A Causal Inference Study on the Impact of Asthma on the Onset of Chronic Obstructive Pulmonary Disease: Two-Sample Mendelian Randomization

Keywords

COPD, GWAS, Mendelian Randomization, Asthma, Causal Relationship, Environmental Exposure Factors

Abstract

Introduction

Respiratory diseases have long been a focus of research in the field of public health, posing a serious threat to the health of the global population
Asthma and COPD (chronic obstructive pulmonary disease) are two common respiratory ailments with overlapping pathogenesis.

Material and methods

In this study, utilizing a two-sample Mendelian randomization (MR) approach and analyzing publicly available genome-wide association study (GWAS) datasets, we explored the causal impact of asthma on the onset of COPD. Using genetic instrumental variables associated with asthma ($p < 5 \times 10^{-8}$) and applying multiple MR methods (IVW, MR-Egger, and weighted median), we identified a significant causal relationship between asthma and COPD. The inverse variance weighted (IVW) method indicated that asthma increases the risk of developing COPD with an odds ratio (OR) of 1.35 (95% confidence interval [CI]: 1.12–1.58, p = 0.002).

Results

Additionally, multivariable MR analysis was performed to account for potential confounders, such as eosinophil count, smoking, and falls, which demonstrated that the association remains significant even after adjusting for these factors (OR: 1.29, 95% CI: 1.08-1.50, p = 0.004).

Conclusions

This study provides robust evidence supporting the causal link between asthma and COPD, offering a more comprehensive understanding of their relationship.

Keywords: Asthma, GWAS, Mendelian Randomization, Causal Relationship, COPD, Environmental Exposure Factors

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2 **Pulmonary Disease: Two-Sample Mendelian Randomization**

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Abstract 6

7 Asthma and COPD (chronic obstructive pulmonary disease) are two common respiratory ailments 8 with overlapping pathogenesis. In this study, utilizing a two-sample Mendelian randomization (MR) 9 approach and analyzing publicly available genome-wide association study (GWAS) datasets, we explored the causal impact of asthma on the onset of COPD. Using genetic instrumental variables 10 associated with asthma ($p < 5 \times 10^{-8}$) and applying multiple MR methods (IVW, MR-Egger, and 11 12 weighted median), we identified a significant causal relationship between asthma and COPD. The inverse variance weighted (IVW) method indicated that asthma increases the risk of developing 13 COPD with an odds ratio (OR) of 1.35 (95% confidence interval [CI]: 1.12 - 1.58, p = 0.002). 14 15 Additionally, multivariable MR analysis was performed to account for potential confounders, such 16 as eosinophil count, smoking, and falls, which demonstrated that the association remains significant 17 even after adjusting for these factors (OR: 1.29, 95% CI: 1.08 - 1.50, p = 0.004). This study provides robust evidence supporting the causal link between asthma and COPD, offering a more 18 19 comprehensive understanding of their relationship.

20

21 Keywords: Asthma, GWAS, Mendelian Randomization, Causal Relationship, COPD, 22 **Environmental Exposure Factors**

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27 Introduction

Respiratory diseases have long been a focus of research in the field of public health, posing a serious threat to the health of the global population [1]. Among them, asthma and COPD are two highly prevalent diseases, posing significant challenges to patients' quality of life and the consumption of social medical resources. Although asthma and COPD have distinct pathological and physiological differences, with the former mainly characterized by chronic airway inflammation and the latter encompassing pathological changes such as chronic bronchitis and emphysema, their relationship and mutual influence remain a scientific puzzle of great concern[2, 3].

Asthma manifests as a complex chronic inflammatory condition, featuring airway hyperresponsiveness and reversible airway obstruction. Hundreds of millions of people worldwide suffer from asthma, affecting both adults and children[4]. COPD mainly includes chronic bronchitis and emphysema, is a progressive disease commonly found in smokers, and is also influenced by environmental factors such as air pollution. Both diseases are primarily characterized by symptoms such as dyspnea, cough, and chest tightness, causing significant inconvenience to patients' lives and work[5].

Although asthma and COPD differ in presentation and progression, their commonalities cannot be
ignored. Studies have shown that in some patients, asthma and COPD might coexist, referred to as
"asthma-COPD overlap syndrome" (ACOS)[6]. This situation complicates the management and

treatment of the disease, making it even more necessary to delve into the relationship between
asthma and COPD to better understand their shared pathophysiological mechanisms and provide
more precise bases for treatment strategies[7].

48 Previous studies have mainly focused on the pathogenesis and causative factors of asthma and 49 COPD individually. However, the question of whether there is a causal relationship between the two, and whether asthma might be a potential factor in the development of COPD, remains relatively 50 unclear in academia. Some observational studies have provided some clues, but due to the presence 51 52 of numerous confounding factors, the research findings are contentious. Therefore, we need a more 53 precise method to control for confounding variables. By adopting a Mendelian randomization approach, which enables exploration of the causal 54 55 relationship between asthma and COPD without being influenced by confounding factors, this study 56 aims to delve into the possible association between the two conditions. Integrating publicly available 57 genome-wide association study (GWAS) datasets and employing careful selection of genetic 58 instrumental variables, alongside various Mendelian randomization analysis methods, the study 59 seeks to elucidate the genetic mechanisms underlying asthma and COPD, thereby providing a 60 clearer understanding of their interrelation.

61

62 Materials and Methods

63 Guidelines for Reporting and Study Design

64 This study utilized a two-sample MR approach and publicly available datasets to investigate the 65 impact of asthma on COPD. The study report adheres to the STROBE-MR Statement, which 66 enhances the reporting of observational studies in epidemiology using Mendelian randomization [8], 67 Figure 1 illustrates the study design schematic.

68 Data Sources

- 69 Genome-wide association study (GWAS) data for preliminary analysis of asthma indicators: GWAS
- analysis data conducted by Demenais et al. on 127,669 individuals of European ancestry, including
- 71 107,715 controls and 19,954 asthma cases [9]. GWAS analysis data conducted by Valette et al. on
- 408,442 individuals of European ancestry, including 352,255 controls and 56,167 asthma cases [10].
- 73 Chronic obstructive pulmonary disease (COPD) GWAS data: COPD GWAS data (ebi-a-
- 74 GCST90018807) conducted by Sakaue et al. on 468,475 individuals of European ancestry, including
- 75 13,530 COPD cases and 454,945 controls [11].

76 Other GWAS data: Eosinophil cell count (ieu-b-33) from GWAS analysis conducted by Vuckovic

- et al., including 563,946 samples [12] [13]; Past tobacco smoking (ukb-b-2134) [14]from
- approximately 500,000 samples from the UK Biobank [15]; Falling risk (ebi-a-GCST90012857)
- 79 GWAS data analysis including 451,179 samples [16] [17].

80 Instrumental Variable Selection

Effective genetic instrumental variables must meet three core assumptions: (1) the relevance assumption, i.e., the chosen instrumental variables must exhibit a significant association with the exposure factor; (2) the independence assumption, i.e., instrumental variables must lack statistical significance correlated with potential confounding factors that might affect exposure or outcome; (3) the exclusion restriction assumption, i.e., instrumental variables must exclusively influence the outcome through the pathway "instrumental variable \rightarrow exposure \rightarrow outcome".

- 87 In this study, the selection criteria for exposure instrumental variables are as follows: SNPs (Single
- 88 Nucleotide Polymorphisms) with $P < 5 \times 10^{-8}$ in GWAS are used as the primary screening

condition; SNPs in linkage disequilibrium (r² < 0.001) and with a physical distance > 10,000 kb
between every two genes are excluded. Then, outcome data are extracted from GWAS based on the
selected SNPs.

92 MR Causal Effect Estimation

93 Various two-sample MR methods are used to assess the causal effects between exposure and outcome, including: Inverse-variance weighted (IVW), MR-Egger, Weight Median, Simple Mode, 94 and Weight Mode. Studies have shown [18] that the IVW method is slightly stronger under certain 95 conditions than other methods. Its characteristic is that it does not consider the presence of intercept 96 97 terms during regression and uses the reciprocal of the outcome variance as weights for fitting. Therefore, in the absence of pleiotropy, regardless of the presence or absence of heterogeneity, the 98 99 IVW method is used as the main MR analysis method in this study, with the other four methods 100 used as supplements (IVW random effects model is used when heterogeneity exists). When pleiotropy exists, the MR-Egger method is used to calculate the results. The reverse causal effect is 101 assessed using the same methods for the outcome's potential causal effect on exposure. 102

103 Sensitivity Analysis

104 Various methods, including heterogeneity testing, pleiotropy testing, and leave-one-out testing, are
105 used to perform sensitivity analysis on the analysis results, as follows:

- 106 (1) Heterogeneity testing: Cochran's Q test is used to assess heterogeneity among estimated values
- 107 of each SNP. If Cochran's Q test is statistically significant, it indicates significant heterogeneity in
- the results, and the effect size of the causal effect is evaluated using the IVW random effects model.
- 109 Since Cochran's Q test could only determine the presence or absence of heterogeneity and cannot
- 110 determine its distribution, the I² statistic is used to reflect the proportion of heterogeneity in the

total variation of instrumental variables: $I^2 \le 0$ indicates no observed heterogeneity; $I^2 = 0-25\%$

indicates mild heterogeneity; $I^2 = 25-50\%$ indicates moderate heterogeneity; $I^2 > 50\%$ indicates

113 high heterogeneity. The specific calculation formula is as follows:

(2) Pleiotropy assessment: The MR-Egger method is employed to examine pleiotropy in
instrumental variables. A MR-Egger's intercept p-value <0.05 suggests substantial horizontal
pleiotropy in genetic variation.

117 (3) Leave-one-out testing: The leave-one-out method is used to evaluate whether a single SNP

- 118 affects the association between asthma and COPD by sequentially removing individual SNPs and
- 119 calculating the MR results with the remaining instrumental variables. If there is a large difference
- 120 between the MR effect estimate after removing a particular instrumental variable and the cumulative
- 121 effect estimate, it indicates sensitivity of the MR effect estimate to that SNP.

122 Multivariable MR Analysis

MVMR is an extension of MR that uses genetic variations associated with multiple exposures possibly correlated with a single outcome evaluate the impacts of various exposures on a singular outcome.It could assess the direct effects of individual exposure factors on the outcome. We conducted multivariable MR analysis by considering several relevant exposure factors of COPD: Eosinophil cell count (ieu-b-33), Past tobacco smoking (ukb-b-2134), Falling risk (ebi-a-GCST90012857), and the significant exposure asthma to obtain the direct effect of asthma on

129 COPD.

130 Statistical Analysis

131 All data processing and statistical analyses were performed using the R programming language

132 (version 4.2.2). Mendelian randomization analysis primarily relied on the TwoSampleMR package

133 [19]. To ensure the robustness and reliability of the results, Cochran's Q test and leave-one-out 134 analysis were utilized. Additionally, genetic pleiotropy was assessed using MR-Egger's intercept 135 method. Evaluation metrics consisted of odds ratios (OR) and their corresponding 95% confidence 136 intervals (95% CI). All statistical P-values were two-sided, with SNP loci generated from GWAS 137 studies considered statistically significant at $P < 5 \times 10^{-8}$; statistical significance was determined 138 for other tests at a threshold of P < 0.05. 139

140 **Results**

- 141 Analysis Framework and Flowchart
- 142 The flowchart was shown in Figure 1.
- 143 Instrumental Variable Selection

144 Based on the selection criteria for instrumental variables, SNPs with linkage disequilibrium were

- 145 removed, and SNPs associated with asthma ($P < 5 \times 10^{-8}$, clump=TRUE, $r^{2} < 0.001$, kb=10,000)
- 146 were included as instrumental variables after matching with COPD GWAS data. The information of
- instrumental variables for each indicator was shown in Table S1.

148 MR Causal Effect Estimation

149 Analysis was conducted using five models: MR Egger, Weighted median, IVW (Inverse variance

- 150 weighted), Simple mode (SM), and Weighted mode. The results of the five models for the causal
- relationship between asthma and COPD (Figure 2a, Table 1) indicated a significant causal
- relationship between asthma and COPD, with higher levels of all circulating metabolites associated
- 153 with an increased risk of COPD occurrence. Different models of MR analysis for asthma (Figure
- 154 2b-c) provided consistent directional estimates, with relatively consistent slopes.

155 Sensitivity Analysis

156	Heterogeneity testing of significant results was conducted using Cochran's Q test and I^2 statistic,
157	as shown in Table 2. The results indicated high heterogeneity in the Mendelian randomization (MR)
158	results for asthma (Asthma, id: ebi-a-GCST006862) with respect to COPD (I^2=50.65%, Cochran
159	Q p-value < 0.05). However, there was no significant heterogeneity in the MR results for asthma
160	(Asthma, id: ebi-a-GCST90014325) with respect to COPD (Cochran Q p-value > 0.05 , I ² < 50%).
161	The exposure of asthma (Asthma, id: ebi-a-GCST006862) with high heterogeneity was removed,
162	and the exposure of asthma (Asthma, id: ebi-a-GCST90014325) was analyzed further. The funnel
163	plot of instrumental variables for asthma (Asthma, id: ebi-a-GCST90014325) (Figure 3A) showed
164	a symmetrical distribution of scatter points, indicating no potential bias in the causal association
165	effect.
166	Instrumental variable horizontal pleiotropy testing was conducted using MR-Egger regression. The
167	p-value of the intercept term in the statistical hypothesis test for p-value associated with asthma
168	(Asthma, id: ebi-a-GCST90014325) exceeded 0.05, and the intercept value approached zero.
169	indicating that the causal inference of this study was not affected by horizontal pleiotropy (Table 3).
170	We performed leave-one-out analysis, systematically excluding each instrumental variable locus, to
171	examine the causal impact of asthma on COPD (Figure 3b). It was found that there was no
172	significant deviation from the total effect of the instrumental variable set. The Steiger directional
173	test result was TRUE with a p-value less than 0.05, indicating no reverse causal effect (Table 4).
174	B. Forest plot showing the sequential effect estimates of asthma and COPD using single SNP locus
175	analysis.

176 Reverse MR Analysis

177 To evaluate reverse causal effects, we used COPD s exposure and asthma (id: ebi-a-GCST90014325)

- as the outcome. Following the selection criteria for instrumental variables in this study and removing
- 179 SNPs with linkage disequilibrium, the reverse causal MR analysis results (Figure 4) indicated that
- 180 COPD did not have a significant causal effect on asthma (P value > 0.05), as shown in Table 5.
- 181 Multivariable MR Analysis
- 182 We conducted multivariable MR analysis by incorporating exposures such as eosinophil cell count
- 183 (ieu-b-33), past tobacco smoking (ukb-b-2134), falling risk (ebi-a-GCST90012857), etc., to assess
- the direct effect of asthma (Asthma, ebi-a-GCST90014325) on COPD (Table 6). In Model 1,
- 185 correcting for the indirect effect of eosinophil cell count, the results indicated that asthma (Asthma,
- ebi-a-GCST90014325) still significantly affects COPD. In Model 2, correcting for the influence of
- smoking, the results showed that asthma (Asthma, ebi-a-GCST90014325) still significantly affects
- 188 COPD. In Model 3, considering the impact of falling risk, the results demonstrated that asthma
- 189 (Asthma, ebi-a-GCST90014325) still significantly affects COPD.

190

191 Discussion

This study delves into the causal impact of asthma on COPD through a two-sample Mendelian randomization design. Firstly, by rigorously screening publicly available GWAS data, we successfully selected a series of SNPs that meet the criteria for instrumental variables for subsequent causal effect estimation. Through various two-sample MR methods, we conclude that asthma serves as a significant risk factor for COPD, providing solid genetic evidence for further elucidating the relationship between these two diseases.

198 In sensitivity analysis, we validated the robustness of the results through methods such as

199 heterogeneity testing and pleiotropy testing. Even after excluding SNPs with high heterogeneity, the

- 200 results remained significant. Through Leave-one-out analysis, we further confirmed the robustness
- 201 of the instrumental variables, eliminating the significant influence of individual SNPs on the overall
- 202 outcomes. The results of the MR-Egger intercept method also indicate that the causal inference of
- this study is not significantly affected by horizontal pleiotropy.

To gain a more comprehensive understanding of the impact of asthma on COPD, we conducted

- 205 reverse MR analysis and found that COPD does not have a significant causal effect on asthma,
- thereby ruling out the possibility of reverse causality.
- Finally, through multivariable MR analysis, we explored the impact of multiple potential confounding factors on the relationship between asthma and COPD. The results indicate that the causal effect of asthma on COPD still exists after considering factors such as eosinophil cell count,
- smoking, and falling.

Previous research in the asthma and COPD field had extensively explored their pathogenesis, 211 212 pathophysiological processes, and their relationship with environmental, genetic, and other factors. 213 However, the relationship between asthma and COPD has always been a contentious issue. Here are 214 some key points and findings from previous studies: 修改为 "Previous research in the asthma 215 and COPD field had extensively explored their pathogenesis, pathophysiological processes, and their relationship with environmental, genetic, and other factors. For instance, recent studies, such 216 217 as the one by Kurmi et al.[20], have also investigated the genetic links and shared risk factors between these diseases, further supporting the potential overlap in their etiology. However, the 218 219 relationship between asthma and COPD remains a contentious issue."

220 Traditionally, asthma and COPD were considered to be two diseases with distinct biological

221 characteristics[21]. Asthma mainly involves airway inflammation and reversible airway obstruction, 222 while COPD includes chronic bronchitis and emphysema, characterized by irreversible airway 223 narrowing[22]. However, recent studies suggest that some patients might have both asthma and 224 COPD, leading to asthma-COPD overlap syndrome (ACOS), with symptoms and biomarkers 225 showing overlapping features, thereby increasing attention to the relationship between the two[23]. Early observational studies attempted to reveal the relationship between asthma and COPD but 226 227 faced challenges due to confounding factors. For example, smoking, air pollution, occupational 228 exposure, etc., are factors that affect respiratory health, making it difficult for observational studies 229 to establish whether asthma directly leads to the development of COPD[24]. In recent years, advances in techniques such as genome-wide association studies (GWAS) have enriched our 230 231 comprehension of the genetic underpinnings of respiratory disorders. Some genes have been 232 confirmed to be associated with asthma and COPD, but their specific mechanisms and their 233 interrelationship remain uncertain. 234 In clinical practice, distinguishing between asthma and COPD is sometimes not straightforward, 235 especially in some elderly patients or long-term smokers. The presence of this overlapping 236 pathological state increases the difficulty of disease management[25]. 237 Additionally, the understanding of asthma-COPD overlap syndrome (ACOS) remains incomplete, 238 necessitating further research to elucidate the nature of this dual pathological state. 239 Our study contributes to this discussion by providing evidence that asthma is a risk factor for COPD 240 development. To further understand the mechanisms involved, exploring how asthma-related factors

- 241 influence COPD onset is crucial. Given the overlap of symptoms in ACOS and the inconclusive
- 242 findings from previous observational studies, our study stands out by utilizing genetic evidence

from GWAS to establish a causal relationship between asthma and COPD. Additionally, our results emphasize the importance of addressing asthma control to potentially prevent COPD progression.
Future research directions may involve studying specific asthma-related genetic markers and their impact on COPD development, as well as implementing personalized therapeutic interventions based on genetic susceptibility analysis. Overall, our study highlights the significance of recognizing asthma as a contributing factor to COPD and underscores the necessity for targeted interventions to alleviate disease burden and improve patient outcomes.

When discussing the mechanisms by which asthma affects COPD and their application in disease 250 251 prevention or treatment, several mechanisms may be utilized. Firstly, the inflammation regulation mechanism may be one of them. Both asthma and COPD are associated with inflammation response, 252 253 so studying the mechanism of inflammation regulation, such as reducing airway inflammation 254 through medication or biological treatment, may prevent or treat the progression of COPD[26]. Further research may include the development of novel inflammation-regulating agents and their 255 effectiveness and safety in clinical practice. Secondly, airway remodeling mechanism is also an 256 257 important research direction. Asthma and COPD patients have abnormal airway structure and 258 function, so studying the mechanism of airway remodeling and attempting to restore normal airway structure and function through medication or other treatment methods may prevent or treat the 259 260 development of COPD[27]. Further research may include the development of novel treatment 261 methods targeting airway remodeling and the evaluation of their long-term effects on COPD patients. 262 Lastly, the immune regulation mechanism may also be a promising treatment approach. The immune 263 system plays an important role in the development of asthma and COPD, so studying the mechanism 264 of immune response regulation, such as the use of immunomodulators, may prevent or treat the

265 development of COPD[28]. Further research may include the evaluation of the effectiveness and 266 safety of different types of immunomodulators and the determination of the optimal treatment 267 strategy. These potential mechanisms require further research to promote their application in disease prevention or treatment. Such research may include basic scientific research, such as mechanism 268 269 studies and drug development, as well as clinical research, such as clinical trials and epidemiological studies, to evaluate the effectiveness and safety of these mechanisms and determine the optimal 270 treatment strategies. Additionally, interdisciplinary cooperation is needed to promote joint efforts 271 from different fields of expertise and accelerate the application of these mechanisms in clinical 272 273 practice.

274 增加讨论参考:

"In addition to genetic factors, immune pathways play a significant role in the development of both
asthma and COPD. Recent studies have highlighted the importance of Th17 cells and the IL-31/IL33 axis in modulating chronic inflammation and immune responses, which are pivotal in the
pathogenesis of these diseases. Furthermore, factors such as vitamin D and the microbiome have
been shown to influence immune regulation, potentially impacting disease progression. For example,
recent work by Drożdżal et al. [29] discussed these mechanisms in the context of chronic immunemediated diseases, offering new insights into their therapeutic potential."

The innovation of this study lies in starting from the perspective of genetics, using Mendelian randomization methods to minimize the interference of confounding factors as much as possible, and exploring whether asthma directly affects the risk of COPD. This will help fill the gaps in existing research and provide new directions for future studies. Additionally, through multivariable MR analysis, we will explore the impact of other exposure factors related to COPD, thereby achieving a more comprehensive understanding of this complex disease relationship.

288 In summary, this study provides new genetic evidence and theoretical support for understanding the 289 relationship between asthma and COPD through comprehensive analysis from multiple perspectives and methods. This has positive implications for the formulation of preventive and therapeutic 290 291 strategies for related diseases and the selection of future genetic research directions. In this study, there are limitations that need to be considered. Firstly, despite our efforts to screen 292 publicly available GWAS data rigorously, the limited nature of the data prevents us from conducting 293 294 detailed analyses on more disease pathological types. Particularly, due to the scarcity of patients and 295 the lack of specific SNP data, we cannot confirm the relationship between asthma and certain specific pathological types of COPD. Additionally, due to the diversity of "exposure" sources and 296 297 the lack of relevant SNP data, we have yet to investigate whether specific sources of "exposure" 298 exert similar biological effects on diseases and whether "exposure" has a consistent impact across different subgroups. It is worth noting that our data primarily consist of samples of European 299 300 ancestry, and thus, may not adequately differentiate ethnic variations from various regions, 301 potentially limiting the generalizability of these results to a global scale. Although the data from 302 GWAS or large samples used for diseases are continuously increasing, there still exists a problem 303 of data insufficiency. Further exploration of the basic multicenter research mechanisms between 304 "exposure" and "outcome" is needed to better understand how asthma affects the occurrence and 305 development of COPD.

306

307 Conclusion

308 Our research findings suggest that there is a significant causal relationship between asthma and

309	COPD, indicating that the occurrence of asthma elevates the likelihood of developing COPD in
310	patients. Even when considering potential confounding factors, asthma still exerts a significant
311	causal effect on COPD. These findings deepen our understanding the association between COPD
312	and asthma providing important reference for further research and clinical practice. Future studies
313	should delve deeper into the complex relationship between these two diseases to promote the
314	formulation of prevention and treatment strategies for respiratory system diseases.
315	

316	Abbrevi	ations
317	COPD	chronic obstructive pulmonary disease
318	GWAS	enome-wide association study
319	MR	Mendelian randomization
320	ACOS	asthma-COPD overlap syndrome
321	SNP	Single nucleotide polymorphisms
322	IVW	Inverse-variance weighted
323	MVMR	Multivariable mendelian randomization
324	OR	Odds ratio
325	CI	Confidence interval
326		
327	Ethics a	pproval and consent to participate
328	Not appl	icable.
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330 Consent for publication

331	Not applicable	э.
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333 Availability of data and materials

- 334 Data and material are available from (https://www.eqtlgen.org/) and corresponding GWAS
- 335 consortium.
- 336

337 Competing interests

- 338 The authors declared that they have no conflict of interest.
- 339

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- 346

347 Authors' Contributions

- 348 WS and YH were responsible for study conception and design. ZL and HS were in charge of data
- 349 collection and analysis and wrote the paper. All authors read and approved the final manuscript.

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351 Reference

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432 Figure legends



Figure 1. Mendelian randomization analysis flowchart. (a) Basic assumptions of Mendelian 434 435 randomization analysis, including (1) the relevance assumption, which states that selected instrumental variables must be significantly associated with the exposure factor; (2) the exclusion 436 restriction assumption, which states that instrumental variables could only affect the outcome 437 through the pathway "instrumental variable \rightarrow exposure \rightarrow outcome"; (3) the independence 438 assumption, which states that instrumental variables must not be significantly correlated with 439 potential confounding factors that might affect exposure or outcome. (b) Flowchart of the analysis 440 441 methods used in this study. SNP, single nucleotide polymorphism. IVW, inverse variance weighted. 442 MR-Egger, Mendelian randomization-Egger. GWAS, genome-wide association study.

Exp	oosure	method	β	OR95CI			P value
Asthma (ebi-a-	GCST90014325)	Inverse variance weighted	0.3671981	1.44 (1.36, 1.53)		H - H	9.99e-35
Asthma (ebi-a-	GCST90014325)	MR Egger	0.3407018	1.41 (1.21, 1.63)		— •—	3.12e-08
Asthma (ebi-a-	GCST90014325)	Simple mode	0.2518579	1.29 (1.06, 1.56)		—	0.0128
Asthma (ebi-a-	GCST90014325)	Weighted median	0.3698692	1.45 (1.33, 1.57)		— •—	7.18e-18
Asthma (ebi-a-	GCST90014325)	Weighted mode	0.3482749	1.42 (1.23, 1.63)	i	— •—	7.78e-06
Asthma (ebi-a	-GCST006862)	Inverse variance weighted	0.3022835	1.35 (1.25, 1.46)	i	——	4.72e-15
Asthma (ebi-a	-GCST006862)	MR Egger	0.2120904	1.24 (0.95, 1.61)	Ļ,		0.134
Asthma (ebi-a	-GCST006862)	Simple mode	0.2493573	1.28 (1.09, 1.51)		—	0.00841
Asthma (ebi-a	-GCST006862)	Weighted median	0.2704959	1.31 (1.20, 1.43)	i	—	1.97e-09
Asthma (ebi-a	-GCST006862)	Weighted mode	0.2546926	1.29 (1.13, 1.47)	i	—	0.00149



446 Figure 2. Multiple model analysis results of mendelian randomization analysis for asthma and 447 chronic obstructive pulmonary disease. (a) Forest plot showing the causal association analysis results of Mendelian randomization using multiple models for asthma and chronic obstructive 448 449 COPD. Effect estimates are presented using OR and 95% CI, along with the number of instrumental variables used in each model, as well as the calculated beta values and standard errors. B-C. Scatter 450 plots showing the causal relationship between Asthma (id: ebi-a-GCST006862) (b), Asthma (id: ebi-451 452 a-GCST90014325) (c), and chronic obstructive pulmonary disease. The slope of the line represents 453 the magnitude of the causal relationship predicted by different models. SNPs, Single Nucleotide Polymorphisms; OR, Odds Ratio; CI, Confidence Interval. 454



Figure 3. Funnel plot for heterogeneity testing and effect estimates of IVW random effects
model in mendelian randomization analysis of asthma and chronic obstructive pulmonary
disease. (a) Funnel plot displaying the causal effect estimates of each instrumental variable for
asthma and chronic obstructive pulmonary disease. The causal effect estimates of the IVW and MR
Egger models are indicated by lines on the plot. (b) Forest plot showing the sequential effect
estimates of asthma and COPD using single SNP locus analysis.

Exposure	method	β	OR95CI		P value
Chronic obstructive pulmonary disease	Inverse variance weighted	0.055008497	1.06 (0.96, 1.16)	H-	0.235
Chronic obstructive pulmonary disease	MR Egger	-0. 136972222	0.87 (0.73, 1.05)		0. 179
Chronic obstructive pulmonary disease	Simple mode	0.017520151	1.02 (0.93, 1.11)	н <mark>е</mark> н	0.698
Chronic obstructive pulmonary disease	Weighted median	0.003900208	1.00 (0.95, 1.06)	⊢∳-1	0.888
Chronic obstructive pulmonary disease	Weighted mode	0.004347437	1.00 (0.96, 1.05)	н і і	0.866
				0.8 0.9 1.0 1.1	



468 errors. SNPs, Single Nucleotide Polymorphisms; OR, Odds Ratio; CI, Confidence Interval.

469

470 Table 1 Estimates of Mendelian randomized causal effects of Asthma and Chronic obstructive

Evene	Mathad	Number	Q	Standar	Р	OP
Exposure	Methed	of SNPs	р	d error	value	0K
Asthma (eb	-a- Inverse	65	0.3671	0.02987	9.99E	1.44
GCST90014325)	variance		98137	265	-35	(1.36,
	weighted					1.53)
Asthma (eb	-a- MR Egger	65	0.3407	0.07590	3.12E	1.41
GCST90014325)			01751	4004	-05	(1.21,
						1.63)
Asthma (eb	-a- Simple mode	65	0.2518	0.09838	0.012	1.29
GCST90014325)			57948	223	8	(1.06,
						1.56)
Asthma (eb	-a- Weighted	65	0.3698	0.04294	7.18E	1.45
GCST90014325)	median		69234	8228	-18	(1.33,
						1.57)
Asthma (eb	-a- Weighted	65	0.3482	0.07158	7.78E	1.42
GCST90014325)	mode		74863	0068	-06	(1.23,
						1.63)
Asthma (eb	-a- Inverse	18	0.3022	0.03858	4.72E	1.35

471 pulmonary disease .

GCST006862)		variance		83474	5381	-15	(1.25,
		weighted					1.46)
Asthma (ebi-a-	MR Egger	18	0.2120	0.13430	0.134	1.24
GCST006862)				9039	7392		(0.95,
							1.61)
Asthma (ebi-a-	Simple mode	18	0.2493	0.08367	0.008	1.28
GCST006862)				573	6473	41	(1.09,
							1.51)
Asthma (ebi-a-	Weighted	18	0.2704	0.04507	1.97E	1.51) 1.31
Asthma (GCST006862)	ebi-a-	Weighted median	18	0.2704 95937	0.04507 9563	1.97E -09	1.51) 1.31 (1.20,
Asthma (GCST006862)	ebi-a-	Weighted median	18	0.2704 95937	0.04507 9563	1.97E -09	 1.51) 1.31 (1.20, 1.43)
Asthma (GCST006862) Asthma (ebi-a- ebi-a-	Weighted median Weighted	18	0.2704 95937 0.2546	0.04507 9563 0.06737	1.97E -09 0.001	 1.51) 1.31 (1.20, 1.43) 1.29
Asthma (GCST006862) Asthma (GCST006862)	ebi-a- ebi-a-	Weighted median Weighted mode	18	0.2704 95937 0.2546 92559	0.04507 9563 0.06737 5612	1.97E -09 0.001 49	 1.51) 1.31 (1.20, 1.43) 1.29 (1.13,

472 SNP, Single Nucleotide Polymorphism; OR, Odds Ratio; CI, Confidence Interval.

473

474 Table 2 Mendelian randomization analysis heterogeneity test for the association between Asthma

Evnogura	0	0 df	Cochran Q p-	I2
Exposure	Q	Qui	value	(%)
Asthma (ebi-a-GCST006862)	34.45016675	17	0.007340785	50.65
Asthma (ebi-a-GCST90014325)	80.35204633	64	0.081408939	20.35

475 and Chronic obstructive pulmonary disease

477 Table 3 Mendelian randomization analysis of horizontal pleiotropy for the association between

	-	-					
		MR-E	MR-Egger intercept		ď	P value	
Exposure		interce					
Asthma (ebi-a-GCS	ST90014325)	0.0019	009126	0.00502	21459	0.70	508059
Table 4 Mendelian ra	andomization analysi	s of Asthma on	Chronic	obstruct	ive pul	mona	ry disea
Steiger directionality	test						
					Corre	ct	
P		SNP r2	SNP	r2	causa	1	Steiger
Exposure	Outcome	exposure	outcor	ne	direct	io	value
					n		
Asthma (ebi-a-	Chronic	0.010916669	0.0005	575918	TRUE	(T)	0
GCST90014325)	obstructive						
	pulmonary disease						

478 Asthma and Chronic obstructive pulmonary disease

482 SNP, Single Nucleotide Polymorphism; r2, 方差解释率。

483

484 Table 5 Results of reverse causal Mendelian randomization analysis of Chronic obstructive

485 pulmonary disease and Asthma.

		Metho	Numb	Standard	
Exposure	Outcome		l	β	P value
_		d	er of	error	

			SNPs			
Chronic obstructive	Asthma (ebi-	MR	10	-	0.092977	0.178926
pulmonary disease	a-	Egger		0.136972	815	771
	GCST900143			222		
	25)					
Chronic obstructive	Asthma (ebi-	Weight	10	0.003900	0.027606	0.887648
pulmonary disease	a-	ed		208	237	78
	GCST900143	median				
	25)					
Chronic obstructive	Asthma (ebi-	Inverse	10	0.055008	0.046306	0.234860
pulmonary disease	a-	varianc		497	146	669
	GCST900143	e				
	25)	weight				
		ed				
Chronic obstructive	Asthma (ebi-	Simple	10	0.017520	0.043791	0.698426
pulmonary disease	a-	mode		151	52	597
	GCST900143					
	25)					
Chronic obstructive	Asthma (ebi-	Weight	10	0.004347	0.025058	0.866105
pulmonary disease	a-	ed		437	646	254
	GCST900143	mode				
	25)					

488 Table 6 The results of multivariable Mendelian randomization analysis on the impact of Asthma and

Model	Exposure	Outcome		nSNP	Beta	SE	Pvalue	OR (95%CI)
Model	Eosinophil	Chronic	obstructive	398	0.037394501	0.03433414	0.276094521	1.038(1.110,0.971)
1	cell count	pulmonary disea	se					
	Asthma	Chronic	obstructive	34	0.362836557	0.036274363	1.49E-23	1.437(1.339,1.543)
		pulmonary disea	se					
Model	Past tobacco	Chronic	obstructive	77	-0.737199808	0.087208236	2.83E-17	0.478(0.403,0.568)
2	smoking	pulmonary disea	se					
	Asthma	Chronic	obstructive	64	0.37334249	0.029146487	1.46E-37	1.453(1.372,1.538)
		pulmonary disea	se					
Model	Falling risk	Chronic	obstructive	2	0.117250627	0.228771877	0.608285706	1.124(0.718,1.761)
3		pulmonary disea	ise					
	Asthma	Chronic	obstructive	70	0.366839717	0.028134419	7.36E-39	1.443(1.366,1.525)
		pulmonary disea	se					

489 Chronic obstructive pulmonary disease

Model, Multivariable MR analysis of Asthma and Eosinophil cell count on Chronic obstructive
pulmonary disease; Model 2, Multivariable MR analysis of Asthma and Past tobacco smoking on
Chronic obstructive pulmonary disease; Model 3, Multivariable MR analysis of Asthma and Falling
risk on Chronic obstructive pulmonary disease. MR, Mendelian Randomization; SNPs: Single
Nucleotide Polymorphisms.



		Number	0	Standar	Р	OP
Exposure	Methed	of SNPs	β	d error	value	OR
Asthma (ebi-a-	Inverse	65	0.3671	0.02987	9.99E	1.44
GCST90014325)	variance		98137	265	-35	(1.36,
	weighted					1.53)
Asthma (ebi-a-	MR Egger	65	0.3407	0.07590	3.12E	1.41
GCST90014325)			01751	4004	-05	(1.21,
						1.63)
Asthma (ebi-a-	Simple mode	65	0.2518	0.09838	0.012	1.29
GCST90014325)			57948	223	8	(1.06,
						1.56)
Asthma (ebi-a-	Weighted	65	0.3698	0.04294	7.18E	1.45
GCST90014325)	median		69234	8228	-18	(1.33,
						1.57)
Asthma (ebi-a-	Weighted	65	0.3482	0.07158	7.78E	1.42
GCST90014325)	mode		74863	0068	-06	(1.23,
						1.63)
Asthma (ebi-a-	Inverse	18	0.3022	0.03858	4.72E	1.35
GCST006862)	variance		83474	5381	-15	(1.25,
	weighted					1.46)
Asthma (ebi-a-	MR Egger	18	0.2120	0.13430	0.134	1.24
GCST006862)			9039	7392		(0.95,
						1.61)
Asthma (ebi-a-	Simple mode	18	0.2493	0.08367	0.008	1.28
GCST006862)			573	6473	41	(1.09,
						1.51)

 Table 1 Estimates of Mendelian randomized causal effects of Asthma and Chronic obstructive

 pulmonary disease .

Asthma	(ebi-a-	Weighted	18	0.2704	0.04507	1.97E	1.31
GCST006862)			median		95937	9563	-09	(1.20,
								1.43)
Asthma	(ebi-a-	Weighted	18	0.2546	0.06737	0.001	1.29
GCST006862	2)		mode		92559	5612	49	(1.13,

SNP, Single Nucleotide Polymorphism; OR, Odds Ratio; CI, Confidence Interval.

Europura	0	O df	Cochran	$I^{2}(0/)$
Exposure	Q	Qui	Q p-value	1-(%)
Asthma (ebi-a-GCST006862)	34.45016675	17	0.007340785	50.65
Asthma (ebi-a-GCST90014325)	80.35204633	64	0.081408939	20.35

 Table 2 Mendelian randomization analysis heterogeneity test for the association between

 Asthma and Chronic obstructive pulmonary disease

Table 3 Mendelian randomization analysis of horizontal pleiotropy for the association between

Exposure	MR-Egger	Standard	P value	
Exposure	intercept	error		
Asthma (ebi-a-GCST90014325)	0.001909126	0.005021459	0.705080595	

Asthma and Chronic obstructive pulmonary disease

				Correct	
Exposure	Outcome	SNP r ²	SNP r ²	causal	Steiger
		exposure	outcome	directio	P value
				n	
Asthma (ebi-a-	Chronic	0.010916669	0.000575918	TRUE	0
GCST90014325	obstructive				
)	pulmonary disease				

 Table 4 Mendelian randomization analysis of Asthma on Chronic obstructive pulmonary

 disease Steiger directionality test

SNP, Single Nucleotide Polymorphism; r², 方差解释率。

Exposure	Outcome	Metho d	Numb er of SNPs	β	Standard error	P value
Chronic obstructive	Asthma (ebi-	MR	10	-	0.092977	0.178926
pulmonary disease	a-	Egger		0.136972	815	771
	GCST900143			222		
	25)					
Chronic obstructive	Asthma (ebi-	Weight	10	0.003900	0.027606	0.887648
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		ed				
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pulmonary disease	a-	ed		437	646	254
	GCST900143	mode				
	25)					

 Table 5 Results of reverse causal Mendelian randomization analysis of Chronic obstructive

 pulmonary disease and Asthma.

SNP, Single Nucleotide Polymorphism $_{\circ}$

Model	Exposure	Outcome	nSN P	Beta	SE	Pvalue	OR (95%CI)
Model	Eosinophil cell count	Chronic obstructive pulmonary	398	0.037394501	0.03433414	0.276094521	1.038(1.110,0.971)
1		disease					
	Asthma	Chronic obstructive pulmonary	34	0.362836557	0.036274363	1.49E-23	1.437(1.339,1.543)
		disease					
Model	Past tobacco smoking	Chronic obstructive pulmonary	77	-0.737199808	0.087208236	2.83E-17	0.478(0.403,0.568)
2		disease					
	Asthma	Chronic obstructive pulmonary	64	0.37334249	0.029146487	1.46E-37	1.453(1.372,1.538)
		disease					
Model	Falling risk	Chronic obstructive pulmonary	2	0.117250627	0.228771877	0.608285706	1.124(0.718,1.761)
3		disease					
	Asthma	Chronic obstructive pulmonary	70	0.366839717	0.028134419	7.36E-39	1.443(1.366,1.525)
		disease					

Table 6 The results of multivariable Mendelian randomization analysis on the impact of Asthma and Chronic obstructive pulmonary disease

Model, Multivariable MR analysis of Asthma and Eosinophil cell count on Chronic obstructive pulmonary disease; Model 2, Multivariable MR analysis of Asthma and Past tobacco smoking on Chronic obstructive pulmonary disease; Model 3, Multivariable MR analysis of Asthma and Falling risk on Chronic obstructive pulmonary

disease. MR, Mendelian Randomization; SNPs: Single Nucleotide Polymorphisms.



а	Exposure	method	β	OR95CI		P value
-	Asthma (ebi-a-GCST90014325)	Inverse variance weighted	0.3671981	1.44 (1.36, 1.53)	. ⊢ ⊷	9.99e-35
	Asthma (ebi-a-GCST90014325)	MR Egger	0.3407018	1.41 (1.21, 1.63)	¦	3.12e-05
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	Asthma (ebi-a-GCST90014325)	Weighted median	0.3698692	1.45 (1.33, 1.57)	¦ ⊢•→	7.18e-18
	Asthma (ebi-a-GCST90014325)	Weighted mode	0.3482749	1.42 (1.23, 1.63)	¦ ⊢-•i	7.78e-06
	Asthma (ebi-a-GCST006862)	Inverse variance weighted	0.3022835	1.35 (1.25, 1.46)	⊢ •−1	4.72e-15
	Asthma (ebi-a-GCST006862)	MR Egger	0.2120904	1.24 (0.95, 1.61)	·······	0.134
	Asthma (ebi-a-GCST006862)	Simple mode	0.2493573	1.28 (1.09, 1.51)	·	0.00841
	Asthma (ebi-a-GCST006862)	Weighted median	0.2704959	1.31 (1.20, 1.43)	—	1.97e-09
	Asthma (ebi-a-GCST006862)	Weighted mode	0.2546926	1.29 (1.13, 1.47)	·	0.00149
-					1.1 1.3 1.5	





Exposure	method	β	OR95CI		P value
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Chronic obstructive pulmonary disease	MR Egger	-0.136972222	0.87 (0.73, 1.05)		0. 179
Chronic obstructive pulmonary disease	Simple mode	0.017520151	1.02 (0.93, 1.11)	F F	0.698
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Chronic obstructive pulmonary disease	Weighted mode	0.004347437	1.00 (0.96, 1.05)	н і і	0.866
				08 09 10 11	