

Association between metabolic syndrome and inflammatory bowel disease: a bidirectional two-sample Mendelian randomized study

Keywords

Metabolic Syndrome, Inflammatory Bowel Disease, Ulcerative Colitis, Crohn's Disease

Abstract

Introduction

Epidemiological studies have revealed parallel increases in the incidences of metabolic syndrome (MetS) and inflammatory bowel disease (IBD). Clinical observational studies have shown a link between MetS and a poor prognosis of IBD. However, the causal relationship between MetS and IBD remains unclear. This study used bidirectional two-sample Mendelian randomisation to investigate potential causal links between MetS and IBD, including ulcerative colitis (UC) and Crohn's disease (CD).

Material and methods

Genetic associations of MetS and components, IBD were sourced from public databases of European populations. Inverse variance weighting was conducted, with weighted median, Mendelian randomisation–Egger, and Mendelian randomization (MR) Pleiotropy Residual Sum and Outlier methods used as sensitivity analyses. This process was repeated in the opposite direction.

Results

The Inverse variance weighted (IVW) method showed that genetic prediction of MetS may be a potential risk factor for CD (OR=1.34, 95% CI: 1.009–1.779; P=0.043). In further estimating the different components of MetS, waist circumference may increase the risk of CD (OR=1.33, 95% CI: 1.05–1.684; P=0.018) and hypertension may increase the risk of UC (OR=1.61, 95% CI: 1.084–2.39; P=0.018). In reverse analysis, IBD may increased triglyceride levels (OR=1.019, 95% CI: 1.000–1.038; P=0.049).

Conclusions

This MR Analysis showed a causal relationship between genetically predicted MetS and CD, and genetically predicted hypertension and UC. Therefore, these patients need to be closely monitored clinically for the risk of CD/UC comorbidities. In patients with IBD, close monitoring of MetS-associated cardiovascular risk is required.

Association between metabolic syndrome and inflammatory bowel disease: a bidirectional two-sample Mendelian randomized study

running head: metabolic syndrome and inflammatory bowel disease

Abstract

Introduction: Epidemiological studies have revealed parallel increases in the incidences of metabolic syndrome (MetS) and inflammatory bowel disease (IBD). Clinical observational studies have shown a link between MetS and a poor prognosis of IBD. However, the causal relationship between MetS and IBD remains unclear. This study used bidirectional two-sample Mendelian randomisation to investigate potential causal links between MetS and IBD, including ulcerative colitis (UC) and Crohn's disease (CD).

Material and methods: Genetic associations of MetS and components, IBD were sourced from public databases of European populations. Inverse variance weighting was conducted, with weighted median, Mendelian randomisation–Egger, and **Mendelian randomization (MR)** Pleiotropy Residual Sum and Outlier methods used as sensitivity analyses. This process was repeated in the opposite direction.

Results: The **Inverse variance weighted (IVW)** method showed that genetic prediction of MetS may be a potential risk factor for CD (OR=1.34, 95% CI: 1.009–1.779; $P=0.043$). In further estimating the different components of MetS, waist circumference may increase the risk of CD (OR=1.33, 95% CI: 1.05–1.684; $P=0.018$)

and hypertension may increase the risk of UC (OR=1.61, 95% CI: 1.084–2.39; $P=0.018$). In reverse analysis, IBD may increased triglyceride levels (OR=1.019, 95% CI: 1.000–1.038; $P=0.049$).

Conclusions: This MR Analysis showed a causal relationship between genetically predicted MetS and CD, and genetically predicted hypertension and UC. Therefore, these patients need to be closely monitored clinically for the risk of CD/UC comorbidities. In patients with IBD, close monitoring of MetS-associated cardiovascular risk is required.

Keywords:

Metabolic Syndrome, Inflammatory Bowel Disease, Ulcerative Colitis, Crohn's Disease

Introduction

Inflammatory bowel disease (IBD) is a chronic, immune-mediated inflammatory disease of the intestine. Ulcerative colitis (UC) and Crohn's disease (CD) are the two main types of IBD. The pathogenesis of IBD remains unknown, but it involves complex interactions among genetic, environmental, microbial, and immune factors. The incidence of IBD is increasing globally [1]. Metabolic syndrome (MetS) is a group of complex metabolic disorders that includes obesity, dyslipidaemia, hypertension, and insulin resistance; the syndrome has a global incidence of

approximately 12%–31% [2,3]. The clinical features of MetS include elevated diastolic or systolic blood pressure, increased fasting blood glucose and triglyceride levels, increased waist circumference, and decreased levels of high-density lipoprotein cholesterol [4].

Epidemiological studies have revealed similar upward trends in the incidences of IBD and MetS in recent decades, suggesting a common environmental link between these two diseases. Both diseases share clinically relevant features, such as an increased risk of cardiovascular disease [5,6] and increased incidences of non-alcoholic cirrhosis [7,8] and obesity [9,10]. **MetS is a common comorbidity of IBD, and their co-occurrence is increasing in incidence. MetS and IBD have some several similar pathophysiological features,** including immune imbalance, chronic inflammation, adipose tissue dysfunction, and disorders of the gut microbiota [11]. Although studies have suggested an association between MetS and a poor prognosis of IBD [12,13], previous studies on the relationship between IBD and MetS have largely been limited to observational or single-centre studies with small sample sizes. Consequently, the causal relationship between IBD and MetS remains unclear [14].

Mendelian randomization (MR) is a genetics-based research method used to assess the causal effects of exposure factors on outcomes. It employs genetic variations associated with these factors as instrumental variables. The core concept of this method is that genetic variation in the population is randomly distributed, similar to the randomisation employed in randomised controlled trials; this effectively

controls the influence of confounding factors [15]. Therefore, this study aimed to use MR to explore the causal relationship between MetS and IBD based on the latest summary statistics of genome-wide association studies (GWASs), providing new insights into the prevention and treatment of IBD.

Material and methods

The overall study design of this bidirectional two-sample MR analyses is shown in Figure 1. To be used as instrumental variables, single nucleotide polymorphisms (SNPs) were required to meet three assumptions: (1) they are associated with the exposure, (2) they are independent of any confounding factors in the exposure–outcome relationship, and (3) they affect the outcome solely through the exposure [16]. The detailed summary data used in the present study are shown in Table 1. This study was conducted in accordance with the Strengthening the Reporting of Observational Studies in Epidemiology Using Mendelian Randomization reporting guidelines [17]. All data used in this study are derived from published public databases; therefore, no additional ethical approval was required.

Source of GWAS data

MetS GWAS

GWAS data for MetS were obtained from the Center for Neurogenomics and Cognitive Research database, including data from a study by Van Walree et al [18]. The Van Walree et al. study is, to date, the largest GWAS study to focus on MetS and includes data from 461,920 individuals of European ancestry. The GWAS summary

data of the five components of MetS (waist circumference (WC), high blood pressure, fasting blood glucose (FBG), high-density lipoprotein cholesterol (HDL-C) and triglyceride (TG) were obtained from IEU Open GWAS database (<https://gwas.mrcieu.ac.uk/>).

IBD GWAS

GWAS data regarding IBD and its subtypes, UC and CD were obtained from the latest FinnGen R12 dataset [19], which includes 10,960 cases and 489,388 controls for IBD, 7,220 cases and 492,160 controls for UC, and 2,489 cases and 497,622 controls for CD.

Instrument selection

Strict selection criteria and linkage disequilibrium clumping were used to identify suitable instrumental variables for the MR analyses. SNPs with a genome-wide significance level of $P < 5 \times 10^{-8}$ were included. Further, we performed a linkage disequilibrium clumping and excluded SNPs with an r^2 value of ≥ 0.001 and a clump distance of $\leq 10,000$ kb to eliminate SNPs that correlated more strongly with outcomes than with exposure [20]. The F statistic was calculated separately for each SNP. Weak instrumental variables were defined as those with an F statistic of < 10 and all weak instrumental variables were excluded from the analyses [21].

Statistical analyses

A generalised **inverse variance weighted (IVW)** MR approach was used for the principal analysis. MR analysis was conducted for each of the three European

databases, and the overall effect of each specific outcome was assessed using a meta-analysis. Cochrane's Q was used to calculate the I^2 statistics to assess the heterogeneity of the SNP estimates. A random effects model was used when significant heterogeneity was detected ($P < 0.05$); otherwise, a fixed effects model was used. Several complementary methods were applied to provide reliable and consistent causal estimates, including the weighted median [22], MR-Egger [23], and Mendelian Randomization Pleiotropy RESidual Sum and Outlier (MR-PRESSO) [24] methods. The P -value of the MR-Egger method intercept was used to evaluate the horizontal pleiotropy, with $P < 0.05$ indicating the presence of horizontal pleiotropy. An MR-PRESSO analysis was performed to identify and eliminate outliers, and to evaluate whether a significant difference in the causal effect could be observed after these outliers were removed ($P < 0.05$). The leave-one-out method was used to determine whether the overall causal effect was influenced by any single SNP, which could potentially introduce bias. Multiple tests were performed using the Benjamini-Hochberg correction to control the false discovery rate; correlations with $P < 0.05$ were considered significant.

All statistical analyses were performed using the R packages MR-PRESSO and TwoSampleMR within the open-source statistical software R (version 4.4.0; R Foundation for Statistical Computing, Vienna, Austria).

Results

Causal role of MetS in IBD, UC, and CD

Our results suggested that MetS could increase the risk of CD (OR=1.34, 95% CI: 1.009–1.779; $P=0.043$) with low heterogeneity. Genetically predicted MetS was also not associated with IBD and UC (Figure 2). In further analysis, we found a causal relationship between waist circumference and CD (OR=1.33, 95% CI: 1.05–1.684; $P=0.018$) and a causal relationship between hypertension and UC (OR=1.61, 95% CI: 1.084–2.39; $P=0.018$) in the MetS component with low heterogeneity (Figure 2). Our study found no causal relationship between genetically predicted FBG, HDL-C, triglycerides and IBD, UC, and CD. The scatter plots for the forward analyses and the leave-one-out analyses for each SNP association are summarised in Supplementary Figure S1 and Figure S2, respectively. Detailed information regarding the instrumental variables for MetS and components is provided in Supplementary Tables. Sensitivity analyses, including the weighted median, MR-Egger, and MR-PRESSO methods, yielded consistent findings (Table 2). The statistical results between FBG and UC based on the MR-Egger intercept show horizontal pleiotropy, while other statistical results do not show horizontal pleiotropy (Table 2).

Causal role of IBD, UC, and CD in MetS

In reverse analysis, our findings showed that genetically predicted IBD, UC, and CD were not associated with MetS (Figure 3). In addition, we found a causal relationship between IBD and triglycerides (OR=1.019, 95% CI: 1.000–1.038; $P=0.049$) with no heterogeneity (Figure 3). Our results suggest no causal relationship between genetically predicted IBD, UC, and CD and FBG, WC, HDL-C, hypertension.

No horizontal pleiotropy was observed for all outcomes. The sensitivity analysis revealed similar findings (Table 3). Scatter plots for the reverse analyses and the plots of the leave-one-out analyses for each SNP are summarised in Supplementary Figure S3 and Supplementary Figure S4, respectively. Detailed information regarding the instrumental variables for IBD, UC, and CD is provided in Supplementary Tables.

Discussion

This is the first study to comprehensively examine the causal relationship between MetS and IBD, including the IBD subtypes UC and CD. After rigorous reverse variance weighted analysis and sensitivity analysis, our results revealed a significant association between MetS, WC, hypertension, triglycerides, and IBD.

The comorbidities of IBD must be considered during treatment as they can alter disease activity and parenteral manifestations, ultimately affecting the disease prognosis and drug treatment responses. The global incidences of MetS and IBD have increased in tandem, and approximately 19.4% of patients with IBD also have MetS [25]. As a comorbidity of IBD, MetS increases the risk of cardiovascular disease, liver disease, and surgical complications and reduces patients' quality of life [12,26-28]. Obesity, a characteristic of MetS, may increase the incidence and severity of CD and the risk of cancer, and affect the patient's response to treatment, although MetS does not have the same impact in patients with UC [9,29-31]. Previous studies on the effects of MetS on IBD were observational, rendering them susceptible to reverse causality and other biases. The causal relationship between MetS and IBD remains

unclear, as some studies have reported conflicting results [32,33]. Through the use of different estimation models and rigorous sensitivity analyses, MR effectively reduces potential biases such as confounding and reverse causality, enhancing the causal reasoning and ensuring the reliability and robustness of the study findings. The results of this study suggest that MetS may increase the risk of CD. In a further analysis of MetS components, increased WC appeared to have a more significant effect on risk of CD. In addition, we observed that hypertension may increase the risk of UC, and IBD may lead to elevated triglyceride levels.

Although there is limited research on the causal relationship between MetS and IBD, there have been several studies exploring whether obesity, a core component of MetS, has an impact on the development of IBD, but have provided inconsistent and conflicting evidence [34-41]. While most cohort studies have proposed that general obesity, as represented by body mass index, increases the risk of CD and decreases the risk of UC [36,39-41], a few studies have shown inconsistent findings [36,38,40]. As for abdominal obesity as measured by WC, limited cohort study evidence suggests a positive association with the risk of CD [35,41]. WC usually reflects abdominal adipose tissue, including visceral adipose tissue (VAT) and subcutaneous adipose tissue (SAT). The VAT, adipocyte dysfunction, chronic low-grade inflammation, and insulin resistance are components of MetS [42], and VAT plays a central role in the pathophysiology of MetS. Therefore, VAT may contribute to chronic systemic inflammation in patients with MetS or IBD [11]. As an important endocrine organ that

integrates and body's energy homeostasis, the adipose tissue serves as an important metabolic regulator by secreting adipokines with pro-inflammatory and anti-inflammatory activities [43]. In a normal metabolic state, the balance between pro-inflammatory and anti-inflammatory adipokines maintains homeostasis; however, excessive calorie intake can cause fat cells to become hypertrophic, leading to central obesity. If this state persists and exceeds the buffering capacity of the fat cells, the cells are subjected to oxidative stress, resulting in disruption and the adipose tissue producing abnormal levels of resistin, leptin, and adiponectin [44,45]. Hypertrophic adipocytes secrete interleukin-6, tumour necrosis factor-alpha, and monocyte chemoattractant protein-1, which recruit monocytes and promote their differentiation into pro-inflammatory macrophages. These macrophages infiltrate the VAT and promote chronic, low-grade inflammation throughout the body [46]. Unlike subcutaneous adipose tissue, VAT actively promotes local systemic inflammation [47]. Individuals with obesity and VAT are more likely to develop MetS and IBD than individuals with SAT [48]. In patients with CD, VAT will cover the intestinal surface to form "creeping fat", creeping fat cells have inflammatory characteristics, and the expression of cytokines and adipokines involved in inflammation is increased [49]. As an important indicator of disease activity, creeping fat is found in 100% of patients with CD, whereas it is generally absent in UC [50]. In addition, compared with UC patients, the visceral adipose tissue of CD patients is more prone to inflammation and colonization by intestinal bacteria [51]. Therefore, MetS and WC may be more

associated with the risk of CD than UC.

Hypertension is an important component of MetS, and hypertension and IBD share some common core pathways in pathogenesis. Pro-inflammatory signaling molecules, including interleukin-1 β , tumor necrosis factor- α , and interleukin-6, are significantly elevated in both diseases. These molecules coordinate chronic inflammation, endothelial dysfunction, and smooth muscle cell proliferation, which leads to plaque formation and vascular damage. In essence, the persistence of systemic inflammation triggered by these cytokines is the common driving force behind the development and progression of IBD and cardiovascular disease. The endothelium is an important regulator of vascular function and plays a key role in maintaining cardiovascular health. In healthy conditions, endothelium promotes vasodilation, inhibits thrombosis, and regulates inflammation. Endothelial dysfunction disrupts this delicate balance, resulting in impaired vasoconstriction and vasodilation [52]. A previous cohort study suggested that UC patients have a higher cumulative risk of developing hypertension than the general population [53]. Another meta-analysis found that patients with IBD had a higher risk of co-existing hypertension [54]. However, the evidence for a causal relationship between hypertension and IBD remains limited. This study found for the first time that hypertension may increase the risk of UC through MR Analysis, but its mechanism still needs further study. In addition, previous studies have shown elevated triglyceride levels in IBD patients [55,56], triglycerides play an important role in

atherosclerosis [57], similarly, the results of this study indicate that IBD increases triglyceride levels, suggesting that IBD may increase the risk of cardiovascular disease.

Although this study used MR to effectively control for confounding factors and inverse causality, some limitations remain. First, MR Analyses infer causal hypotheses by randomly assigning genetic variants, so it is difficult to fully distinguish between pleiotropy and mediations during the analysis. Second, the genetic data used in this study were primarily derived from European populations and may limit the applicability of the findings to other ethnicities and regions. Finally, larger sample sizes and more advanced methodologies are needed to confirm these findings and enhance the statistical power.

Conclusions

This MR Analysis showed a causal relationship between genetically predicted MetS and CD, and genetically predicted hypertension and UC. Therefore, these patients need to be closely monitored clinically for the risk of CD/UC comorbidities. In patients with IBD, close monitoring of triglycerides-associated cardiovascular risk is required.

Ethics approval

Informed consent and ethical approval are not required for Mendelian randomization studies.

Availability of data and materials

Corresponding authors can be contacted to request access to the datasets used or analyzed in this study.

Competing interests

The authors declare that they have no competing interests.

Funding

This study was supported by the Natural Science Foundation of Guangdong Province, China (Grant No: 2022A1515011230) and the President Foundation of Nanfang Hospital, Southern Medical University (2023B050).

Acknowledgements

The authors thank for access to the GWAS data of MetS by van Walree et al. We also thank the FinnGen consortium and UK biobank for sharing the summary-level GWAS data.

References

1. Buie MJ, Quan J, Windsor JW, Coward S, Hansen TM, King JA, et al. Global Hospitalization Trends for Crohn's Disease and Ulcerative Colitis in the 21st Century: A Systematic Review With Temporal Analyses. Clin Gastroenterol Hepatol. 2023 Aug;21(9):2211-2221.
2. Noubiap JJ, Nansseu JR, Lontchi-Yimagou E, Nkeck JR, Nyaga UF, Ngouo AT, et al. Geographic distribution of metabolic syndrome and its components in the general adult population: A meta-analysis of global data from 28 million individuals. Diabetes Res Clin Pract. 2022 Jun;188:109924.

3. Dobrowolski P, Prejbisz A, Kuryłowicz A, Baska A, Burchardt P, Chlebus K, et al. Metabolic syndrome - a new definition and management guidelines: A joint position paper by the Polish Society of Hypertension, Polish Society for the Treatment of Obesity, Polish Lipid Association, Polish Association for Study of Liver, Polish Society of Family Medicine, Polish Society of Lifestyle Medicine, Division of Prevention and Epidemiology Polish Cardiac Society, "Club 30" Polish Cardiac Society, and Division of Metabolic and Bariatric Surgery Society of Polish Surgeons. Arch Med Sci. 2022;18(5):1133-1156.
4. Huang PL. A comprehensive definition for metabolic syndrome. Dis Model Mech. 2009 May-Jun;2(5-6):231-7.
5. Tune JD, Goodwill AG, Sassoon DJ, Mather KJ. Cardiovascular consequences of metabolic syndrome. Transl Res. 2017 May;183:57-70.
6. Cheng K, Faye AS. Venous thromboembolism in inflammatory bowel disease. World J Gastroenterol. 2020 Mar 28;26(12):1231-1241.
7. Principi M, Iannone A, Losurdo G, Mangia M, Shahini E, Albano F, et al. Nonalcoholic Fatty Liver Disease in Inflammatory Bowel Disease: Prevalence and Risk Factors. Inflamm Bowel Dis. 2018 Jun 8;24(7):1589-1596.
8. Yki-Järvinen H. Non-alcoholic fatty liver disease as a cause and a consequence of metabolic syndrome. Lancet Diabetes Endocrinol. 2014 Nov;2(11):901-10.
9. Chan SSM, Chen Y, Casey K, Olen O, Ludvigsson JF, Carbonnel F, et al. Obesity is Associated With Increased Risk of Crohn's disease, but not Ulcerative Colitis: A

Pooled Analysis of Five Prospective Cohort Studies. Clin Gastroenterol Hepatol. 2022 May;20(5):1048-1058.

10. Bacha RA, Bouhnik Y, Serrero M, Filippi J, Roblin X, Bourrier A, et al. Obesity in adult patients with inflammatory bowel disease: Clinical features and impact on disability. A cross-sectional survey from the GETAID. Dig Liver Dis. 2023 Dec;55(12):1632-1639.

11. Michalak A, Mosińska P, Fichna J. Common links between metabolic syndrome and inflammatory bowel disease: Current overview and future perspectives. Pharmacol Rep. 2016 Aug;68(4):837-46.

12. Khakoo NS, Ioannou S, Khakoo NS, Vedantam S, Pearlman M. Impact of Obesity on Inflammatory Bowel Disease. Curr Gastroenterol Rep. 2022 Jan;24(1):26-36.

13. Argollo M, Gilardi D, Peyrin-Biroulet C, Chabot JF, Peyrin-Biroulet L, Danese S. Comorbidities in inflammatory bowel disease: a call for action. Lancet Gastroenterol Hepatol. 2019 Aug;4(8):643-654.

14. Adolph TE, Meyer M, Jukic A, Tilg H. Heavy arch: from inflammatory bowel diseases to metabolic disorders. Gut. 2024 Jul 11;73(8):1376-1387.

15. Doyle SL, Donohoe CL, Lysaght J, Reynolds JV. Visceral obesity, metabolic syndrome, insulin resistance and cancer. Proc Nutr Soc. 2012 Feb;71(1):181-9.

16. Burgess S, Davey Smith G, Davies NM, Dudbridge F, Gill D, Glymour MM, et al. Guidelines for performing Mendelian randomization investigations: update for summer 2023. Wellcome Open Res. 2023 Aug 4;4:186.

17. Skrivankova VW, Richmond RC, Woolf BAR, Davies NM, Swanson SA, VanderWeele TJ, et al. Strengthening the reporting of observational studies in epidemiology using mendelian randomisation (STROBE-MR): explanation and elaboration. *BMJ*. 2021 Oct 26;375:n2233.
18. van Walree ES, Jansen IE, Bell NY, Savage JE, de Leeuw C, Nieuwdorp M, van der Sluis S, Posthuma D. Disentangling Genetic Risks for Metabolic Syndrome. *Diabetes*. 2022 Nov 1;71(11):2447-2457.
19. Kurki MI, Karjalainen J, Palta P, Sipilä TP, Kristiansson K, Donner KM, et al. FinnGen provides genetic insights from a well-phenotyped isolated population. *Nature*. 2023 Jan;613(7944):508-518.
20. Hemani G, Tilling K, Davey Smith G. Orienting the causal relationship between imprecisely measured traits using GWAS summary data. *PLoS Genet*. 2017 Nov 17;13(11):e1007081.
21. Davies NM, Holmes MV, Davey Smith G. Reading Mendelian randomisation studies: a guide, glossary, and checklist for clinicians. *BMJ*. 2018 Jul 12;362:k601.
22. Bowden J, Davey Smith G, Haycock PC, Burgess S. Consistent Estimation in Mendelian Randomization with Some Invalid Instruments Using a Weighted Median Estimator. *Genet Epidemiol*. 2016 May;40(4):304-14.
23. Bowden J, Davey Smith G, Burgess S. Mendelian randomization with invalid instruments: effect estimation and bias detection through Egger regression. *Int J Epidemiol*. 2015 Apr;44(2):512-25.

24. Verbanck M, Chen CY, Neale B, Do R. Detection of widespread horizontal pleiotropy in causal relationships inferred from Mendelian randomization between complex traits and diseases. *Nat Genet.* 2018 May;50(5):693-698.
25. Shen Z, Zhang M, Liu Y, Ge C, Lu Y, Shen H, et al. Prevalence of metabolic syndrome in patients with inflammatory bowel disease: a systematic review and meta-analysis. *BMJ Open.* 2024 Mar 7;14(3):e074659.
26. Tune JD, Goodwill AG, Sassoon DJ, Mather KJ. Cardiovascular consequences of metabolic syndrome. *Transl Res.* 2017 May;183:57-70.
27. Méndez-Sánchez N, Bugianesi E, Gish RG, Lammert F, Tilg H, Nguyen MH, et al. Global multi-stakeholder endorsement of the MAFLD definition. *Lancet Gastroenterol Hepatol.* 2022 May;7(5):388-390.
28. Jiang K, Chen B, Lou D, Zhang M, Shi Y, Dai W, et al. Systematic review and meta-analysis: association between obesity/overweight and surgical complications in IBD. *Int J Colorectal Dis.* 2022 Jul;37(7):1485-1496.
29. Sehgal P, Su S, Zech J, Nobel Y, Luk L, Economou I, et al. Visceral Adiposity Independently Predicts Time to Flare in Inflammatory Bowel Disease but Body Mass Index Does Not. *Inflamm Bowel Dis.* 2024 Apr 3;30(4):594-601.
30. Yarur AJ, Bruss A, Moosreiner A, Beniwal-Patel P, Nunez L, Berens B, et al. Higher Intra-Abdominal Visceral Adipose Tissue Mass Is Associated With Lower Rates of Clinical and Endoscopic Remission in Patients With Inflammatory Bowel Diseases Initiating Biologic Therapy: Results of the Constellation Study.

Gastroenterology. 2023 Oct;165(4):963-975.e5.

31. Wunderlich CM, Ackermann PJ, Ostermann AL, Adams-Quack P, Vogt MC, Tran ML, et al. Obesity exacerbates colitis-associated cancer via IL-6-regulated macrophage polarisation and CCL-20/CCR-6-mediated lymphocyte recruitment. *Nat Commun.* 2018 Apr 25;9(1):1646.

32. Nagahori M, Hyun SB, Totsuka T, Okamoto R, Kuwahara E, Takebayashi T, et al. Prevalence of metabolic syndrome is comparable between inflammatory bowel disease patients and the general population. *J Gastroenterol.* 2010 Oct;45(10):1008-13.

33. Jovanovic M, Simovic Markovic B, Gajovic N, Jurisevic M, Djukic A, Jovanovic I, et al. Metabolic syndrome attenuates ulcerative colitis: Correlation with interleukin-10 and galectin-3 expression. *World J Gastroenterol.* 2019 Nov 21;25(43):6465-6482.

34. Milajerdi A, Abbasi F, Esmailzadeh A. A systematic review and meta-analysis of prospective studies on obesity and risk of inflammatory bowel disease. *Nutr Rev.* 2022 Feb 10;80(3):479-487.

35. Je Y, Han K, Chun J, Kim Y, Kim JH, Hoon Youn Y, et al. Association of Waist Circumference with the Risk of Inflammatory Bowel Disease: a Nationwide Cohort Study of 10 Million Individuals in Korea. *J Crohns Colitis.* 2023 May 3;17(5):681-692.

36. Harpsøe MC, Basit S, Andersson M, Nielsen NM, Frisch M, Wohlfahrt J, et al.

Body mass index and risk of autoimmune diseases: a study within the Danish National Birth Cohort. *Int J Epidemiol*. 2014 Jun;43(3):843-55.

37. Rahmani J, Kord-Varkaneh H, Hekmatdoost A, Thompson J, Clark C, Salehisahlabadi A, et al. Body mass index and risk of inflammatory bowel disease: A systematic review and dose-response meta-analysis of cohort studies of over a million participants. *Obes Rev*. 2019 Sep;20(9):1312-1320.

38. Chan SS, Luben R, Olsen A, Tjønneland A, Kaaks R, Teucher B, et al. Body mass index and the risk for Crohn's disease and ulcerative colitis: data from a European Prospective Cohort Study (The IBD in EPIC Study). *Am J Gastroenterol*. 2013 Apr;108(4):575-82.

39. Jensen CB, Ängquist LH, Mendall MA, Sørensen TIA, Baker JL, Jess T. Childhood body mass index and risk of inflammatory bowel disease in adulthood: a population-based cohort study. *Am J Gastroenterol*. 2018 May;113(5):694-701.

40. Khalili H, Ananthakrishnan AN, Konijeti GG, Higuchi LM, Fuchs CS, Richter JM, et al. Measures of obesity and risk of Crohn's disease and ulcerative colitis. *Inflamm Bowel Dis*. 2015 Feb;21(2):361-8.

41. He Z, Fu T, Lu S, Sun Y, Zhang Y, Shi W, et al. Adiposity as a risk factor for inflammatory bowel disease and the mediating effect of metabolic and inflammatory status: A population-based cohort study. *United European Gastroenterol J*. 2023 Dec;11(10):973-984.

42. Zafar U, Khaliq S, Ahmad HU, Manzoor S, Lone KP. Metabolic syndrome: an

update on diagnostic criteria, pathogenesis, and genetic links. *Hormones (Athens)*. 2018 Sep;17(3):299-313.

43. Choe SS, Huh JY, Hwang IJ, Kim JI, Kim JB. Adipose Tissue Remodeling: Its Role in Energy Metabolism and Metabolic Disorders. *Front Endocrinol (Lausanne)*. 2016 Apr 13;7:30.

44. Leal Vde O, Mafra D. Adipokines in obesity. *Clin Chim Acta*. 2013 Apr 18;419:87-94.

45. Kahn CR, Wang G, Lee KY. Altered adipose tissue and adipocyte function in the pathogenesis of metabolic syndrome. *J Clin Invest*. 2019 Oct 1;129(10):3990-4000.

46. Chung S, Cuffe H, Marshall SM, McDaniel AL, Ha JH, Kavanagh K, et al. Dietary cholesterol promotes adipocyte hypertrophy and adipose tissue inflammation in visceral, but not in subcutaneous, fat in monkeys. *Arterioscler Thromb Vasc Biol*. 2014 Sep;34(9):1880-7.

47. Tchernof A, Després JP. Pathophysiology of human visceral obesity: an update. *Physiol Rev*. 2013 Jan;93(1):359-404.

48. Hamdy O, Porramatikul S, Al-Ozairi E. Metabolic obesity: the paradox between visceral and subcutaneous fat. *Curr Diabetes Rev*. 2006 Nov;2(4):367-73.

49. Gonçalves P, Magro F, Martel F. Metabolic inflammation in inflammatory bowel disease: crosstalk between adipose tissue and bowel. *Inflamm Bowel Dis*. 2015 Feb;21(2):453-67.

50. Xiong Z, Wu P, Zhang Y, Chen J, Shen Y, Kamel I, et al. Radiological biomarkers

reflecting visceral fat distribution help distinguish inflammatory bowel disease subtypes: a multicenter cross-sectional study. *Insights Imaging*. 2024 Mar 13;15(1):70.

51. Zulian A, Canello R, Ruocco C, Gentilini D, Di Blasio AM, Danelli P, et al. Differences in visceral fat and fat bacterial colonization between ulcerative colitis and Crohn's disease. An in vivo and in vitro study. *PLoS One*. 2013 Oct 24;8(10):e78495.

52. Sinha T, Zain Z, Bokhari SFH, Waheed S, Reza T, Eze-Odurukwe A, et al. Navigating the Gut-Cardiac Axis: Understanding Cardiovascular Complications in Inflammatory Bowel Disease. *Cureus*. 2024 Feb 29;16(2):e55268.

53. He J, Zhang S, Qiu Y, Liu F, Liu Z, Tan J, et al. Ulcerative colitis increases risk of hypertension in a UK biobank cohort study. *United European Gastroenterol J*. 2023 Feb;11(1):19-30.

54. Jaiswal V, Batra N, Dagar M, Butey S, Huang H, Chia JE, et al. Inflammatory bowel disease and associated cardiovascular disease outcomes: A systematic review. *Medicine (Baltimore)*. 2023 Feb 10;102(6):e32775.

55. Koutroumpakis E, Ramos-Rivers C, Regueiro M, Hashash JG, Barrie A, Swoger J, et al. Association Between Long-Term Lipid Profiles and Disease Severity in a Large Cohort of Patients with Inflammatory Bowel Disease. *Dig Dis Sci*. 2016 Mar;61(3):865-71.

56. Chen H, Li W, Hu J, Xu F, Lu Y, Zhu L, et al. Association of serum lipids with inflammatory bowel disease: a systematic review and meta-analysis. *Front Med*

(Lausanne). 2023 Aug 24;10:1198988.

57. Toth PP. Triglycerides and Cardiovascular Risk: Getting to the Heart of the Matter.

J Am Coll Cardiol. 2024 Sep 10;84(11):1007-1009.

Preprint

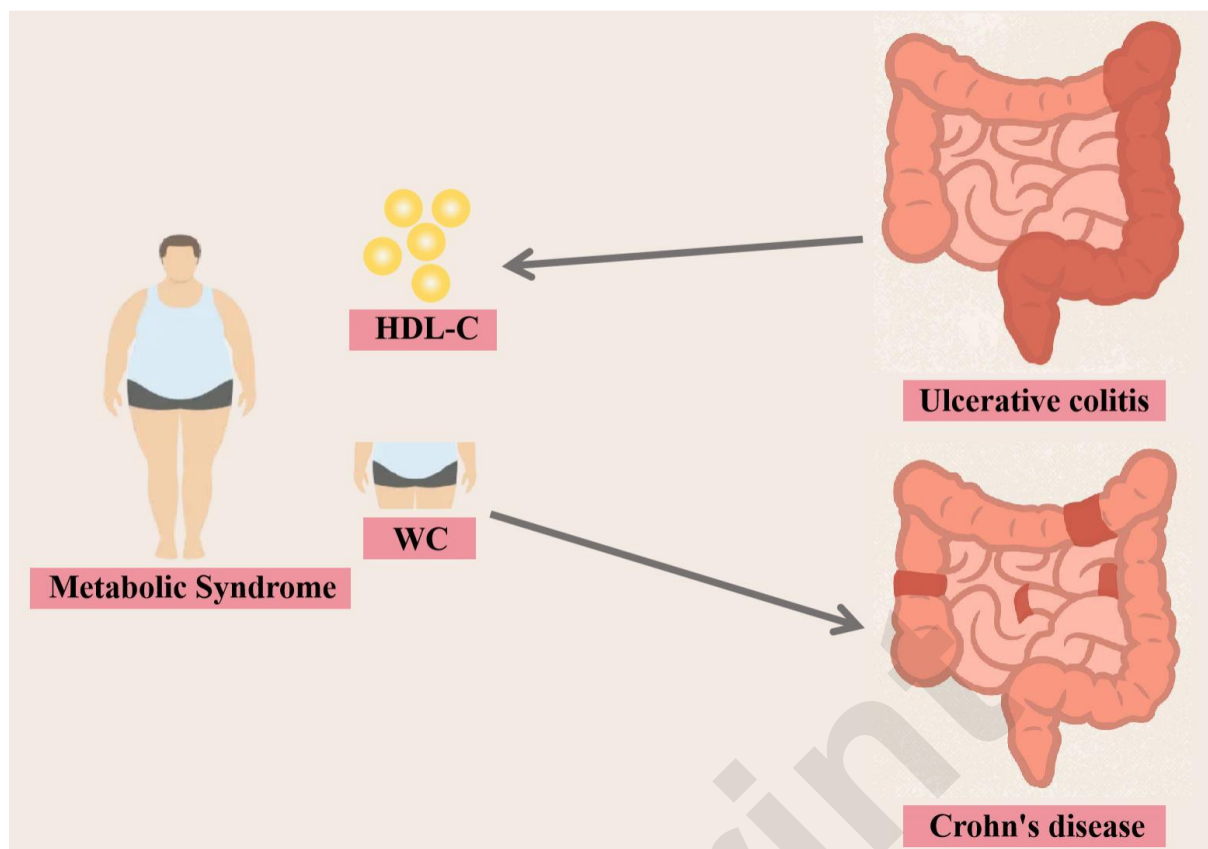
Figure Legends

Figure 1 The overall design of the present Mendelian randomization analysis.

Figure 2 Genetically predicted association between MetS and IBD by forward MR Analysis. **Abbreviations:** IBD, Inflammatory bowel disease; UC, Ulcerative colitis; CD, Crohn's disease; FBG, Fasting blood glucose; WC, Waist circumference.

Figure 3 Genetically predicted association between IBD and MetS by reverse MR Analysis. **Abbreviations:** IBD, Inflammatory bowel disease; UC, Ulcerative colitis; CD, Crohn's disease; FBG, Fasting blood glucose; WC, Waist circumference.

Preprint



Tables

Table 1 Characteristics of the used genome-wide association study in the study

Phenotypes	Ancestry	Sample size	Data sources
MetS	European	461,920	van Walree et al
FBG	European	58,074	https://gwas.mrcieu.ac.uk/datasets/ebi-a-GCST005186/
WC	European	462,166	https://gwas.mrcieu.ac.uk/datasets/ukb-b-9405/
Hypertension	European	484,598	https://gwas.mrcieu.ac.uk/datasets/ebi-a-GCST90038604/
HDL-C	Mixed (96% European)	187,167	https://gwas.mrcieu.ac.uk/datasets/ieu-a-299/
Triglycerides	Mixed (96% European)	177,861	https://gwas.mrcieu.ac.uk/datasets/ieu-a-302/
IBD	European	500,348	FinnGen consortium
UC	European	499,380	FinnGen consortium
CD	European	500,111	FinnGen consortium

Table 2 Results of the forward Mendelian randomization analysis for the effect of metabolic syndrome on inflammatory bowel disease

Exposure	Outcome	No. of SNPs	Methods	OR	Lower 95% CI	Upper 95% CI	P	MR-Egger intercept (P value)	Cochran's Q test (I ²)	P
MetS	IBD	184	IVW	1.023	0.861	1.216	0.793	0.522	348.189 (47.44%)	<0.001
			WM	1.070	0.852	1.343	0.562			
			MR-Egger	0.885	0.549	1.426	0.616			
			MRPRESSO	1.023	1.010	1.037	0.793			
FBG	IBD	20	IVW	0.869	0.715	1.056	0.158	0.140	20.109 (5.51%)	0.388
			WM	0.990	0.769	1.275	0.937			
			MR-Egger	1.124	0.770	1.641	0.551			
			MRPRESSO	0.869	0.832	0.908	0.174			
WC	IBD	363	IVW	1.058	0.938	1.195	0.358	0.973	486.05 (25.52%)	<0.001
			WM	1.025	0.861	1.220	0.784			
			MR-Egger	1.064	0.752	1.507	0.725			
			MRPRESSO	1.058	1.052	1.065	0.358			
Hypertension	IBD	236	IVW	1.350	0.971	1.878	0.075	0.397	383.977 (38.8%)	<0.001
			WM	2.005	1.309	3.071	0.001			
			MR-Egger	1.905	0.805	4.508	0.144			
			MRPRESSO	1.350	1.321	1.379	0.076			
HDL-C	IBD	82	IVW	1.029	0.925	1.144	0.600	0.883	166.921 (51.47%)	<0.001
			WM	0.988	0.871	1.122	0.854			
			MR-Egger	1.016	0.834	1.238	0.873			
			MRPRESSO	1.029	1.017	1.041	0.601			

Triglycerides	IBD	50	IVW	0.980	0.870	1.103	0.736	0.227	81.006	0.003
			WM	0.933	0.808	1.077	0.344		(39.51%)	
			MR-Egger	0.896	0.744	1.079	0.252			
			MRPRESSO	0.980	0.964	0.996	0.738			
MetS	UC	185	IVW	0.944	0.762	1.169	0.596	0.523	359.544	<0.001
			WM	0.939	0.700	1.261	0.676		(48.82%)	
			MR-Egger	0.788	0.436	1.426	0.432			
			MRPRESSO	0.944	0.929	0.959	0.597			
FBG	UC	21	IVW	0.825	0.656	1.037	0.099	0.046	20.227	0.444
			WM	0.926	0.658	1.303	0.659		(1.12%)	
			MR-Egger	1.280	0.805	2.034	0.310			
			MRPRESSO	0.825	0.785	0.867	0.115			
WC	UC	365	IVW	1.015	0.876	1.176	0.845	0.562	483.305	<0.001
			WM	0.946	0.752	1.191	0.637		(24.69%)	
			MR-Egger	0.902	0.591	1.378	0.634			
			MRPRESSO	1.015	1.007	1.023	0.845			
Hypertension	UC	240	IVW	1.610	1.084	2.390	0.018	0.112	374.894	<0.001
			WM	2.211	1.326	3.686	0.002		(36.25%)	
			MR-Egger	3.498	1.245	9.828	0.018			
			MRPRESSO	1.610	1.569	1.651	0.019			
HDL-C	UC	82	IVW	1.017	0.897	1.154	0.788	0.970	152.668	<0.001
			WM	0.985	0.827	1.174	0.869		(46.94%)	
			MR-Egger	1.013	0.801	1.282	0.911			
			MRPRESSO	1.017	1.003	1.032	0.788			
Triglycerides	UC	50	IVW	0.999	0.880	1.135	0.993	0.381	69.875	0.027
			WM	0.909	0.763	1.083	0.285		(29.88%)	

MetS	CD	190	MR-Egger	0.933	0.765	1.138	0.497	0.642	231.022 (18.19%)	0.02
			MRPRESSO	0.999	0.982	1.018	0.993			
			IVW	1.340	1.009	1.779	0.043			
			WM	1.336	0.887	2.014	0.166			
FBG	CD	21	MR-Egger	1.594	0.726	3.501	0.247	0.508	27.131 (26.28%)	0.132
			MRPRESSO	1.340	1.312	1.368	0.045			
			IVW	0.767	0.489	1.205	0.250			
			WM	0.995	0.569	1.740	0.985			
WC	CD	369	MR-Egger	1.012	0.401	2.553	0.980	0.323	443.917 (17.1%)	0.004
			MRPRESSO	0.767	0.695	0.847	0.263			
			IVW	1.330	1.050	1.684	0.018			
			WM	1.389	0.971	1.987	0.072			
Hypertension	CD	242	MR-Egger	1.834	0.930	3.617	0.081	0.566	304.114 (20.75%)	0.004
			MRPRESSO	1.330	1.313	1.346	0.018			
			IVW	1.000	0.553	1.810	0.999			
			WM	0.780	0.340	1.789	0.557			
HDL-C	CD	88	MR-Egger	0.656	0.138	3.114	0.596	0.294	126.869 (31.43%)	0.003
			MRPRESSO	1.000	0.963	1.039	0.999			
			IVW	0.835	0.696	1.001	0.052			
			WM	0.905	0.706	1.161	0.434			
Triglycerides	CD	53	MR-Egger	0.716	0.511	1.004	0.056	0.968	72.906 (28.68%)	0.029
			MRPRESSO	0.835	0.819	0.851	0.055			
			IVW	1.098	0.880	1.371	0.407			
			WM	1.056	0.795	1.403	0.707			
			MR-Egger	1.105	0.771	1.584	0.590			
			MRPRESSO	1.098	1.065	1.132	0.411			

MetS, metabolic syndrome; FBG, fasting blood glucose; WC, waist circumference; HDL-C, high-density lipoprotein cholesterol; IBD, inflammatory bowel disease; UC, ulcerative colitis; CD, crohn's disease

Table 3 Results of the reverse Mendelian randomization analysis for the effect of inflammatory bowel disease on metabolic syndrome

Exposure	Outcome	No. of SNPs	Methods	OR	Lower 95% CI	Upper 95% CI	P	MR-Egger intercept (P value)	Cochran's Q test (I ²)	P
IBD	MetS	21	IVW	1.007	0.998	1.016	0.141	0.164	23.041 (13.2%)	0.287
			WM	1.008	0.995	1.021	0.249			
			MR-Egger	0.986	0.956	1.016	0.362			
			MRPRESSO	1.007	1.005	1.009	0.157			
IBD	FBG	25	IVW	0.994	0.980	1.009	0.450	0.385	30.476 (21.25%)	0.169
			WM	0.984	0.964	1.004	0.125			
			MR-Egger	0.972	0.924	1.024	0.298			
			MRPRESSO	0.994	0.991	0.997	0.458			
IBD	WC	53	IVW	1.004	0.998	1.009	0.190	0.533	72.895 (28.67%)	0.029
			WM	1.006	0.999	1.013	0.073			
			MR-Egger	0.999	0.986	1.014	0.938			
			MRPRESSO	1.004	1.003	1.004	0.196			
IBD	Hypertension	63	IVW	1.001	0.999	1.004	0.427	0.876	83.587 (25.83%)	0.035
			WM	1.001	0.998	1.005	0.400			

IBD	HDL-C	22	MR-Egger	1.000	0.995	1.007	0.777	0.840	32.505 (35.39%)	0.052
			MRPRESSO	1.001	1.001	1.002	0.275			
			IVW	0.995	0.972	1.018	0.644			
			WM	0.995	0.967	1.025	0.751			
IBD	Triglycerides	23	MR-Egger	1.002	0.930	1.079	0.960	0.294	24.074 (8.61%)	0.343
			MRPRESSO	0.995	0.990	0.999	0.649			
			IVW	1.019	1.000	1.038	0.049			
			WM	1.009	0.983	1.035	0.511			
UC	MetS	14	MR-Egger	0.988	0.932	1.048	0.699	0.696	15.195 (19.97%)	0.295
			MRPRESSO	1.019	1.015	1.023	0.062			
			IVW	0.996	0.987	1.006	0.457			
			WM	1.004	0.992	1.017	0.495			
UC	FBG	19	MR-Egger	1.003	0.968	1.039	0.861	0.054	20.707 (13.07%)	0.294
			MRPRESSO	0.996	0.994	0.999	0.470			
			IVW	1.002	0.989	1.016	0.726			
			WM	0.988	0.970	1.006	0.206			
UC	WC	38	MR-Egger	0.957	0.913	1.002	0.077	0.704	72.739 (49.13%)	<0.001
			MRPRESSO	1.002	0.999	1.006	0.730			
			IVW	1.003	0.997	1.010	0.279			
			WM	1.006	0.999	1.013	0.070			
UC	Hypertension	42	MR-Egger	1.007	0.989	1.025	0.468	0.443	90.352 (47.98%)	<0.001
			MRPRESSO	1.003	1.002	1.004	0.286			
			IVW	1.000	0.998	1.003	0.545			
			WM	1.000	0.997	1.003	0.842			
			MR-Egger	1.004	0.996	1.011	0.352			
			MRPRESSO	1.000	1.000	1.001	0.538			

UC	HDL-C	15	IVW	1.000	0.980	1.021	0.966	0.949	19.628	0.142
			WM	0.997	0.973	1.021	0.803		(28.67%)	
			MR-Egger	1.003	0.930	1.082	0.942			
			MRPRESSO	1.000	0.995	1.006	0.966			
UC	Triglycerides	17	IVW	1.011	0.995	1.028	0.173	0.877	11.301	0.791
			WM	1.005	0.983	1.027	0.667		(41.58%)	
			MR-Egger	1.016	0.962	1.072	0.581			
			MRPRESSO	1.011	1.008	1.015	0.124			
CD	MetS	5	IVW	0.997	0.987	1.007	0.560	0.423	3.516	0.475
			WM	0.997	0.983	1.011	0.648		(13.77%)	
			MR-Egger	1.031	0.960	1.107	0.466			
			MRPRESSO	0.997	0.993	1.001	0.568			
CD	FBG	5	IVW	1.008	0.989	1.027	0.414	0.883	6.291	0.178
			WM	1.012	0.991	1.033	0.251		(36.42%)	
			MR-Egger	0.998	0.884	1.127	0.978			
			MRPRESSO	1.008	0.999	1.016	0.460			
CD	WC	8	IVW	1.002	0.996	1.008	0.514	0.661	2.252	0.944
			WM	1.004	0.997	1.012	0.254		(210.87%)	
			MR-Egger	1.006	0.989	1.023	0.533			
			MRPRESSO	1.002	1.000	1.003	0.288			
CD	Hypertension	48	IVW	1.001	0.998	1.003	0.535	0.443	90.352	<0.001
			WM	1.000	0.997	1.003	0.842		(47.98%)	
			MR-Egger	1.004	0.996	1.011	0.352			
			MRPRESSO	1.000	1.000	1.001	0.538			
CD	HDL-C	7	IVW	0.991	0.966	1.018	0.520	0.368	12.371	0.054
			WM	0.995	0.969	1.022	0.705		(51.5%)	

CD	Triglycerides	7	MR-Egger	1.072	0.916	1.254	0.425	0.169	838.822 (99.28%)	<0.001
			MRPRESSO	0.991	0.982	1.001	0.544			
			IVW	1.149	0.931	1.417	0.196			
			WM	1.025	0.997	1.055	0.081			
			MR-Egger	0.464	0.151	1.425	0.237			
			MRPRESSO	1.149	1.061	1.244	0.243			

MetS, metabolic syndrome; FBG, fasting blood glucose; WC, waist circumference; HDL-C, high-density lipoprotein cholesterol; IBD, inflammatory bowel disease; UC, ulcerative colitis; CD, crohn's disease

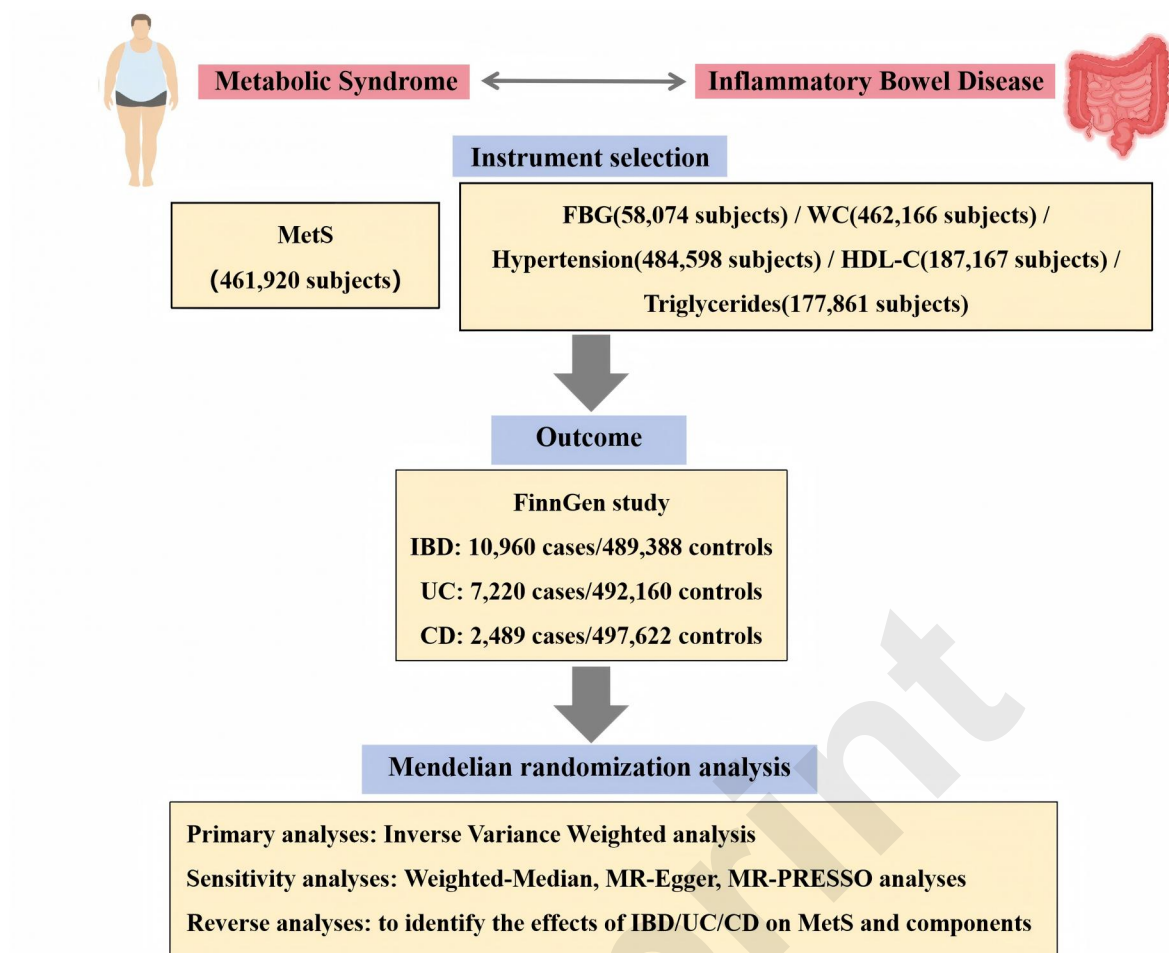


Figure 1 The overall design of the present Mendelian randomization analysis.

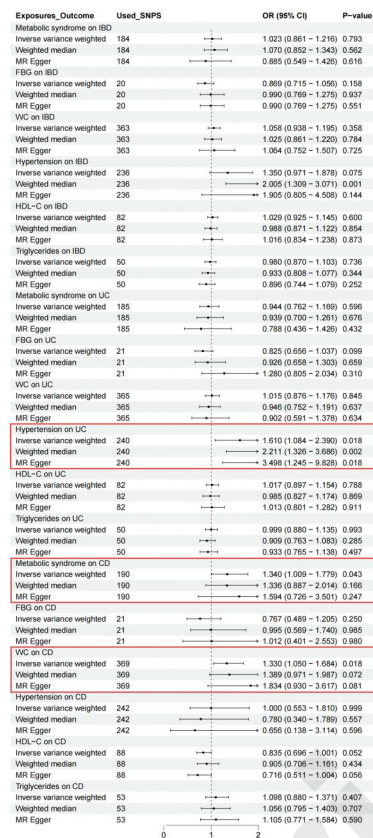


Figure 2 Genetically predicted association between MetS and IBD by forward MR Analysis.

Exposures_Outcome	Used_SNPs		OR (95% CI)	P-value
IBD on Metabolic syndrome				
Inverse variance weighted	21	+	1.007 (0.998 – 1.016)	0.141
Weighted median	21	+	1.008 (0.995 – 1.021)	0.249
MR Egger	21	+	0.986 (0.956 – 1.016)	0.362
IBD on FBG				
Inverse variance weighted	25	+	0.994 (0.980 – 1.009)	0.450
Weighted median	25	+	0.984 (0.964 – 1.004)	0.125
MR Egger	25	+	0.972 (0.924 – 1.024)	0.298
IBD on WVC				
Inverse variance weighted	53	+	1.004 (0.998 – 1.009)	0.190
Weighted median	53	+	1.006 (0.999 – 1.013)	0.073
MR Egger	53	+	0.999 (0.986 – 1.014)	0.938
IBD on Hypertension				
Inverse variance weighted	63	+	1.001 (0.999 – 1.004)	0.271
Weighted median	63	+	1.001 (0.998 – 1.005)	0.400
MR Egger	63	+	1.001 (0.995 – 1.007)	0.777
IBD on HDL-C				
Inverse variance weighted	22	+	0.995 (0.972 – 1.018)	0.644
Weighted median	22	+	0.995 (0.967 – 1.025)	0.751
MR Egger	22	+	1.002 (0.930 – 1.079)	0.960
IBD on Triglycerides				
Inverse variance weighted	23	+	1.019 (1.000 – 1.038)	0.049
Weighted median	23	+	1.009 (0.983 – 1.035)	0.511
MR Egger	23	+	0.988 (0.932 – 1.048)	0.699
UC on Metabolic syndrome				
Inverse variance weighted	14	+	0.996 (0.987 – 1.005)	0.457
Weighted median	14	+	1.004 (0.992 – 1.017)	0.495
MR Egger	14	+	1.003 (0.965 – 1.039)	0.861
UC on FBG				
Inverse variance weighted	19	+	1.002 (0.989 – 1.016)	0.726
Weighted median	19	+	0.988 (0.970 – 1.006)	0.206
MR Egger	19	+	0.957 (0.913 – 1.002)	0.077
UC on WVC				
Inverse variance weighted	38	+	1.003 (0.997 – 1.010)	0.279
Weighted median	38	+	1.006 (0.999 – 1.013)	0.070
MR Egger	38	+	1.007 (0.989 – 1.025)	0.468
UC on Hypertension				
Inverse variance weighted	48	+	1.001 (0.998 – 1.003)	0.535
Weighted median	48	+	1.000 (0.997 – 1.003)	0.942
MR Egger	48	+	1.004 (0.996 – 1.011)	0.352
UC on HDL-C				
Inverse variance weighted	15	+	1.000 (0.980 – 1.021)	0.960
Weighted median	15	+	0.997 (0.973 – 1.021)	0.803
MR Egger	15	+	1.003 (0.930 – 1.082)	0.942
UC on Triglycerides				
Inverse variance weighted	17	+	1.011 (0.995 – 1.028)	0.173
Weighted median	17	+	1.005 (0.983 – 1.027)	0.667
MR Egger	17	+	1.016 (0.962 – 1.072)	0.581
CD on Metabolic syndrome				
Inverse variance weighted	5	+	0.997 (0.987 – 1.007)	0.560
Weighted median	5	+	0.997 (0.983 – 1.011)	0.648
MR Egger	5	+	1.031 (0.960 – 1.107)	0.466
CD on FBG				
Inverse variance weighted	5	+	1.008 (0.989 – 1.027)	0.414
Weighted median	5	+	1.012 (0.991 – 1.033)	0.251
MR Egger	5	+	0.998 (0.884 – 1.127)	0.978
CD on WVC				
Inverse variance weighted	8	+	1.002 (0.996 – 1.008)	0.514
Weighted median	8	+	1.004 (0.997 – 1.012)	0.254
MR Egger	8	+	1.008 (0.989 – 1.029)	0.533
CD on Hypertension				
Inverse variance weighted	9	+	1.003 (1.000 – 1.006)	0.085
Weighted median	9	+	1.003 (0.999 – 1.007)	0.203
MR Egger	9	+	1.002 (0.994 – 1.011)	0.574
CD on HDL-C				
Inverse variance weighted	7	+	0.991 (0.966 – 1.016)	0.520
Weighted median	7	+	0.995 (0.965 – 1.022)	0.705
MR Egger	7	+	1.072 (0.918 – 1.254)	0.425
CD on Triglycerides				
Inverse variance weighted	7	+	1.149 (0.931 – 1.417)	0.196
Weighted median	7	+	1.025 (0.997 – 1.055)	0.081
MR Egger	7	+	0.464 (0.151 – 1.425)	0.237

Figure 3 Genetically predicted association between IBD and MetS by reverse MR Analysis.

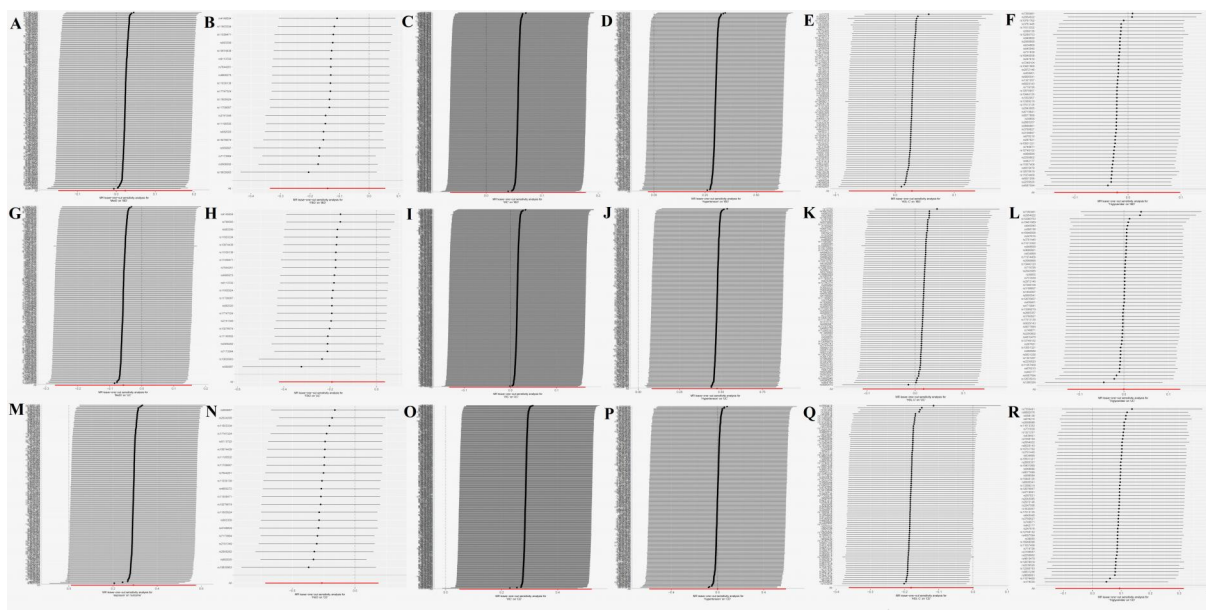


Figure S2 The forward MR analyses: Plots of “leave-one-out” analyses for MR analyses of the causal effect of metabolic syndrome with the risk of inflammatory bowel disease and subtypes.

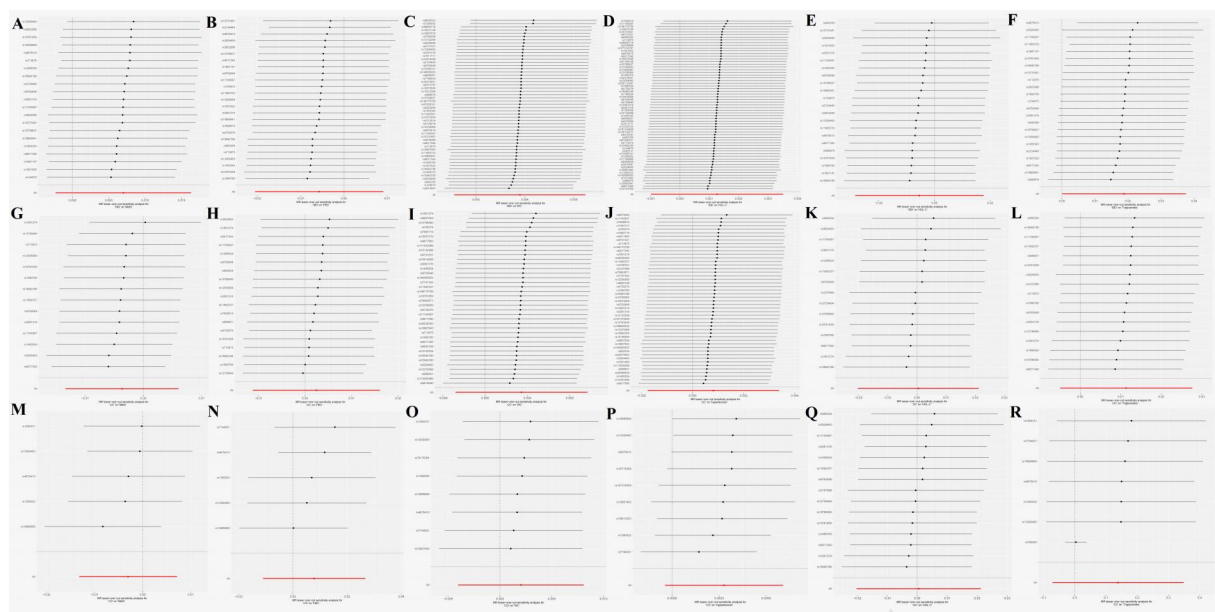


Figure S4 The reverse MR analyses: Plots of "leave-one-out" analyses for MR analyses of the causal effect of inflammatory bowel disease and subtypes with the risk of metabolic syndrome.