Causal Associations Between Dietary Habits and Cardiovascular Disease: A Mendelian Randomization Study

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Abstract

Over the years, numerous observational studies have substantiated that various dietary choices have opposing effects on cardiovascular disease. However, the causal effect has not yet been established. Thus, we conducted a Mendelian randomization (MR) analysis to reveal the causal impact of dietary habits on cardiovascular disease (CVD). Genetic variants strongly associated with 20 dietary habits were selected from publicly available genome-wide association studies conducted on the UK Biobank cohort (n = 449,210). Summary-level data on CVD were obtained from the FinnGen study, MEGASTROKE consortium, CARDIoGRAMplusC4D consortium, HERMES consortium, and other consortia (n=159,836 - 977,323). The inverse variance weighted method (IVW) was the primary outcome, while MR-Egger, weighted median, and MR Pleiotropy RESidual Sum and Outlier were used to assess heterogeneity and pleiotropy. We found compelling evidence of a protective causal effect of genetic predisposition towards cheese consumption on myocardial infarction (IVW OR = 0.67; 95% CI = 0.544-0.826; P = 1.784×10^{-4}) and heart failure (IVW OR = 0.646; 95% CI = 0.513-0.814; P = 2.135×10^{-4}). Poultry intake was found to be a detrimental factor for hypertension (IVW OR = 4.306; 95% CI = 2.158-8.589; P = 3.416×10^{-5}), while dried fruit intake was protective against hypertension (IVW OR = 0.473; 95% CI = 0.348-0.642; P = 1.683×10^{-6}). Importantly, no evidence of pleiotropy was detected. Mendelian randomization estimates provide robust evidence for a causal relationship between genetic predisposition to different dietary habits and CVD risk, suggesting that well-planned diets may help prevent and reduce the risk of CVD.

Keywords: Mendelian randomization; Cardiovascular disease; Dietary habits; GWAS.

Acronyms: MR: Mendelian randomization; CVD: cardiovascular disease; IVW: inverse variance weighted method; GWAS: genome-wide association studies; CHD: coronary heart disease; HP: hypertension; HF: heart failure; MI: myocardial infarction; AF: atrial fibrillation; IS: ischemic stroke; SNPs: single-nucleotide polymorphisms; MRPs: Milrad reaction products; AGEs: advanced glycation end products.
1. Introduction

Although the past few years have witnessed remarkable progress in the management of cardiovascular disease (CVD), it continues to be a major global health burden with high disability and mortality rates\(^1\). Current estimates indicate that the prevalence of cardiovascular disease in the United States is increasing, with approximately 85.6 million Americans currently living with CVD\(^2\). Accordingly, further emphasis should be placed on preventing the occurrence and development of the disease, and a healthy lifestyle throughout life is widely thought to be the most important way to prevent CVD\(^3\).

A healthy dietary habit may play the most important role in preventing premature death and disability of CVD compared to other habits, such as avoiding smoking and moderate physical activities. Ample evidence substantiates that dietary habits may influence the risk of CVD. A comprehensive review indicated that the Mediterranean diet is associated with better cardiovascular health outcomes, including decreased prevalence of coronary heart disease, ischemic stroke, and cardiovascular disease in general\(^4\). Two randomized clinical studies revealed that vegetarian dietary patterns could reduce the risk of CVD mortality and morbidity\(^5, 6\). In addition, dietary patterns with proinflammatory potential were positively associated with CVD risk in a prospective observational study\(^7\). It is widely thought that cardiometabolic risk factors affected by diet include obesity, lipoprotein function and concentration, glucose-insulin homeostasis, inflammation, oxidative stress, endothelial health, microbiome, and others\(^8\).

While studies have shown promising results regarding the benefits of healthy dietary patterns in preventing and managing cardiovascular disease, no consensus has been reached on the effects of individual food items or categories on CVD. For instance, some intervention studies have shown that blood pressure reduction (diastolic/systolic blood pressure > 21.5/24.0 mm Hg) was significantly associated with an increased fruit and vegetables (FV) intake\(^9,11\), while other studies revealed little or no effect\(^12, 13\). Similarly, research on the association between consumption of unprocessed red meat (i.e., beef, pork) and coronary heart disease (CHD) has yielded heterogeneous findings. In this respect, three meta-analyses based on observational prospective studies revealed inconsistent results; one reported a significant 27% increase in CHD incidence with red meat\(^14\), while the other two studies showed no association\(^15, 16\). In addition, several other potential risk factors for CHD were identified through observational studies and randomized controlled trials, including salt intake\(^17\), less coffee\(^18\) and tea intake\(^19\), decreased cheese intake\(^20\) and fish intake\(^16\). Conversely, studies also showed no association between CVD and these risk factors\(^21-25\).
These inconsistencies may be due to the presence of potential biases, such as confounding factors or reverse causality. Therefore, whether dietary habits play a causal role in the development of CVD remains undetermined. Mendelian randomization (MR) offers a novel approach to address these potential biases by utilizing genetic variants as instrumental variables to assess the causal relationship between exposure to risk factors and disease outcomes\(^\text{26, 27}\). Recent genome-wide association studies (GWAS) have shown that dietary habits are heritable traits\(^\text{28, 29}\). Besides, a two-sample MR study confirmed a causal relationship between dietary influence and major depressive disorder\(^\text{30}\). Herein, we conducted an MR analysis to investigate the causality of genetic liability for 20 dietary habits on the risk of developing coronary heart disease, hypertension (HP), heart failure (HF), myocardial infarction (MI), atrial fibrillation (AF), and ischemic stroke (IS).

2. Materials and Methods

Study Design

We applied a two-sample MR approach to investigate the causal relationship between dietary habits and CVD, including CHD, HP, MI, HF, AF and IS. The study framework is shown in Figure 1. To ensure the validity of our MR analysis, we considered three fundamental assumptions: (1) The genetic variants used as instrumental variables are strongly associated with the exposure of interest (20 dietary habits in this study); (2) The instrumental variables have an impact on the outcome (CAD in this study) only through the exposure; (3) The instrumental variables are independent of other factors that influence the outcome\(^\text{31}\).

2.1 Data sources for dietary habits

The genome-wide associations for 20 dietary habits used in our study were derived from UK Biobank (UKBB) GWAS summary statistics provided by Benjamin Neale’s lab (http://www.nealelab.is/uk-biobank/)\(^\text{32}\). UKBB is a large prospective study with over 500,000 participants aged 40–69 years recruited between 2006 and 2010. To prevent selective reporting bias, we evaluated all dietary habit variables (n=20) in the UKBB GWAS results obtained from Neil’s laboratory to explore potential causal associations with cardiovascular disease. At the same time, we conducted an extensive review of relevant studies, as mentioned in the introduction. We observed that the impact of dietary variables on heart disease remains controversial in observational studies. It is worth noting that certain dietary factors have not been extensively studied in relation to heart disease (such as hot drink temperature). In the UKBB, information on the 20 dietary habits was collected retrospectively by a shortened food frequency touchscreen questionnaire at baseline. Data on
the 20 single food intakes consist of quantitative continuous variables, (the number of heaped tablespoons of cooked vegetables per day), categorical variables (the main type of food consumed), and ordinal non-quantitative variables (frequency of consumption of oily fish). The questions on dietary habits and their corresponding answers are summarized in Supplementary Table S1, based on publicly available information on the UKBB website (http://biobank.ctsu.ox.ac.uk/crystal/label.cgi?id=100052). The submission would not be accepted if the answer were unrealistic.

2.2 Selection of genetic instruments

After obtaining GWAS data for 20 dietary habits from the UKBB, we performed various quality control procedures to select appropriate genetic instruments. Firstly, we extracted single nucleotide polymorphisms (SNPs) that were highly associated with each dietary habit \((P < 5 \times 10^{-8})\). Secondly, we conducted Linkage disequilibrium (LD)-based clumping with \(r^2 < 0.01\) and set a window size of 10,000 kb to evaluate whether there is a Linkage disequilibrium between snps of one trait located on the same chromosome. In this case, SNPs with the largest \(P\) values were discarded to ensure the independence between IVs. Thirdly, we assessed the strength of the genetic instruments and mitigated potential weak instrument bias by calculating \(F\)-statistics. The \(F\)-statistics were computed using the formula: \(F = [R^2 \times (N - k - 1)] / [k \times (1 - R^2)]^{33}\). We calculated the \(R^2\) using the following equation: \(R^2 = [2 \times EAF \times (1 - EAF) \times \beta^2] / [2 \times EAF \times (1 - EAF) \times \beta^2 + 2 \times EAF \times (1 - EAF) \times n \times se^2]\). A threshold of an \(F\)-statistic higher than 10 indicated a strong genetic variant. Additionally, we excluded SNPs associated with the outcome traits (at \(P < 5 \times 10^{-8}\)) and not present in the outcome data. We employed the MRPRESSO method to identify and remove SNPs and outliers to address potential horizontal pleiotropy. Furthermore, we used Phenoscanner (www.phenoscanner.medschl.cam.ac.uk) to exclude SNPs associated with confounding factors to account for potential confounding.

2.3 Data sources for cardiovascular disease

Summary genetic statistics for CHD and MI were extracted from a large-scale GWAS meta-analysis from the CARDIoGRAMplusC4D consortium, which included up to 184,305 subjects across 48 studies (77% of whom were of European descent, 60,801 CHD cases and 123,504 controls; 43,676 MI cases and 128,199 controls).\(^{34}\) Summary level-data for stroke were obtained from the MEGASTROKE consortium, compromising 446,696 participants of European descent (405,855 cases and 406,111 controls). Summary statistics for AF were derived from by a meta-analysis of GWAS, which compromised 55,114 cases and 482,295 control samples of European ancestry.\(^{35}\) Summary data for HF were obtained from the
HERMES consortium, comprising 977,323 participants of European descent (47,309 cases and 930,014 controls) from 26 studies. HP GWAS data were extracted from the FinnGen Study, compromising 13 cohorts and biobanks. The data were adjusted for age, sex, genotyping batch, and 10 principal components. 15,870 cases were defined as having essential or primary hypertension, and 74,345 controls were without essential hypertension or any other hypertensive diseases. The detailed definition of essential hypertension and outcome data resources are provided in Table S3 and S4. We retrieved these summary data and extracted the SNPs associated with the 20 dietary habits.

2.4 Mendelian randomization analyses

After harmonizing the SNPs between exposures and outcomes and excluding palindromic SNPs with an allele frequency > 0.3, we conducted three different approaches of MR, including random-effect inverse-variance weighted (IVW), MR Egger, and weighted median to assess the effect of variant heterogeneity and pleiotropy. We used the IVW method as the main MR analysis, which estimates the causal effect by performing a weighted regression of the SNP-outcome effect on the SNP exposure effect. In this method, the intercept is constrained to zero, assuming there are no other pathways through which the genetic variants affect the outcome besides the exposure of interest. We conducted additional MR analyses, including MR-Egger and weighted median, to enhance the robustness of our findings, although these methods may yield wider confidence intervals due to lower efficiency (36). In cases where the results of these three methods were inconsistent, a stricter instrument p-value threshold was implemented to reperform another MR analysis. We further examined the horizontal pleiotropy using the MR-Egger intercept test and leave-one-out analyses.

Sensitivity analysis is crucial to detect the pleiotropy and heterogeneity for MR estimates. Cochran’s Q test (p < 0.05) from the IVW method is used to represent heterogeneity. We further examined the intercept obtained from the MR-Egger regression and leave-one-out analyses for horizontal pleiotropy (p < 0.05) (37). A funnel plot was also conducted to examine the potential directional pleiotropy. SNPs and outliers identified by MRPRESSO were removed to correct horizontal pleiotropy.

We performed power analyses for significant and nominally significant MR via an online tool (https://shiny.cnsgenomics.com/mRnd/) to assess the likelihood of a non-zero causal relationship between the dietary habits and their corresponding cardiovascular diseases (Supplementary Table S5).
2.5 Statistical analysis

Statistical analysis was performed using the two-sample MR package (version 0.5.6) and MRPRESSO (version 1.0) in R (version 4.2.2). We checked for potential mediators, including obesity, body mass index, serum lipids, type 2 diabetes, and smoking in PhenoScanner (www.phenoscanner.medschl.cam.ac.uk) to assess their influence on the SNPs. A threshold of \( P < 0.000417 \) (Bonferroni correction \( P = 0.05/120 \)) was considered significant for establishing a causal relationship. The risk of CVDs was quantified using odds ratios (OR) with a 95% confidence interval (CI) per unit in our MR results.

3. Results

The F statistics of all extracted SNPs ranged from 29 to 811, indicating strong instruments. Besides, all power values were greater than 0.8, indicating a high confidence level in the results' significance.

In our comprehensive MR analysis investigating the genetic predisposition towards dietary habits and their influence on CVD, we identified several significant or nominally significant associations (Table 1). We found evidence of a protective causal effect of genetic predisposition towards cheese intake on MI (IVW OR = 0.67; 95% CI = 0.544-0.826; \( P = 1.784 \times 10^{-4} \); 60 SNPs) and HF (IVW OR = 0.646; 95% CI = 0.513-0.814; \( P = 2.135 \times 10^{-4} \); 53 SNPs). Poultry intake was found to be a detrimental factor for HP (IVW OR = 4.306; 95% CI = 2.158-8.589; \( P = 3.416 \times 10^{-5} \); 6 SNPs), while dried fruit intake was protective against HP (IVW OR = 0.473; 95% CI = 0.348-0.642; \( P = 1.683 \times 10^{-6} \); 28 SNPs). Except for the effect of cheese on HF, the Cochran-Q test derived \( P \) values were > 0.05. Since we used the random effect IVW as the primary outcome, the observed heterogeneity was acceptable\(^{(38)}\). The MR-Egger intercept (\( P > 0.05 \)) confirmed the absence of directional pleiotropy was identified (Figure 2). Furthermore, the leave-one-out plot indicated no individual SNPs influenced the pooled IVW estimates (Figure 3).

Additionally, we identified several nominally significant variables in our analysis. Cooked vegetables, coffee, and pork consumption were associated with an increased risk of CHD. Coffee, cooked vegetables, and alcohol intake frequency were associated with increased risk of MI and HP, respectively (Table 1). To assess the presence of horizontal pleiotropy, we conducted Cochran-Q tests, leave-one-out plots, MR-Egger intercept tests, and funnel plots (Supplementary Figure S1-3). The MR-Egger intercept tests yielded \( P \)-values greater than 0.05, indicating no evidence of horizontal pleiotropy.

To assess potential confounding effects, we performed a sensitivity analysis by removing specific SNPs (rs10938397, rs13107325, rs1051730, rs1421085, and rs782221) that were...
associated with known confounders such as obesity, body traits, lipid levels, type 2 diabetes, and smoking, or directly affected MI, HF, or HP. After removing these SNPs, we re-evaluated the MR results and found that the causal associations remained consistent with the previous findings.

4. Discussion

This study represents the first and largest MR study to explore the causal relationship between 20 dietary habits and CVD. Our results suggest that different dietary intakes may have protective or deleterious effects on CVD, consistent with the literature. We found that certain dietary intakes may have protective effects on CVD, such as cheese intake on myocardial infarction and heart failure and dried fruit intake on hypertension. Conversely, we observed the deleterious effects of poultry intake on HP. Furthermore, our study revealed a higher risk of coronary heart disease associated with increased intake of cooked vegetables, coffee, and pork. Besides, coffee, cooked vegetable, and alcohol intake frequency were associated with increased risk of MI and HP.

Dairy fat, characterized by a high content of saturated fatty acids, is widely thought to increase coronary heart disease (CHD) mortality (38). Hence, only low-fat dairy products are recommended for a healthy diet. Surprisingly, dairy fat intake was associated with ischemic heart disease when cheese was excluded from the analysis (39), while several related studies have found no or even an inverse association between cheese and CVD (40, 41). The inconsistency in the results of previous studies may be explained by the heterogeneity in study design and methods, including different food frequency questionnaires, population differences, and other residual confounding factors. A two-sample Mendelian randomization study, utilizing single nucleotide polymorphisms as instrumental variables provides a robust approach to assess the causal relationship between cheese intake and cardiovascular disease while minimizing the impact of residual confounding factors. In the present study, we found that cheese intake has a beneficial effect on IM and HF, consistent with the results of previous studies (42). Several potential mechanisms may explain the protective effects of cheese on CVD. Firstly, cheese is rich in calcium, which has been reported to reduce total and LDL cholesterol concentrations (43) by binding to intestinal fatty acids to form insoluble soaps that may inhibit fatty acid absorption. Moreover, calcium is reportedly negatively correlated with stroke (44) and hypertension (45). Secondly, cheese contains a variety of bioactive proteins that protect vascular endothelial function through a variety of putative mechanisms, which include reducing NADPH oxidase (39), inhibiting lipid peroxidation (40), scavenging free radicals (41, 42) and enhancing antioxidant enzyme capacity (43, 44). These studies suggest that dairy proteins
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may protect endothelial function by limiting the production of reactive oxygen species. In addition, microarray analysis of the aorta from a spontaneously hypertensive rat model showed that targeted changes in the expression of genes related to vascular function were alterations in pro-/ anti-inflammatory transcription factors and increases in endothelial NO synthase and gap junction protein 40, including peroxisome proliferator-activated receptor γ and NF-κB, respectively (45). Thirdly, cheese is a probiotic food, containing many live microorganisms. A clinical trial (47) demonstrated that two probiotics in cheese, Lactobacillus and Bifidobacterium, yielded beneficial effects on immunity, inflammation, and cardiovascular risk factors.

Interestingly, we substantiated the harmful effect of poultry intake on HP, but consumption of beef, pork, processed meat and lamb/mutton did not significantly influence HP. There is an increasing consensus suggesting that unprocessed red meat increases the risk of hypertension (49, 50), although there are conflicting data on the relationship between blood pressure and poultry. Poultry consumption has been associated with an increased risk of HP in some studies (51), while others reported contrasting results (52). Three prospective cohorts from America (52) consistently demonstrated a significant association between animal flesh consumption and an increased risk of HP, independent of other dietary factors. Besides, higher consumption of poultry was associated with a higher risk for HP in two of the three cohorts. However, our results differ from previous studies in terms of the positive association observed between processed and red meat consumption and hypertension. The disparity in findings could be attributed to the inherent limitations and biases present in observational studies. It is highly conceivable that previous studies might not have consistently classified meat types, which could have contributed to the inconsistent outcomes. The increased risk of hypertension in poultry may be attributed to the formation of Milrad reaction products (MRPs) in cooked meat (53), such as heterocyclic amines and advanced glycation end products (AGEs). The MRPs reportedly increase oxidative stress and inflammation, both of which are potential factors in the development of HP (54, 55). Prospective studies conducted in Poland have shown a positive association between a proinflammatory diet and high-fat poultry (46). A proinflammatory diet is associated with higher dietary energy expenditure (47, 48), as fat cells store excess energy leading to adipose tissue hyperplasia and hypertrophy. This process triggers the release of high levels of proinflammatory molecules, including tumor necrosis factor receptors, interleukin-1 receptors, NF-kB transcription factors, and Toll-like receptors from macrophages (49). More importantly, the excess lipid is redistributed to other organs (e.g., blood vessels), causing the expression of proinflammatory mediators, differentiation of
monocytes into macrophages, and recruitment of macrophages. This can lead to a vicious cycle of impaired endothelial function and vascular inflammation, contributing to the development and progression of hypertension. In addition to poultry consumption, we found a negative correlation between dried fruit intake and the risk of HP. One clinical study supported the role of raisin intake in BP reduction. Dried fruit is a nutritionally concentrated form of fresh fruit with low water content. The health benefits of frequent dried fruit consumption include improvements in CVD and cardiometabolic syndrome (e.g., endothelial function, inflammation, lipid levels, and blood pressure) and insulin homeostasis. Although some estimates did not pass Bonferroni correction, their IVW derived P < 0.05 should also be treated with caution. Interestingly, we found that cooked vegetable intake can be a risk factor for CHD and MI, attributed to the effects of cooking, which can dramatically alter the chemical composition of vegetables and affect the concentration and bioavailability of bioactive compounds, such as heat-sensitive, antioxidant and water-soluble nutrients. It has been established that protein glycosylation occurs in food preparation processes such as baking, cooking and frying. High temperature and long cooking time are also conducive to the glycosylation reaction and formation of AGEs. By interacting with their corresponding receptor, AGEs can exacerbate oxidative stress by generating reactive oxygen species via NADPH oxidase in mitochondria. This, in turn, adversely impacts mitochondrial function and subsequently alters cellular metabolism, particularly in pathological circumstances. Therefore, AGEs are considered a risk factor for the onset of diet-related diseases such as cardiovascular disease.

Our two-sample MR Study, which assessed the association between genetic predisposition to 20 dietary habits and CVD, has several strengths. Firstly, implementing MR analysis allows for the mitigation of residual confounding and the minimization of issues related to reverse causality, frequently encountered in traditional observational studies. Secondly, a meticulous screening process was employed to carefully select robust genetic instruments, eliminate SNPs associated with outcomes, and mitigate potential confounding variables. Thirdly, the study employed three complementary MR methods to assess and address any potential violations of the MR assumption. This study offers a novel perspective to explore the causal relationship between dietary habits and the risk of CVD. Future studies should elucidate the mechanism underlying the association between diet and CVD or explore dietary treatment modalities to prevent or treat different types of CVD.
5. Limitation
The limitations inherent in MR studies affected the robustness of the present study findings to a certain extent. Firstly, the instrumental variable approach using genetic variation may introduce pleiotropy and residual confounding, where the genetic variants affect outcomes through pathways other than the exposure of interest. This could potentially lead to biased estimates. However, our study observed robust effect estimates across different MR models, and sensitivity analyses based on various assumptions did not indicate any horizontal pleiotropy. Additionally, the genetic variants used as instruments in our study may not completely capture the biological pathways through which dietary habits affect cardiovascular health. Secondly, the participants included in our study were predominantly of European descent, limiting the generalizability of the causal relationship between dietary habits and CVD to other populations. Thirdly, although the effect of diet on CVD may differ by sex and age group, we only analyzed the overall effect adjusted for sex and age due to limited GWAS summary data. Furthermore, certain dietary patterns, such as the Mediterranean diet, could not be examined due to insufficient data. Finally, some dietary habits were treated as ordered variables, and the assumption of a linear effect of SNPs on diet may introduce bias toward zero in GWAS results, potentially resulting in false-negative findings regarding the effect of dietary habits on CVD. We were also unable to confirm the presence of a nonlinear causal relationship between dietary habits and CVD. Linear regression-based GWAS is reportedly unsuitable for examining the nonlinear association between exposure and outcome. Despite these limitations, our MR results still serve as valid tests of the null hypothesis of causality, although some negative causal relationships in our study may have been underestimated. Accordingly, we recommend the prospective collection of detailed dietary data across different populations for future MR analyses or implementing randomized controlled trials with adequate sample sizes.

6. Conclusion
This is the first comprehensive MR analysis to reveal an association between dietary habits and CVD. Multiple dietary habits exhibit protective or harmful effects on CVD. Accordingly, adopting healthy dietary habits may prevent and reduce the risk of cardiovascular disease. However, further validation and investigation of the underlying mechanisms are warranted.
7. Acknowledgement
We want to acknowledge all participants and investigators from UK biobank, the FinnGen study, MEGASTROKE consortium, CARDIoGRAMplusC4D consortium, HERMES consortium and Roselli et al. for the AF GWAS meta-analysis.

J.P., M.M.Y. and X.G. designed the study, drafted the manuscript and verified the data; M.M.Y., X.G. conducted statistical analyses; L.Z.X. drafting the work and revising it critically for important intellectual content; Z.Z.L, X.S.Y. and J.Y.O. provided substantial contribution to interpretation of data; J.P. was accountable for all aspects of the work. All authors checked the paper critically and approved the final manuscript.

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The authors declare no conflict of interest.

9. Ethics
This study only used publicly summary data which has been approved to conduct human experimentation by an ethical standards committee. Hence, additional ethical approval was not required in current study.

10. References


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<table>
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<th>Exposure</th>
<th>Outcome</th>
<th>MR Egger intercept P value</th>
<th>Cochran’s Q-derived P value</th>
<th>MR method</th>
<th>OR</th>
<th>OR (95% Confidence intervals)</th>
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<td>CHD/MI/HP</td>
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<td>Weighted median</td>
<td></td>
<td></td>
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<td></td>
<td></td>
</tr>
<tr>
<td>Pork intake</td>
<td>0.72</td>
<td>0.128</td>
<td>IVW 1.992</td>
<td>1.008-3.936</td>
<td>0.047</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td>MR-Egger 4.638</td>
<td>0.048-4.879</td>
<td>0.524</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td>Weighted median</td>
<td>2.27</td>
<td>1.024-5.032</td>
<td>0.043</td>
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<tr>
<td>Cooked vegetable intake</td>
<td>0.911</td>
<td>0.380</td>
<td>IVW 2.947</td>
<td>1.354-6.415</td>
<td>6.464×10⁻³</td>
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<tr>
<td></td>
<td></td>
<td></td>
<td>MR-Egger 5.937</td>
<td>5.154×10⁻³-6.839×10⁻⁵</td>
<td>0.777</td>
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<td></td>
<td></td>
<td></td>
<td>Weighted median</td>
<td>2.762</td>
<td>0.921-8.279</td>
<td>0.070</td>
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<tr>
<td>Coffee intake</td>
<td>0.862</td>
<td>0.004</td>
<td>IVW 1.545</td>
<td>1.107-2.155</td>
<td>0.010</td>
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<td></td>
<td>MR-Egger 1.463</td>
<td>0.726-2.948</td>
<td>0.295</td>
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<td>Weighted median</td>
<td>1.404</td>
<td>0.924-2.135</td>
<td>0.101</td>
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<tr>
<td>Alcohol intake frequency</td>
<td>0.647</td>
<td>0.001</td>
<td>IVW 1.256</td>
<td>1.092-1.445</td>
<td>0.001</td>
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<td></td>
<td>MR-Egger 1.138</td>
<td>0.729-1.775</td>
<td>0.571</td>
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<td>Weighted median</td>
<td>1.203</td>
<td>1.042-1.388</td>
<td>0.013</td>
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**Significant estimate is defined as IVW-derived P < 4.17×10⁻⁴, *nominal significant estimate is defined as IVW-derived P < 0.05. Cochran’s Q-derived P value and MR-Egger intercept-derived P value < 0.05 is significant.**

MR, Mendelian randomization; IVW, Inverse-variance weighted; MI: myocardial infarction; HF, heart failure; HP, hypertension; CHD, coronary heart disease.
**Figure 1:** The central framework of Mendelian randomization analyses of the causal relationship of 20 dietary habits on the risk of cardiovascular disease. Assumption 1: IVs are strongly associated with the exposure; Assumption 2: IVs have an impact on outcome through the exposure only; Assumption 3: IVs are independent of other factors that influence outcome. SNPs=single nucleotide polymorphism; CHD: coronary heart disease, HP: hypertension, HF: heart failure, MI: myocardial infarction, AF: atrial fibrillation, IS: ischemic stroke.
Figure 2: Scatter plots of significant estimates from genetically predicted dietary habits on CVD. (a) cheese intake on MI; (b) cheese intake on HF; (c) poultry intake on HP; (d) dried fruit intake on HP.
Figure 3: Leave-one-out plots of significant estimates from genetically predicted dietary habits on CVD. (a) cheese intake on MI; (b) cheese intake on HF; (c) poultry intake on HP; (d) dried fruit intake on HP.