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# Causal effect of iron status on lung function: A Mendelian randomization study

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**Background:** The association between systemic iron status and lung function was conflicting in observational studies. We aim to explore the potential causal relationships between iron status and the levels of lung function using the two-sample Mendelian randomization (MR) design.

**Methods:** Genetic instruments associated with iron status biomarkers were retrieved from the Genetics of Iron Status (GIS) consortium (N = 48,972). Summary statistics of these genetic instruments with lung function were extracted from a meta-analysis of UK Biobank and SpiroMeta consortium (N = 400,102). The main analyses were performed using the inverse-variance weighted method, and complemented by multiple sensitivity analyses.

**Results:** Based on conservative genetic instruments, MR analyses showed that genetically predicted higher iron (beta: 0.036 per 1 SD increase, 95% confidence interval (CI): 0.016 to 0.056,  $P = 3.51 \times 10^{-4}$ ), log10-transformed ferritin (beta: 0.081, 95% CI: 0.047 to 0.116,  $P = 4.11 \times 10^{-6}$ ), and transferrin saturation (beta: 0.027, 95% CI: 0.015 to 0.038,  $P = 1.09 \times 10^{-5}$ ) were associated with increased forced expiratory volume in 1s (FEV1), whereas higher transferrin was associated with decreased FEV1 (beta: -0.036, 95% CI: -0.064 to -0.008, P = 0.01). A significant positive association between iron status and forced vital capacity (FVC) was also observed. However, there is no causal association between iron status and FEV1-to-FVC ratio (P = 0.10). Similar results were obtained from the liberal instruments analyses and multiple sensitivity analyses.

**Conclusion:** Our study provided strong evidence to support that higher iron status is causally associated with higher levels of FEV1 and FVC, but has no impact on airway obstruction, confirming iron status as an important target for lung function management.

#### KEYWORDS

Mendelian randomization, lung function, FEV1, FVC, iron

# Introduction

Chronic obstructive pulmonary disease (COPD) is a common disorder worldwide that has caused heavy disease burden (1). The global prevalence of COPD was estimated at  ${\sim}174$  million cases based on the Global Burden of Disease Study 2015 (2) and the proportion can reach 28% in populations aged 80 years (3). Persistent decline in lung function as assessed by forced expiratory volume in 1 second (FEV1) underlies the pathogenesis of COPD (4). It has been long considered that exposure to particles and gases from tobacco smoking and biomass fuel are the major etiology of lung function decline. However, emerging evidence supports the notion that events early in life affect lung function in adults (5). This fundamentally changed the concept of COPD to show that factors determining the maximal lung capacity attained during development may play important roles in the pathogenesis of the disease (5). Seeking potential risk factors that are involved in lung function will assist in preventing the deterioration of lung function and associated death.

As an essential trace element, iron participates in various aspects of lung diseases, including oxygen delivery, immune response, and oxidative stress (6). Iron has been reported to be critical for effective immune response in defending respiratory tract infection (7). Besides, the early life iron status may determine the maximum lung function by influencing the development of airways and lungs. Current evidence on the association between iron status and lung function is conflicting. An observational study conducted in US women aged 20-49 years in National Health and Nutrition Examination Survey (NHANES) reported that total body iron status was positively associated with FEV1 (8). Similar results were achieved in children aged 10-12 years, where participants with lower iron status have decreased lung function (9). However, these findings contrast with the results of a large cross-sectional study performed in 42,927 Korean men that suggests a negative association between ferritin level and lung function as indicated by FEV1 and forced vital capacity (FVC) (10). In addition, serum ferritin has been associated with increased risk of restrictive respiratory disease in postmenopausal women (11).

Making causal inference from relationships that may be attributable to confounding in these observational studies is challenging. Recently, Mendelian randomization (MR) has emerged as a reliable method to estimate causal association between exposure and outcome by using genetic variants as proxy for exposure (12). Given the non-modifiable and unconfounded nature of genetic instruments, MR study can largely overcome these issues in traditional observational studies. The MR design has been previously adopted to study the causal associations of iron status with risk of asthma and lung cancer (13, 14). However, the causal nature between systemic iron status and lung function remains to be investigated. The aim of the present study was to explore the association of genetically instrumented iron status with lung function using the MR approach.

## **Methods**

#### Study design

Figure 1 depicted the overall design of the present study. The causal association between iron status and lung function was examined in a two-sample MR design using genome-wide association study (GWAS) summary-level data. Three crucial assumptions should be hold for a valid genetic instrument: (1) it is robustly associated with exposure of interest; (2) it is not associated with potential confounders; and (3) it should exert no effect on the outcome through factors other than exposure, i.e., no horizontal pleiotropy (15). We extracted single-nucleotide polymorphism (SNP)-iron status associations from GWAS of iron status (16). SNP-lung function associations were retrieved from a meta-analysis of UK Biobank and SpiroMeta consortium (17). After data harmonization, MR effect estimates were calculated followed by multiple sensitivity analyses to test the robustness of the results. The associations of iron status with potential confounding factors were also examined.

We performed this MR study as per the Strengthening the Reporting of Observational Studies in Epidemiology-Mendelian randomization (STROBE-MR) guidelines (18). All datasets used in the present study are publicly available from each original studies, where patient consent and ethical approval had been obtained. No additional informed consent and ethical approval are required.

### Data sources for iron status

Genetic instruments associated with four serum markers of iron status, i.e., iron ( $\mu$ mol/l), ferritin (log<sub>10</sub>-transformed, ug/l), transferrin saturation (%), and transferrin (g/l) were obtained from the Genetics of Iron Status (GIS) consortium, the largest GWAS on iron status to date (16). The GIS consortium identified genetic loci associated with iron status by meta-analyzing results from 11 cohorts among 23,986 participants of European descent in its discovery phase, followed by replication in up to 24,986 participants of European descent (16). The mean age of participants was 46.89  $\pm$  17.84 years, and 45% of the total participants were men. Genetic associations between SNPs and systemic iron markers were adjusted for age, principal component scores and other study specific covariates (16).



#### Genetic instruments selection

SNPs showed significant associations ( $P < 5 \times 10^{-8}$ ) with iron status were extracted from GIS consortium (Supplementary Table S1). A total of 5, 6, 5, and 9 SNPs were identified as genetic instruments for serum iron, ferritin, transferrin saturation, and transferrin, respectively (Supplementary Table S2). Then, SNPs with linkage disequilibrium (LD) were clumped based on 1000 Genomes LD reference panel (European population) using the TwoSampleMR package of R. All of the SNPs were in linkage equilibrium ( $r^2 \leq 0.01$  and window size = 10 MB). One SNP, rs1799945, was missing from the FEV1 dataset without suitable proxy SNP ( $r^2 > 0.8$ ) was excluded. Finally, exposure and outcome SNPs were harmonized to align alleles on the forward strand. Palindromic SNPs were also omitted prior to further MR analysis.

We selected two sets of genetic instruments for iron status: (1) Conservative genetic instruments, which comprised 3 core SNPs (rs855791, rs1799945, rs1800562; Supplementary Table S1) associated with elevated serum iron, ferritin and transferrin saturation, but with decreased serum transferrin level at genome-wide significance ( $P < 5 \times 10^{-8}$ ). These 3 SNPs were used for the main analyses, since systemic iron status was positively associated with serum iron, ferritin and transferrin level (19). (2) Liberal genetic instruments, where SNPs that were associated with at least one of the four

iron biomarkers at genome-wide significance level ( $P < 5 \times 10^{-8}$ ) were included (Supplementary Table S2) (16). The liberal genetic instruments contained more SNPs than conservative genetic instruments and were expected to increase the statistical power, but at the expense of increased pleiotropic bias risk.

#### Data sources for lung function

3 spirometric indices of lung function were considered in current study: FEV1, FVC, and FEV1-to-FVC Ratio (FEV1/FVC). Summary-level statistics for lung function were obtained from a meta-analysis of UK Biobank (n = 321,047) and SpiroMeta consortium (n = 79,055) with up to 400,102 participants of European ancestry (17). UK Biobank is a population-based cohort study that comprises participants aged between 40 to 69 years (20). Only participants who had at least two measures of FEV1 and FVC, and complete information for age, sex, height, ever smoking status, and spirometry method used were included. Haplotype Reference Consortium was used as reference panel to impute genotypes, where age, sex, height, smoking status and genotyping array were adjusted (17). For SpiroMeta consortium, genotypes were imputed to either 1000 genomes project phase 1 panel or haplotype reference consortium panel. age, sex and height were adjusted in the model. Details of UK Biobank and SpiroMeta consortium have been described elsewhere (17, 20).

TABLE 1	Baseline characteristics of participants included in
the analy	sis.

Characteristics	UK Biobank	SpiroMeta study
Participants, n	321,047	79,055
Age, y; mean (SD)	56.44 (8.02)	49.67 (15.80)
Age at lung function measurement, y	39-72	14-99
Male (%)	178,489 (44.40)	35, 734 (45.20)
Height, cm; mean (SD)	168.57 (9.13)	168.27 (9.24)
FEV1, L; mean (SD)	2.84 (0.76)	2.98 (0.93)
FVC, L; mean (SD)	3.74 (0.96)	3.75 (1.12)
FEV1/FVC; mean (SD)	0.76 (0.06)	0.79 (0.10)
PEF, L/min; mean (SD)	406.19 (117.55)	NA
Never smokers (%)	173,658 (54.09)	37,939 (47.99)
Ever smokers (%)	147,389 (45.91)	41,116 (52.01)

 ${\rm FEV1},$  forced expired volume in 1 sec; FVC, forced vital capacity; PEF, Peak expiratory flow.

Baseline characteristics of the 400,102 participants of European ancestry included for lung function analysis are shown in the Table 1. The mean age of participants was 55.10 years, and 53.54% were men. The mean FEV1, FVC, FEV1/FVC, and PEF were 2.86, 3.74, 0.76, 406.19 L/min, respectively. Of the participants, 188,505 (47.11%) were ever smoker, and 211,597 (52.89%) were never-smokers.

#### Mendelian randomization analysis

After data harmonization, Wald ratio for individual SNP was calculated by SNP-outcome coefficient divided by SNP-exposure coefficient. Random- or fixed-effect inverse-variance weighted (IVW) method was applied to combine the Wald ratio estimates to give an overall estimate of the causal effect across all SNPs included for each iron status. This method can provide unbiased causal estimates when instrumental variables are valid and pleiotropy is absent or balanced (21). To evaluate the heterogeneity in MR analysis, Cochran's Q statistic and corresponding *P*-value for the IVW method were calculated. When significant heterogeneity was observed (P < 0.05), the random-effect IVW method was applied to pool the Wald ratio estimates; otherwise, the fixed-effect IVW model was used (22).

#### Sensitivity analysis

MR estimates can be influenced by invalid instrument bias or potential pleiotropy. To test the validity and robustness of the MR results, we performed several sensitivity analyses: weighted median, MR-Egger regression, MR pleiotropy residual sum and outlier test (MR-PRESSO), and MR-robust adjusted profile scores (MR-RAPS) methods. Limited by the number of SNPs (n = 3), we can only conduct partial sensitivity analyses for the conservative instrumental variable analyses. The weighted median method can provide consistent effect estimates when more than 50% of the weight in the analysis was derived from valid genetic instruments (23). MR-Egger regression was used to check and adjust for potential directional pleiotropy. The value of intercept term significantly deviates from 0 (*P* for MR-Egger intercept < 0.05) suggests the existence of directional pleiotropy. Meanwhile, MR-Egger regression can provide pleiotropy-corrected effect estimates but with relatively low precision, particularly when the number of instrumental variables is small (24). The MR-PRESSO method was applied to detect horizontal pleiotropic outlying SNPs (25). It can also provide outlier-removed causal estimates. MR-RAPS is a method which accounts for weak instruments bias and reports pleiotropy-corrected causal effect using robust adjusted profile scores (21).

To further explore the possibility of pleiotropy, we search the online database PhenoScanner (http://www.phenoscanner. medschl.cam.ac.uk/phenoscanner) for potential secondary phenotypes associated with SNPs that were used as genetic instruments for iron status (26, 27). Two lung functionassociated risk factors were considered in our analyses: body mass index (BMI) and tobacco smoking. All statistical analyses were performed using R (version 3.6.1, R Core Team, Vienna, Austria) with TwoSampleMR, MR pleiotropy residual sum and outlier, MR-robust adjusted profile scores packages. Results are reported as beta with corresponding 95% confidence interval [CI] per SD unit increase in each iron status biomarker. A 2-sided *P*-value < 0.05 was set as the threshold of statistical significance.

### Results

### Genetic instruments for iron status

Conservative and liberal genetic instruments for iron status that were used in MR analyses are shown in Supplementary Tables S1, S2, respectively. Briefly, the conservative genetic instruments contain 3 independent SNPs that have a concordant association with all of the 4 iron-related phenotypes: rs1800562, rs1799945, and rs855791 (Supplementary Table S1). For liberal genetic instruments, a total of 5, 6, 5 and 9 independent SNPs were selected for serum iron, ferritin, transferrin saturation, and transferrin, respectively (Supplementary Table S2). The total variances of iron status explained by genetic instruments were 3.4% for iron, 0.9% for (log<sub>10</sub>) ferritin, 6.7% for transferrin saturation, and 7.2% for transferrin (16). The F statistics for included SNPs ranged from 31 to 2947, indicating weak instrument bias was unlikely

Lung Function	SNPs			Beta [95% CI]	P-value
FEV1					
Ferritin (log10)	3		<b>⊢</b> ●	0.081 [0.047, 0.116]	4.10E-06
Iron	3		<b>⊢</b> −−−1	0.036 [0.016, 0.056]	3.51E-04
Transferrin saturation	3		<b>⊢</b> ●–-i	0.027 [0.015, 0.038]	1.09E-05
Transferrin	3	<b>⊢</b>		-0.036 [-0.064,-0.008]	0.01
FVC					
Ferritin (log10)	3		<b>⊢</b> •−-i	0.039 [0.023, 0.055]	9.8E-07
Iron	3		<b>⊢</b> •−-i	0.039 [0.023, 0.055]	9.8E-07
Transferrin saturation	3		H●H	0.029 [0.024, 0.033]	1.2E-34
Transferrin	3	<b>⊢</b>		-0.039 [-0.061,-0.018]	2.76E-04
FEV1/FVC					
Ferritin (log10)	3	H	• 1	0.006 [-0.047, 0.060]	0.81
Iron	3	i	•i	0.003 [-0.022, 0.027]	0.82
Transferrin saturation	3	<u> </u>	•I	0.002 [-0.016, 0.019]	0.84
	2	<b>—</b>		-0.001 [-0.027 0.025]	0.92

The IVW MR estimates for relationship between iron status and lung function using conservative instrumental variables associated with all four iron biomarkers. The causal effects of iron status on levels of FEV1, FVC, and FEV1/FVC ratio are estimated. Results are reported as beta with corresponding 95% CI per SD unit increase in each iron marker. IVW, inverse variance weighted; MR, Mendelian randomization; FEV1, forced expiratory volume in 1 s; FVC, forced vital capacity.

(Supplementary Table S2). We summarized the associations of genetic instruments with iron status and spirometric indices in Supplementary Tables S4, S5, respectively.

# Causal association of iron status with lung function

The main IVW findings between iron status and lung function (FEV1, FVC, and FEV1/FVC ratio) using conservative instrumental variables are shown in Figure 2. The results, reported as beta for spirometric indices per SD unit increase in the iron status marker, demonstrated that genetically predicted higher iron (beta = 0.036, 95% CI: 0.016 to 0.056, P = 3.51 $\times$  10<sup>-4</sup>), log<sub>10</sub>-transformed ferritin (beta = 0.081, 95% CI: 0.047 to 0.116,  $P = 4.11 \times 10^{-6}$ ), and transferrin saturation (beta = 0.027, 95% CI: 0.015 to 0.038,  $P = 1.09 \times 10^{-5}$ ) levels were associated with increased FEV1. Genetically instrumented higher transferrin level, which reflects lower iron status, was significantly associated with decreased FEV1 (beta = -0.036, 95% CI: -0.064 to -0.008, P = 0.01, Figure 2). Similarly, genetical predisposition to increased levels of iron (beta = 0.039, 95%CI: 0.023 to 0.055,  $P = 9.76 \times 10^{-7}$ ), log<sub>10</sub>-transformed ferritin (beta = 0.039, 95% CI: 0.023 to 0.055,  $P = 9.76 \times 10^{-7}$ ), and transferrin saturation (beta = 0.029, 95% CI: 0.024 to 0.033,  $P = 1.23 \times 10^{-34}$ ) were causally associated with higher FVC,

whereas transferrin level was negatively associated with FVC (beta = -0.039, 95% CI: -0.061 to -0.018,  $P = 2.76 \times 10^{-4}$ ). No significant association between iron status and FEV1/FVC ratio was observed (Figure 2).

We further assessed the association between iron status and lung function based on liberal genetic instruments, which included more SNPs and were expected to have greater statistical power (Figure 3). Consistent results were obtained using the liberal genetic instruments, where genetically predicted iron status was positively associated with FEV1 and FVC. However, there is still no significant association between iron status and FEV1/FVC ratio using these expanded instrumental variables (Figure 3, Supplementary Table S7). None of the included SNPs exhibited any relationships with BMI or tobacco smoking (Supplementary Table S3).

#### Sensitivity analysis

Using conservative instrumental variable, the causal effects of iron status on lung function were robust and consistent in all sensitivity analyses except for MR-Egger, which has wider CIs (Table 2, Supplementary Table S6). There was no substantial heterogeneity between individual SNPs in the analyses (all *P*-values for heterogeneity were >0.05). Potential pleiotropies were indicated by MR-Egger regression for the analyses of serum

Lung Function	SNPs		Beta [95% CI]	P-value
FEV1				
Ferritin (log10)	6	<b>⊢</b>	⊣ 0.082 [0.019, 0.145]	0.011
Iron	4	<b>⊢</b> −−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−	0.035 [0.003, 0.067]	0.0347
Transferrin saturation	5	⊢●┥	0.028 [0.018, 0.038]	5.90E-08
Transferrin	9		-0.023 [-0.036,-0.009]	9.66E-04
FVC				
Ferritin (log10)	6	⊢ → → → → → → → → → → → → → → → → → → →	0.079 [0.020, 0.139]	8.75E-03
Iron	5	<b>⊢</b> I	0.031 [0.002, 0.060]	0.034
Transferrin saturation	5	⊢●┤	0.029 [0.020, 0.039]	3.93E-10
Transferrin	9		-0.021 [-0.035,-0.006]	0.005
FEV1/FEVC				
Ferritin (log10)	6		0.027 [-0.019, 0.073]	0.245
Iron	5	●	0.005 [-0.014, 0.024]	0.622
Transferrin saturation	5	<b>⊢</b>	0.003 [-0.011, 0.016]	0.718
Transferrin	9	<b>⊢</b> −● <mark> </mark> −1	-0.007 [-0.029, 0.014]	0.490
			0.15	

The IVW MR estimates for relationship between iron status and lung function using liberal instrumental variables associated with each iron biomarker. The causal effects of iron status on levels of FEV1, FVC, and FEV1/FVC ratio are estimated. Results are reported as beta with corresponding 95% CI per SD unit increase in each iron marker. IVW, inverse variance weighted; MR, Mendelian randomization; FEV1, forced expiratory volume in 1 s; FVC, forced vital capacity.

iron on FVC, and serum transferrin on FEV1/FVC ratio, as the intercepts were significantly departed from zero. However, the estimates from other MR methods—MR-Egger, weighted median, and MR-RAPS—were in the same direction as of those from the primary IVW analyses, providing support for the causal associations of iron status on FEV1 and FVC (Table 2). The effect estimates of each SNP on iron status and lung function outcomes are depicted in scatter plots (Supplementary Figures S1–S3). The forest plots are shown in Supplementary Figures S7–S9.

In liberal instruments analyses, significant heterogeneities were observed in most analyses (Supplementary Table S7). Despite this, the heterogeneity was unlikely to influence the results, as weighted median approach provided same direction estimates compared to the main IVW MR (Table 3, Supplementary Table S7). The MR-Egger regression identified directional pleiotropy in the analyses of iron on FVC (*P* for MR-Egger intercept = 0.023), and transferrin on FEV1/FVC ratio (*P* for MR-Egger intercept = 0.012, Supplementary Table S7). However, even after correcting for pleiotropy, MR analyses still showed a causal association between serum iron and FVC (beta = 0.037, 95%CI: 0.013 to 0.060, P = 0.002 in MR-RAPS), and no association between transferrin and FEV1/FVC ratio (beta = -0.006, 95%CI: -0.024 to 0.012, P = 0.506 in MR-RAPS; beta = -0.010, 95%CI: -0.040 to

0.019, P = 0.522 in MR-PRESSO). The scatter plots and forest plots displaying the estimates of each SNP on lung function outcomes are shown in Supplementary Figures S4–S6, Supplementary Figures S10–S12, respectively.

# Discussion

For the first time, the causal associations of genetically instrumented systemic iron status with lung function outcomes were detailly examined using the MR design. The present MR study went beyond previous work, to provide genetic causal estimates from larger studies of lung function. Our MR analyses demonstrated consistent evidence for associations between increased iron status and higher FEV1 and FVC levels, whenever using the only 3 SNPs associated with all four iron status biomarkers (conservative genetic instruments) or additional genetic instruments associated with each iron status (liberal genetic instruments) at genome-wide significance. Specifically, we identified 27 to 81 ml increase of FEV1, and 29 to 39 ml increase of FVC per 1 SD increase in each systemic iron status. However, no association between systemic iron biomarkers and FEV1-to-FVC ratio was found. Although pleiotropy was

Iron status	Outcome	Method	No. of SNPs	Beta (95% CI)	<i>P</i> -value	P for heterogeneity	P for intercept
Iron	FEV1	IVW MR	3	0.036 (0.016: 0.056)	3.51E-04	0.214	0.167
		Weighted median	3	0.040 (0.021; 0.058)	4.22E-05		
		MR-Egger	3	0.073 (-0.012; 0.159)	0.342		
		MR-RAPS	3	0.037 (0.017; 0.056)	2.47E-04		
Ferritin (log <sub>10</sub> )	FEV1	IVW MR	3	0.081 (0.047; 0.116)	4.11E-06	0.375	0.641
		Weighted median	3	0.082 (0.044; 0.120)	2.18E-05		
		MR-Egger	3	0.072 (-0.015; 0.158)	0.353		
		MR-RAPS	3	0.081 (0.039; 0.123)	1.37E-04		
Transferrin saturation	FEV1	IVW MR	3	0.027 (0.015; 0.038)	1.09E-05	0.344	0.763
		Weighted median	3	0.029 (0.016; 0.041)	9.80E-06		
		MR-Egger	3	0.027 (-0.009; 0.063)	0.384		
		MR-RAPS	3	0.027 (0.015; 0.038)	1.05E-05		
Transferrin	FEV1	IVW MR	3	-0.036 (-0.064; -0.008)	0.012	0.062	0.183
		Weighted median	3	-0.033 (-0.051; -0.015)	3.54E-04		
		MR-Egger	3	-0.023(-0.059; 0.013)	0.429		
		MR-RAPS	3	-0.034 (-0.052; -0.016)	1.83E-04		
Iron	FVC	IVW MR	3	0.039 (0.023; 0.055)	9.76E-07	0.397	0.023
		Weighted median	3	0.039 (0.023; 0.055)	9.76E-07		
		MR-Egger	3	0.081 (0.016; 0.146)	0.248		
		MR-RAPS	3	0.039 (0.022; 0.056)	5.51E-06		
Ferritin (log <sub>10</sub> )	FVC	IVW MR	3	0.039 (0.023; 0.055)	9.76E-07	0.878	0.849
		Weighted median	3	0.037 (0.017; 0.056)	2.05E-04		
		MR-Egger	3	0.081 (0.016; 0.146)	0.248		
		MR-RAPS	3	0.039 (0.022; 0.056)	5.51E-06		
Transferrin saturation	FVC	IVW MR	3	0.029 (0.024; 0.033)	1.23E-34	0.854	0.790
		Weighted median	3	0.030 (0.018; 0.043)	2.78E-06		
		MR-Egger	3	0.031 (0.006; 0.056)	0.251		
		MR-RAPS	3	0.029 (0.017; 0.041)	2.48E-06		
Transferrin	FVC	IVW MR	3	-0.039(-0.061; -0.018)	2.76E-04	0.207	0.653
		Weighted median	3	-0.037(-0.055; -0.020)	2.68E-05		
		MR-Egger	3	-0.027(-0.050; -0.005)	0.255		
		MR-RAPS	3	-0.039(-0.057; -0.021)	1.40E-05		

TABLE 2 Associations of genetically predicted iron status with FEV1 and FVC using the 3 SNPs associated with all 4 iron biomarkers.

The effect estimates were reported as beta coefficients and corresponding 95% CIs. CI, confidence interval; IVW, the inverse-variance weighted method; MR, Mendelian randomization; MR-RAPS, MR-robust adjusted profile scores; SNPs, single nucleotide polymorphisms.

observed in some analyses, the overall results were consistent using sensitivity analyses robust to pleiotropy.

The mechanisms underlying the association between iron status and lung function remain to be elucidated, but potentially include iron or iron-containing proteins promote lung and airway development. Iron is an important micronutrient element which has been implicated in various cellular processes such as DNA synthesis, oxygen transport, energy metabolism and mitochondrial respiration (28). Iron deficiency induced lower oxygen binding and inadequate oxygen supply could result in a growth retardation of lung during development (29). Hypoxic condition arising from iron deficiency redistributed the cardiac output to vital organs such as brain, heart and kidney, leading to reduced blood flow to the lung and restriction of lung growth (29). In animal experiments, iron chelation in lung buds from embryonic mice significantly reduced vascular branching and airway development (30). In support of these evidence, population-based investigations found that maternal iron status during pregnancy was positively associated with lung function levels in offspring at age of 10 years (31).

On the other hand, iron status may influence lung function *via* regulating immune system. Evidence from cohort studies showed that airway and lung damages caused by respiratory tract infections in early life increased the risk TABLE 3 Associations of genetically predicted iron status with FEV1 and FVC using the separately selected SNPs associated with each iron biomarker.

Iron status	Outcome	Method	No. of SNPs	Beta (95% CI)	P-value	P for heterogeneity	P for intercept
Iron	FEV1	IVW MR	4*	0.035 (0.003; 0.067)	0.03	0.012	0.167
		Weighted median	4*	0.042 (0.022; 0.061)	3.47E-05		
		MR-Egger	4*	0.072 (0.031; 0.112)	0.07		
		MR-RAPS	4*	0.040 (0.018; 0.062)	4.00E-04		
		MR-PRESSO	4*	0.035 (0.003; 0.067)	0.125		
Ferritin (log <sub>10</sub> )	FEV1	IVW MR	6	0.082 (0.019; 0.145)	0.010	2.76E-04	0.641
		Weighted median	6	0.064 (0.027; 0.101)	7.16E-04		
		MR-Egger	6	0.052 (-0.082; 0.187)	0.490		
		MR-RAPS	6	0.073 (0.028; 0.119)	0.001		
		MR-PRESSO	5†	0.065 (0.030; 0.100)	0.022		
Transferrin saturation	FEV1	IVW MR	5	0.028 (0.018; 0.038)	5.90E-08	0.512	0.763
		Weighted median	5	0.029 (0.016; 0.042)	7.33E-06		
		MR-Egger	5	0.025 (0.007; 0.044)	0.07		
		MR-RAPS	5	0.028 (0.016; 0.039)	1.97E-06		
		MR-PRESSO	5	0.028 (0.018; 0.038)	0.005		
Transferrin	FEV1	IVW MR	9	-0.023 (-0.036; -0.009)	9.66E-04	0.059	0.183
		Weighted median	9	-0.017(-0.029; -0.005)	0.005		
		MR-Egger	9	-0.014 (-0.031; 0.003)	0.15		
		MR-RAPS	9	-0.023 (-0.036; -0.011)	2.01E-04		
		MR-PRESSO	9	-0.023(-0.036; -0.009)	0.010		
Iron	FVC	IVW MR	5	0.031 (0.002; 0.060)	0.030	0.008	0.023
		Weighted median	5	0.034 (0.014; 0.054)	0.001		
		MR-Egger	5	0.077 (0.046; 0.107)	0.020		
		MR-RAPS	5	0.037 (0.013; 0.060)	0.002		
		MR-PRESSO	5	0.031 (0.002; 0.060)	0.101		
Ferritin (log <sub>10</sub> )	FVC	IVW MR	6	0.079 (0.020; 0.139)	0.008	0.001	0.849
		Weighted median	6	0.083 (0.043; 0.122)	4.20E-05		
		MR-Egger	6	0.068 (-0.060; 0.196)	0.360		
		MR-RAPS	6	0.076 (0.030; 0.122)	0.001		
		MR-PRESSO	5	0.065 (0.025; 0.105)	0.034		
Transferrin saturation	FVC	IVW MR	5	0.029 (0.020; 0.039)	3.93E-10	0.607	0.790
		Weighted median	5	0.030 (0.018; 0.043)	3.20E-06		
		MR-Egger	5	0.031 (0.014; 0.049)	0.040		
		MR-RAPS	5	0.030 (0.018; 0.041)	6.20E-07		
		MR-PRESSO	5	0.029 (0.020; 0.039)	0.003		
Transferrin	FVC	IVW MR	9	-0.021(-0.035; -0.006)	0.005	0.027	0.653
		Weighted median	9	-0.015(-0.028; -0.003)	0.020		
		MR-Egger	9	-0.024(-0.045; -0.003)	0.060		
		MR-RAPS	9	-0.024(-0.039;-0.009)	0.002		
		MR-PRESSO	9	-0.021(-0.035; -0.006)	0.024		

The effect estimates were reported as beta coefficients and corresponding 95% CIs. CI, confidence interval; IVW, the inverse-variance weighted method; MR, Mendelian randomization; MR-PRESSO, MR-Pleiotropy Residual Sum and Outlier; MR-RAPS, MR-robust adjusted profile scores; SNPs, single nucleotide polymorphisms. \*One SNP, rs1799945, was missing from FEV1 outcome dataset without suitable proxy SNP. <sup>†</sup> Outlying SNP detected: rs651007.

of FEV1 deficit in adult life (32). Iron homeostasis takes part in both innate and adaptive immunity during infections (33). In innate immunity, iron fine-tunes the function of myeloid cells through controlling the activity of enzymes and transcription factors and thus the production of antimicrobial effectors such as nitric oxide and hydroxyl radicals. Iron also plays crucial roles in adaptive immune response by regulating clonal expansion of lymphocyte subsets (34). However, iron overload or deficiency results in immune deficits may predispose organisms to respiratory infection and airway dysfunction. For instance, iron deficiency has been linked to higher airway inflammatory response and increased risk of allergic asthma (35, 36), both are involved in the pathogenesis of COPD. In contrast, iron supplement decreases the production of proinflammatory cytokines such as interleukin (IL)-1β, IL-9 and IL-17, and subsequently reduced airway hyper-responsiveness (37). Collectively, findings from animal studies and cell line studies provide biological evidence that iron homeostasis is beneficial for lung development, reducing airway hyper-responsiveness, and defending lung infections.

Previous observational studies reported conflicting associations between systemic iron status and lung function (8-10, 38). One study conducted in NHANES III which included participants aged above 20 years provided evidence that iron status was positively associated with FEV1 and FVC levels (39). However, another cohort study performed in 42,927 healthy Korean men suggested that higher serum ferritin, but not iron or transferrin saturation, was associated with decreased FEV1 and FVC (10). Some studies even reported null association between ferritin and lung function (8). Most studies to date linking iron status to poor lung function have co-existing risk factors for lung disease, making it difficult to rule out that lung function decline is not merely a result of confounders, rather than being a cause of iron deficiency. For example, tobacco smoking is a well-recognized risk factor for poor lung function, it can also increase the ferritin concentration in respiratory lavage (40) and serum (41), thus biasing the iron status-lung function associations in those studies. By employing two-sample MR method, we could overcome these defects of observational studies to provide strong evidence for a positive association between iron status and lung function. Moreover, the pooled MR estimates for serum iron, ferritin, transferrin saturation, and transferrin all showed that higher iron status improves FEV1 and FVC, but not FEV1/FVC ratio.

Taking advantages of the random assign of genetic variants at conception, the findings of MR studies can be treated similar to randomized controlled trails (RCTs). However, it is crucial to hold the three principal assumptions; in particular, the exclusion assumption (absence of pleiotropy), to provide unbiased causal evidence between exposure of interest and outcomes (42). To evaluate the violations of these assumptions, we searched the large-scale genetic association database PhenoScanner for potential associations of SNPs with confounders or traits that could influence lung function (26, 27). Based on prior studies, we considered BMI and tobacco smoking as potential confounding factors or sources of pleiotropy as both traits have been associated with lung function. As a result, we identified no associations of used SNPs with these phenotypes, as shown in Supplementary Table S3. However, we did not take asthma into account in our analyses, because if these SNPs have impacts on asthma, which in turn affect the lung function (vertical pleiotropy), then it still should be considered a causal factor for lung function.

There are several strengths in this MR analyses. To the best of our knowledge, this is the first MR study to examine the causal nature between systemic iron status and lung function. As genetic variants used to instrument the effect of modifying iron status were randomly allocated within participants, the frequently occurred bias in conventional epidemiological studies due to potential confounding, measurement error, and reverse causation can be largely avoided. The included genetic instruments all have strong associations with iron status (F statistics > 10) decreased weak instrument bias. Meanwhile, our analyses with large sample size (up to 400,102 participants) maximized the statistical power to detect robust causal associations. No participants overlap between exposure and outcome dataset also lowered the type 1 error rate. Moreover, multiple sensitivity analyses with different underlying assumptions (weighted median, MR-Egger, MR-PRESSO, and MR-RAPS) all provided similar and consistent estimates in our study, suggesting the credibility of our conclusion (21).

Several limitations should be considered in our work. First, the magnitude of causal estimates through genetic effect may be different from the magnitude of effect of randomized trials. This can be partly attributed to the difference in magnitude of genetic instruments and clinical interventions. In fact, MR analyses were typically designed to determine the causal association between exposure and clinical endpoints (43). Although MR study overcomes the bias from confounding, low precision in effect estimates has been deemed a tradeoff for the unconfounded association (44). Second, etiology of altered lung function is multi-factorial and many factors have been linked with lung function other than iron status. For instance, both FEV1 and FVC progressively decline with age after pulmonary maturity, and age has historically been one of the major factors in the evaluation of lung function. Gender, height and ethnic group have also been identified as contributing factors for lung function. However, as only summary-level statistics derived from general population of both genders were available in current MR study, we can hardly to conduct stratified analyses by gender, age, height and ethic group. The causal effects of iron status on lung function in these specialized populations may need further investigation. Third, all data used for genetic analyses were based on participants of European ancestry; therefore, the results are less generalizable to other populations or settings.

# Conclusion

In summary, the present MR study provides robust evidence to show that higher iron status is causally associated with higher levels of FEV1 and FVC, but has no effect on FEV1-to-FVC ratio. Our study highlights the role of systemic iron in influencing lung function among general population. Iron supplementation may be served as a therapeutic strategy for preventing the decline of lung function and associated complications in populations with iron deficiency. Further research is still needed to understand the underlying mechanisms between iron and lung function, as well as to determine the causal role of iron on lung function in specialized populations.

### Data availability statement

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

## **Ethics statement**

Ethical review and approval was not required for the study on human participants in accordance with the local legislation and institutional requirements. The patients/participants provided their written informed consent to participate in this study.

# Author contributions

Conceptualization: ZY, CX, and FZ. Data curation and investigation: CX. Formal analysis: ZY and CF. Methodology

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and software: CF. Resources and writing—original draft: ZY. Supervision and writing—review and editing: FZ. All authors contributed to the article and approved the submitted version.

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# **Conflict of interest**

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

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# Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/ fnut.2022.1025212/full#supplementary-material

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