effects of air pollution and genetic susceptibility on AMD incidence, enabling a more accurate evaluation of air pollution effects across different susceptibility levels. However, the current study has several limitations. First, this observational study was unable to determine a causal relationship between air pollution and AMD, despite adjusting for a range of confounding factors, including demographics, socioeconomic status, lifestyle factors, and disease conditions. Second, air pollution exposure was measured at baseline, which may have overlooked changes in air pollution levels during the follow-up period, leading to misclassification of exposure.

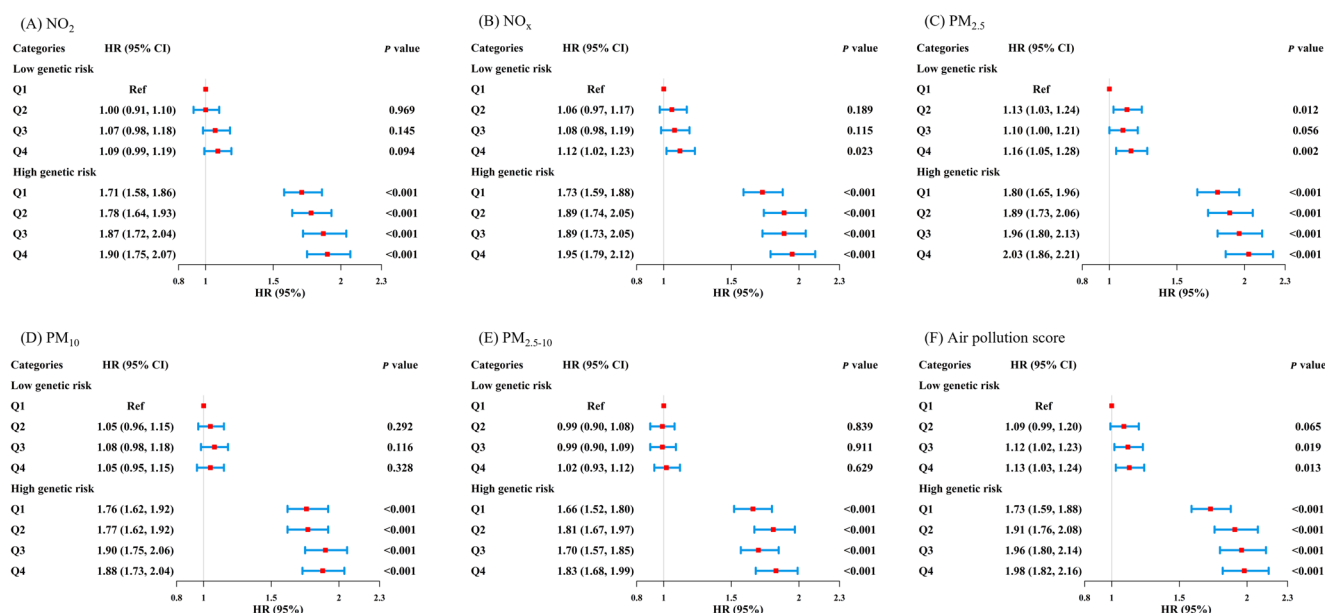


Fig. 2 Joint effect of air pollution score and genetic risk on incident age-related macular degeneration. Note: Model was adjusted for age, sex, education level, ethnicity, BMI, household income, paid employ-

ment status, smoking status, alcohol intake, physical activity, healthy diet score, self-reported hypertension, diabetes and hyperlipidemia, genetic batch and 10 genetic principal components

However, air pollution emissions in the UK remained relatively stable from 2010 to 2019 [14]. Also, the use of a single-year LUR estimate as a surrogate for long-term residential exposure is a well-established methodology in large-scale epidemiological studies of chronic diseases [14, 18, 23]. More cohort studies with repeated measurements of air pollution at different time are required to validate our findings. Third, the air pollution score was estimated by computing the individual pollutants weights by treating each as a continuous variable, whereas the air pollutants may be non-linearly associated with AMD. However, this method has been widely used in previous UK Biobank studies and has been shown to be reliable compared to other methods [23, 36, 39]. Fourth, the air pollution score did not include all air pollutants, such as O₃, which may also be related to the risk of AMD [7]. Fifth, we ascertained incident AMD cases solely using hospital inpatient records, which may have underestimated the prevalence of AMD. Finally, most of participants in our study were of European descent, generalization of the results to other ethnic populations should be undertaken with caution.

Conclusion

In summary, this large-scale prospective study is the first to demonstrated that long-term exposure to NO₂, NO_x, PM_{2.5}, PM₁₀ was significantly associated with an increased risk of AMD, and these effects were more pronounced among subjects with high genetic susceptibility. In addition, our

findings further reveal additive interactions of air pollution and PRS, highlighting the importance of improving air quality for preventing the onset of AMD, particularly among those with high genetic risk. Further studies are needed to validate our findings in different race populations and elucidate the underlying biological mechanisms.

Supplementary Information The online version contains supplementary material available at <https://doi.org/10.1007/s10654-025-01340-8>.

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Author contributions Shengli Chen: Conceptualization, Methodology, Formal analysis, Writing - original draft, Writing - review & editing; Gongyue Wang: Methodology, Writing - review & editing; Xin Guan, Formal analysis, Writing - review & editing; Chenming Wang, Methodology, Writing - review & editing; Yang Xiao: Writing - review & editing; Xingdi Li: Writing - review & editing; Shiru Hong: Data curation; Yuhua Zhou: Methodology, Visualization; Yingqian You, Visualization; Ye Fu: Visualization; Yuxi Wang: Writing - review & editing; Yichi Zhang: Visualization; Hui Zhao: Writing - review & editing; Yingchen Zhang: Data curation; Yang Cheng: Conceptualization; Huan Guo: Project administration, Conceptualization, Writing - original draft, Writing - review & editing, Funding acquisition; Huatao Xie: Methodology, Conceptualization, Writing - review & editing, Funding acquisition.

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Data availability This research was conducted using the UK Biobank

resource under the UKB application No. 8815. The data used in this study are available to all researchers upon application.

Declarations

Conflict of interest We declare no competing interests.

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