



Association between serum iron status and primary liver cancer risk: a Mendelian randomization analysis

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Background: Serum iron status has been reported as associated with primary liver cancer (PLC) risk. However, whether iron status plays a role in the development of PLC remains inconclusive.

Methods: Genetic summary statistics of the four biomarkers (serum iron, ferritin, transferrin saturation, and transferrin) of iron status and PLC were retrieved from two independent genome-wide association studies (GWAS) that had been performed in European populations. Two-sample univariate and multivariate Mendelian randomization (MR) analyses were conducted to determine the causal link between iron status and PLC risk.

Results: No significant horizontal pleiotropy was detected for the four biomarkers according to the Mendelian Randomization Pleiotropy RESidual Sum and Outlier (MR-PRESSO) global test. No evidence of between-single nucleotide polymorphism (SNP) heterogeneity and directional pleiotropy was detected by the Cochran's Q test and MR-Egger regression for serum iron, ferritin, and transferrin. For transferrin saturation, although no heterogeneity was detected, the directional pleiotropy was significant (P value for intercept of MR-Egger regression =0.033). Univariate MR estimates based on inverse variance weighting (IVW) method suggested that there was no causal link between serum iron [odds ratio (OR) =0.71, 95% confidence interval (CI): 0.45 to 1.11], ferritin (OR =0.56, 95% CI: 0.16 to 2.04), and transferrin (OR =0.91, 95% CI: 0.72 to 1.15) and PLC risk. We found a significant causal relationship between transferrin saturation and PLC risk (OR =0.45, 95% CI: 0.22 to 0.90), although this link was non-significant in multivariate MR analysis.

Conclusions: There might be no causal relationship between iron status and PLC risk. However, data from larger sample size and people with different ethnic background were needed to further validate our findings.

Keywords: Primary liver cancer (PLC); Mendelian randomization (MR); iron status; ferritin; transferrin

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Introduction

Primary liver cancer (PLC) is a commonly diagnosed gastrointestinal carcinoma. According to the recently published global statistics, PLC was the sixth most commonly diagnosed cancer and the third leading cause of cancer death worldwide in 2020, with approximately 906,000 new cases and 830,000 deaths (1). Incidence of PLC had shown a decreasing trend in many high-risk countries in Asia, including China, South Korea, and the Philippines over the last 3 decades (2); however, an unfavorable increasing trend was noted in some developed countries where PLC was less diagnosed (2). Owing to the poor prognosis in clinical practice, PLC imposes a heavy disease burden on human health and warrants higher primacy in future schemes of disease prevention.

The malignancy of PLC is deemed to be multi-etiological and involves many risk factors (3,4). For example, previous studies have reported that iron status, which is measured in clinical practice as serum iron, ferritin, transferrin, and transferrin saturation, is associated with PLC risk (5-7). Mechanisms whereby iron may act in carcinogenesis are induction of oxidative stress, facilitation of tumor growth, and modification of the immune system (8). However, the links between serum status and PLC development from observational studies might be biased by underlying confounders and might only reflect an indirect association through other factors (i.e., aging, alcohol consumption, and insulin resistance) (8). Whether iron status plays a role in PLC development remains inconclusive. To address this need, herein, we conducted a 2-sample Mendelian randomization (MR) analysis to determine the causal link between biomarkers for iron status and PLC. The MR is an approach using genetic variants as instrumental variables for assessing causal relationships from observational data and has been widely used to infer the relationship between exposures and outcomes (9-12). We present the following article in accordance with the STROBE reporting checklist (available at <https://dx.doi.org/10.21037/atm-21-4608>).

Methods

Summary statistics of serum iron status and PLC

We retrieved the genetic data of serum iron status from the online platform of Integrative Epidemiology Unit (IEU) open genome-wide association study (GWAS) project (<https://gwas.mrcieu.ac.uk/>) with searching

codes of “ieu-a-1049”, “ieu-a-1050”, “ieu-a-1051”, and “ieu-a-1052” for serum iron, ferritin, transferrin saturation, and transferrin, respectively. The genetic data of serum iron status were derived from a GWAS consisting of 48,972 European individuals (13). In this GWAS, discovery samples consisted of summary data on genome-wide allelic associations between SNP genotypes and iron markers from 23,986 participants of European ancestry gathered from 11 cohorts in 9 participating centers. Replication samples to confirm suggestive and significant associations were obtained from up to 24,986 participants of European ancestry in 8 additional cohorts. Genome-wide association tests, genotype imputation, and associated quality control procedures were performed in each cohort separately (13). The association between genotyped and imputed SNPs and each iron phenotype was performed using an additive model for allelic effects, on the standardized residuals of the phenotype after adjusting for age, principal component scores, and other study specific covariates, for each gender separately (13).

The genetic data of PLC were retrieved from the online database of FinnGen (<https://r4.finnngen.fi/>), which is a public-private partnership project combining genotype data from Finnish biobanks and digital health record data from Finnish health registries (14). In GWAS of PLC, 266 PLC cases and 176,633 controls were included. Individuals with ambiguous gender, high genotype missingness (>5%), excess heterozygosity, and non-Finnish ancestry were excluded. Variants with high missingness (>2%), low Hardy-Weinberg equilibrium test P value (<10⁻⁶), and minor allele count <3 were excluded. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

Selection of genetic instruments

We extracted SNPs showing a significant association with serum iron status at the conventional GWAS threshold ($P < 5 \times 10^{-8}$). We performed a clumping process based on the linkage disequilibrium (LD) estimates from the European samples in 1,000 genomes project. Herein, the LD threshold was set as 0.1 and the window size was 10,000 kb. Among those pairs of SNPs that had LD estimate >0.1, we only retained the SNP with the lower P value. The SNPs with a minor allele frequency <1% were removed. We then retrieved the genetic statistics of the selected SNPs from the PLC GWAS summary data. For SNPs that were absent in the PLC GWAS, we alternatively retrieved data of SNP

proxy that had LD estimate >0.8 with the requested SNP.

Statistical analysis

We first tested the horizontal pleiotropy using Mendelian Randomization Pleiotropy RESidual Sum and Outlier (MR-PRESSO) global test and removed the outliers (i.e., SNPs with $P < 0.05$) if the horizontal pleiotropy was presented. Next, we tested the between-SNP heterogeneity using the inverse variance weighting (IVW) method. The Cochran's Q statistic was used to measure the heterogeneity. In the main analysis, IVW method with fixed-effect was applied if no between-SNP heterogeneity was detected, otherwise IVW method with random-effect was used. We also conducted a set of sensitivity analyses using MR-Egger regression, weighted median, and weighted mode methods. The MR-Egger regression is based on the INstrument Strength Independent of Direct Effect (InSIDE) assumption and consists of 3 parts: (I) a test for directional pleiotropy; (II) a test for a causal effect; and (III) an estimate of the causal effect (15). The analyses of the 4 MR methods were carried out using the *TwoSampleMR* package in R program (version 3.6.3; <https://www.r-project.org/>). We chose the main MR method as follows (16): (I) if no directional pleiotropy was detected, use IVW; (II) if directional pleiotropy was detected in MR-Egger regression, use MR-Egger; and (III) if directional pleiotropy and heterogeneity were both detected, use weighted median. Given the potential correlations between biomarkers of iron status, we also conducted a multivariate MR analysis.

Results

Quality control for genetic instruments

In this analysis, we included 22, 5, 36, and 41 independent genetic variants in MR analysis for serum iron, ferritin, transferrin saturation, and transferrin, respectively (Table 1). We did not detect horizontal pleiotropy for each biomarker of iron status in the MR-PRESSO global test (P value was 0.258, 0.354, 0.289, and 0.758 for serum iron, ferritin, transferrin saturation, and transferrin, respectively). Additionally, no evidence of between-SNP heterogeneity and directional pleiotropy was detected by the Cochran's Q test and MR-Egger regression for serum iron, ferritin, and transferrin (Table 2). For transferrin saturation, although no heterogeneity was detected, the directional pleiotropy was significant (P value for intercept of MR-Egger regression

$=0.033$). As a result, we reported MR estimates of IVW method for serum iron, ferritin, and transferrin, whereas estimates of MR-Egger regression for transferrin saturation.

Univariate MR analysis

The estimated effect sizes of the SNPs on both the exposures (serum iron, ferritin, transferrin saturation, and transferrin) and outcome (PLC) are displayed in scatter plots (Figure 1). According to the MR analysis, we estimated that there was no causal link between serum iron [odds ratio (OR) = 0.71; 95% confidence interval (CI): 0.45 to 1.11], ferritin (OR = 0.56, 95% CI: 0.16 to 2.04), and transferrin (OR = 0.91, 95% CI: 0.72 to 1.15) and PLC risk (Table 2). The findings were consistent with that of sensitivity analyses by another 3 MR methods. By contrast, we found a significant causal relationship between transferrin saturation and PLC risk (OR = 0.45, 95% CI: 0.22 to 0.90), although this link was not significant according to methods other than MR-Egger regression.

Multivariate MR analysis

In the multivariable MR analysis, a total of 76 genetic variants were included. No evidence of between-SNP heterogeneity (Q statistics = 66.10, $P = 0.610$) and directional pleiotropy (intercept of MR-Egger regression = -0.011, $P = 0.748$) was detected (Table 3). The IVW method estimated that there was no significantly causal relationship between the four biomarkers of iron status and PLC risk, which were concordant with estimates from the other three methods (Table 3).

Discussion

In this 2-sample MR study, we estimated the causal link between serum status and PLC risk based on genetic summary data from two previously large-scale GWASs. According to the estimates of MR analyses, we found there was no significant causal relationship between serum iron, ferritin, and transferrin and risk of PLC development. We detected a significant causal effect of transferrin saturation on PLC, although this effect was not significant in multivariable MR analysis. Our findings provide novel insights into the association of serum iron status with PLC genesis.

Links between serum iron status and PLC risk have been reported in previous epidemiological and experimental

Table 1 The genetic instruments used in this study

Rs ID	Chromosome	Position	Genes	Effect allele	Other allele	Beta	Se	P value
Serum iron								
rs10434845	6	25582757	<i>LRRC16A</i>	G	C	0.1047	0.0101	3.77E-25
rs11756569	6	26323026	NA	T	A	0.0748	0.0134	2.28E-08
rs12216125	6	25997458	<i>TRIM38, HIST1H1A</i>	T	C	0.1535	0.0103	1.18E-50
rs13209646	6	28600128	NA	A	T	0.097	0.0147	4.01E-11
rs1525892	3	133484712	<i>TF, SRPRB</i>	A	G	0.0736	0.0104	1.65E-12
rs1800562	6	26093141	<i>HFE, HIST1H4C, HIST1H1T</i>	A	G	0.3724	0.02	3.96E-77
rs1927695	6	25378113	<i>LRRC16A</i>	G	A	0.1066	0.0136	4.08E-15
rs198855	6	26103398	<i>HFE, HIST1H4C, HIST1H1T</i>	T	A	-0.0752	0.0112	1.88E-11
rs2074504	6	30530245	<i>GNL1, PRR3, ABCF1</i>	C	T	-0.063	0.0111	1.57E-08
rs2160907	22	37435900	<i>TST, MPST, KCTD17</i>	G	A	0.1225	0.0141	3.76E-18
rs2744257	6	25267748	<i>LOC101928663, LRRC16A</i>	C	T	-0.0679	0.0115	3.10E-09
rs407934	6	25504562	<i>LRRC16A</i>	T	C	0.0744	0.0103	4.80E-13
rs4712955	6	25684279	<i>SCGN</i>	G	A	0.068	0.0106	1.56E-10
rs518700	6	25889553	<i>SLC17A3</i>	A	T	0.0707	0.0108	6.46E-11
rs5756492	22	37424991	<i>TST, MPST</i>	A	G	0.0817	0.011	1.08E-13
rs7286184	22	37478775	<i>KCTD17, TMPRSS6</i>	G	A	-0.129	0.0124	1.74E-25
rs806794	6	26200677	<i>HIST1H2BE, HIST1H4D, HIST1H3D, HIST1H2AD,</i>	G	A	-0.0691	0.0112	7.34E-10
rs855791	22	37462936	<i>HIST1H2BF, HIST1H4E, HIST1H2BG, HIST1H2AE</i>	G	A	0.1868	0.0101	4.31E-77
rs916213	22	37397550	<i>KCTD17, TMPRSS6</i>	C	T	0.0789	0.01	2.20E-15
rs926331	22	37510072	<i>TEX33, TST, MPST</i>	T	C	0.0756	0.0119	2.01E-10
rs9358858	6	25446308	<i>TMPRSS6, IL2RB</i>	T	G	0.0807	0.0111	3.30E-13
rs9358928	6	26341030	<i>LRRC16A</i>	T	C	0.0953	0.0108	1.31E-18
Ferritin								
rs12216125	6	25997458	<i>TRIM38, HIST1H1A</i>	T	C	0.0687	0.0097	1.20E-12
rs12693541	2	190418690	<i>SLC40A1</i>	T	C	-0.106	0.014	4.18E-14
rs1800562	6	26093141	<i>HFE, HIST1H4C, HIST1H1T</i>	A	G	0.211	0.0187	1.42E-29
rs2413450	22	37470224	<i>KCTD17, TMPRSS6</i>	C	T	0.0559	0.0095	3.57E-09
rs368243	17	56708979	<i>TEX14</i>	C	T	-0.0512	0.0093	3.80E-08
Transferrin saturation								
rs1015811	6	28448086	NA	G	A	0.0833	0.0117	1.03E-12
rs10434845	6	25582757	<i>LRRC16A</i>	G	C	0.1414	0.0101	3.16E-44
rs10946813	6	26345141	NA	A	G	-0.0772	0.0104	8.90E-14

Table 1 (continued)

Table 1 (continued)

Rs ID	Chromosome	Position	Genes	Effect allele	Other allele	Beta	Se	P value
rs12196939	6	27787990	<i>HIST1H2BL, HIST1H2AI, HIST1H3H, HIST1H2AJ, HIST1H2BM, HIST1H4J, HIST1H4K, HIST1H2AK, HIST1H2BN</i>	A	G	-0.0916	0.0133	6.76E-12
rs12216125	6	25997458	<i>TRIM38, HIST1H1A</i>	T	C	0.2047	0.0103	2.29E-87
rs1233333	6	29795421	<i>HLA-G</i>	A	G	0.1063	0.014	3.24E-14
rs13209646	6	28600128	NA	A	T	0.1599	0.0148	3.36E-27
rs13215804	6	28415572	<i>ZSCAN23</i>	G	A	0.0617	0.0108	1.04E-08
rs1736919	6	29697517	<i>HLA-F, HLA-F-AS1</i>	G	A	-0.0732	0.0115	1.93E-10
rs1799852	3	133475722	<i>TF</i>	T	C	0.1458	0.0183	1.51E-15
rs1800562	6	26093141	<i>HFE, HIST1H4C, HIST1H1T</i>	A	G	0.5772	0.0203	1.52E-178
rs1927695	6	25378113	<i>LRRC16A</i>	G	A	0.1577	0.0137	8.34E-31
rs198839	6	26112620	<i>HFE, HIST1H4C, HIST1H1T, HIST1H2BC, HIST1H2AC</i>	G	T	-0.0889	0.0112	2.64E-15
rs2097775	6	30354303	NA	T	A	-0.1087	0.014	8.12E-15
rs2160907	22	37435900	<i>TST, MPST, KCTD17</i>	G	A	0.1203	0.0141	1.86E-17
rs221834	7	100343175	<i>ZAN</i>	G	C	0.1226	0.0205	2.38E-09
rs2393915	6	27080460	<i>HIST1H2BJ</i>	C	A	-0.0556	0.01	2.78E-08
rs2744258	6	25268014	<i>LOC101928663, LRRC16A</i>	C	A	-0.0924	0.0115	1.03E-15
rs301397	6	25464492	<i>LRRC16A</i>	C	T	0.0724	0.0109	2.86E-11
rs3129157	6	29141743	<i>OR2J2</i>	G	A	-0.1548	0.0281	3.46E-08
rs3804111	6	25602926	<i>LRRC16A</i>	G	T	-0.063	0.0099	2.13E-10
rs407934	6	25504562	<i>LRRC16A</i>	T	C	0.108	0.0103	1.24E-25
rs4712955	6	25684279	<i>SCGN</i>	G	A	0.1003	0.0106	4.05E-21
rs4712972	6	25772047	<i>SLC17A4, SLC17A1</i>	G	A	0.1093	0.0149	2.57E-13
rs518700	6	25889553	<i>SLC17A3</i>	A	T	0.0951	0.0109	1.94E-18
rs5756492	22	37424991	<i>TST, MPST</i>	A	G	0.0809	0.011	2.02E-13
rs6939576	6	28669315	NA	A	G	-0.0685	0.011	5.63E-10
rs7286184	22	37478775	<i>KCTD17, TMPRSS6</i>	G	A	-0.1344	0.0124	2.05E-27
rs7759489	6	24950599	<i>FAM65B</i>	T	C	0.0767	0.0137	2.20E-08
rs7762821	6	25335094	<i>LRRC16A</i>	G	C	0.0834	0.0119	2.19E-12
rs7773004	6	26267755	<i>HIST1H3F, HIST1H2BH, HIST1H3G, HIST1H2BI, HIST1H4H</i>	G	A	-0.1172	0.01	4.93E-32
rs8177272	3	133482870	<i>TF</i>	A	G	-0.097	0.0106	5.52E-20
rs855791	22	37462936	<i>KCTD17, TMPRSS6</i>	G	A	0.1921	0.0101	3.50E-80

Table 1 (continued)

Table 1 (continued)

Rs ID	Chromosome	Position	Genes	Effect allele	Other allele	Beta	Se	P value
rs916213	22	37397550	<i>TEX33, TST, MPST</i>	C	T	0.0806	0.01	8.94E-16
rs926331	22	37510072	<i>TMPRSS6, IL2RB</i>	T	C	0.0724	0.012	1.43E-09
rs9358858	6	25446308	<i>LRRC16A</i>	T	G	0.119	0.0112	1.70E-26
Transferrin								
rs10434845	6	25582757	<i>LRRC16A</i>	G	C	-0.1101	0.0104	2.74E-26
rs10456324	6	25600968	<i>LRRC16A</i>	A	G	0.0574	0.0102	1.87E-08
rs1051772	3	133327457	<i>CDV3, TOPBP1</i>	C	T	-0.1196	0.017	2.25E-12
rs10946813	6	26345141	NA	A	G	0.0667	0.0106	2.81E-10
rs1105877	3	133561095	<i>RAB6B</i>	C	T	-0.0865	0.0124	3.00E-12
rs1107413	3	133524717	<i>SRPRB, RAB6B</i>	C	G	-0.161	0.0116	9.74E-44
rs11757000	6	28484869	<i>GPX6, GPX5</i>	C	T	-0.161	0.0148	2.14E-27
rs11963067	6	28560071	<i>ZBED9</i>	T	A	0.085	0.0137	6.27E-10
rs1233333	6	29795421	<i>HLA-G</i>	A	G	-0.1105	0.0144	2.01E-14
rs129128	6	26125342	<i>HIST1H1T, HIST1H2BC, HIST1H2AC</i>	T	C	0.1155	0.0143	7.42E-16
rs13080154	3	133475722	NA	C	T	0.1855	0.0108	9.98E-66
rs1410440	6	25415730	<i>LRRC16A</i>	A	C	-0.0579	0.0104	2.93E-08
rs1444598	3	133553985	NA	C	T	0.1091	0.0104	8.24E-26
rs1444600	3	133599961	<i>CDV3</i>	T	C	0.1365	0.0105	2.64E-38
rs1495741	8	18272881	<i>NAT2</i>	A	G	0.0825	0.0122	1.57E-11
rs174577	11	61604814	<i>FADS2</i>	A	C	0.0684	0.0107	1.90E-10
rs1799852	3	133476852	<i>TF</i>	T	C	-0.3777	0.0191	1.72E-87
rs1800562	6	26093141	<i>HFE, HIST1H4C, HIST1H1T</i>	A	G	-0.5496	0.0208	1.26E-153
rs2032447	6	26044369	<i>HIST1H4B, HIST1H3B, HIST1H2AB, HIST1H2BB, HIST1H3C, HIST1H1C</i>	G	A	-0.0952	0.0108	1.07E-18
rs2393915	6	27080460	<i>HIST1H2BJ</i>	C	A	0.0822	0.0102	8.02E-16
rs2690097	6	25346742	<i>LRRC16A</i>	C	T	-0.0741	0.012	5.69E-10
rs2692670	3	190378750	<i>SRPRB, RAB6B</i>	C	T	-0.1663	0.013	1.06E-37
rs2715630	3	133699575	<i>RAB6B</i>	T	C	0.083	0.0147	1.61E-08
rs3094124	6	30711805	<i>TUBB, FLOT1, IER3</i>	G	C	0.1537	0.0178	5.25E-18
rs3094188	6	31142245	<i>CCHCR1, TCF19, POU5F1, PSORS1C3</i>	A	C	0.0592	0.0106	2.04E-08
rs3811658	3	133476852	<i>TF</i>	T	C	0.3883	0.0109	1.00E-200
rs407934	6	25504562	<i>LRRC16A</i>	T	C	-0.0895	0.0106	3.33E-17
rs4711080	6	25189192	NA	C	A	0.0674	0.0113	2.74E-09

Table 1 (continued)

Table 1 (continued)

Rs ID	Chromosome	Position	Genes	Effect allele	Other allele	Beta	Se	P value
rs4712955	6	25684279	SCGN	G	A	-0.0898	0.0108	1.17E-16
rs5009711	6	25333090	LRRC16A	T	C	-0.1468	0.0144	2.06E-24
rs6923367	6	25745852	HIST1H2AA, HIST1H2BA, HIST1H2APS1, SLC17A4	T	A	-0.3798	0.021	3.09E-73
rs722086	6	25428954	LRRC16A	T	C	-0.0886	0.0134	3.96E-11
rs744653	2	133492471	NA	T	C	0.0916	0.0144	2.00E-10
rs7646392	3	195827205	SLCO2A1	T	C	-0.0615	0.0105	4.49E-09
rs7762537	6	25334967	LRRC16A	C	A	-0.0936	0.0122	1.42E-14
rs9268633	6	32406473	HLA-DRA	G	A	0.0717	0.0128	2.31E-08
rs9358858	6	25446308	LRRC16A	T	G	-0.1108	0.0115	3.99E-22
rs9358928	6	26341030	NA	T	C	-0.109	0.0111	9.07E-23
rs9405124	6	29368813	OR12D2	G	A	-0.0792	0.0111	1.19E-12
rs9824452	3	1.33E+08	TF, SRPRB	A	G	-0.3321	0.0149	1.28E-110
rs9990333	3	1.96E+08	TFRC	T	C	-0.067	0.0101	3.01E-11

The nearest genes were detected within a flanking distance of 20 kb. NA, not available.

Table 2 The association between serum iron status and liver cancer risk according to univariate MR

Variables	Iron	Ferritin	Transferrin saturation	Transferrin
Inverse variance weighted				
OR (95% CI)	0.71 (0.45, 1.11)	0.56 (0.16, 2.04)	0.91 (0.66, 1.26)	0.91 (0.72, 1.15)
Q statistics (P value)	16.24 (0.436)	1.02 (0.796)	29.70 (0.429)	28.40 (0.738)
MR-Egger				
OR (95% CI)	0.53 (0.18, 1.57)	1.09 (0.05, 22.18)	0.45 (0.22, 0.90)	1.26 (0.85, 1.87)
Q statistics (P value)	15.89 (0.389)	0.795 (0.672)	24.65 (0.647)	24.61 (0.853)
Intercept (P value)	0.034 (0.574)	-0.062 (0.681)	0.093 (0.033)	-0.058 (0.060)
Weighted median				
OR (95% CI)	0.66 (0.35, 1.26)	0.62 (0.09, 4.10)	0.80 (0.49, 1.28)	1.05 (0.76, 1.45)
Weighted mode				
OR (95% CI)	0.68 (0.37, 1.24)	0.58 (0.14, 2.40)	0.80 (0.51, 1.26)	1.05 (0.75, 1.47)

MR, Mendelian randomization; OR, odds ratio; CI, confidence interval.

studies (17,18). A recent meta-analysis reported an association between high serum ferritin and PLC risk (HR =1.49, 95% CI: 1.13 to 1.96) and high serum iron and PLC risk (HR =2.47, 95% CI: 1.31 to 4.63) (5). The previous results obtained from observational studies were not in line with our findings. The potential explanations for this

inconsistency might be as follows. First, previous studies involved small numbers of studies and sample size, leading to wide confidence intervals for the association estimates. Additionally, estimates from observational studies might be subject to the inherent defects of residual confounding and reverse causality (19). Findings obtained from

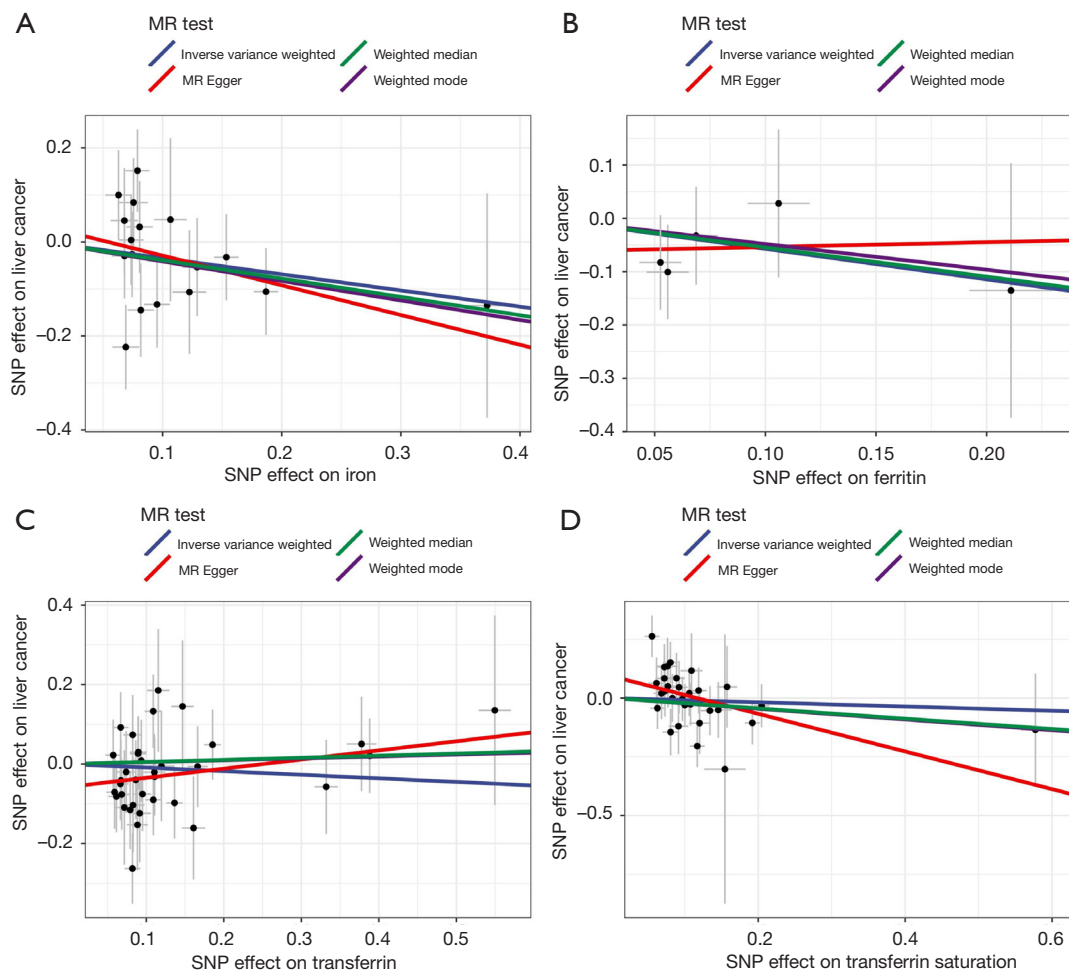


Figure 1 Scatter plots for MR analyses of the causal effect of serum iron status on PLC in clinical practice. (A) Iron; (B) ferritin; (C) transferrin; (D) transferrin saturation. The slope of each line corresponds to the estimated MR effect per method. PLC, primary liver cancer; MR, Mendelian randomization.

experimental studies were not conclusive regarding the direct carcinogenic effect of iron status on the liver, whereas indirect pathways were revealed between iron and PLC such as promoting oxidative stress, cell death, and compensatory proliferation (20). In this regard, the abnormal iron status might be a result of liver carcinogenesis. The causal relationship between iron status and PLC development is debatable and needs further investigations.

In a recent 2-sample MR analysis, Yuan *et al.* reported that genetically predicted iron status was positively associated with liver cancer (21). However, only 3 SNPs were used to serve as genetic instruments of iron status in this study, and thus may have led to a weak instrumental variable bias. On the contrary, in our study, we used all SNPs that reached the GWAS significance threshold as

the genetic instruments after quality-control processes. Different to results of Yuan *et al.*, we found no causal relationship between serum iron, ferritin, and transferrin and risk of PLC. This inconsistency might be attributed to the following reasons: (I) different genetic instrumental tools used; and (II) summary data of PLC GWAS were derived from different populations: Yuan *et al.* retrieved from genetic data from UK Biobank, in which the number of PLC cases included in GWAS was less than that of FinnGen (<https://pan.ukbb.broadinstitute.org/phenotypes/index.html>). Since the GWAS results were largely dependent on sample size and ratio of case-control (22), we speculate that the variations in genetic data of PLC resulted in the aforementioned inconsistency. Future studies based on a more robust GWAS are therefore warranted.

Table 3 The association between serum iron status and liver cancer risk according to multivariate MR

Variables	Iron	Ferritin	Transferrin saturation	Transferrin
Inverse variance weighted				
OR (95% CI)	1.22 (0.91, 1.64)	1.00 (0.63, 1.57)	1.14 (0.77, 1.67)	0.94 (0.85, 1.03)
Q statistics (P value)			66.10 (0.610)	
MR-Egger				
OR (95% CI)	1.28 (0.78, 2.08)	1.00 (0.60, 1.68)	1.17 (0.96, 1.41)	0.94 (0.81, 1.10)
Q statistics (P value)			60.82 (0.758)	
Intercept (P value)			-0.011 (0.748)	
Weighted median				
OR (95% CI)	0.99 (0.64, 1.55)	1.00 (0.50, 1.98)	1.01 (0.74, 1.39)	0.92 (0.75, 1.12)
Lasso method				
OR (95% CI)	1.15 (0.85, 1.56)	0.99 (0.59, 1.66)	1.16 (0.96, 1.41)	0.94 (0.81, 1.10)

MR, Mendelian randomization; OR, odds ratio; CI, confidence interval; Lasso, least absolute shrinkage and selection operator.

We also found a putative causal link between genetically predicted transferrin saturation and PLC risk. However, this link disappeared in the multivariable MR analysis. Our results suggested that there was no compelling evidence with respect to a direct role of iron status in PLC development, although establishment of this link had been attempted in previous experimental studies (18,23). Iron overload and iron deficiency are both related to significant abnormalities in immune function (24). The most mentioned mechanism by which excess iron may promote liver carcinogenesis is through DNA damage from the production of reactive oxygen species, especially hydroxyl radical (25). However, of note is that iron overload was also deemed to be a hepatic presentation of alcoholic liver diseases, which is a well-determined etiology for liver cancer. Alcoholic liver diseases are associated with significant oxidative stress as well as the hepatic accumulation of iron, a transition element also documented to initiate oxidative stress (26). Alcohol intake is therefore an important confounder to consider the observed correlations of iron status with PLC risk.

In our study, we included several genetic variants such as rs1800562 and rs198855 in *HFE* and rs855791 in *TMPRSS6* that show a robust and consistent association with a systemic iron status. Mutations in *HFE* and *TMPRSS6* were reported to be associated with hereditary hemochromatosis (HH) (27,28), which is a risk factor for the development of liver carcinoma (29). However, the population-attributable fraction of HH might be small due to HH being rarely

diagnosed in general population. On the other hand, variants that directly cause HH were not always associated with elevated risk of liver cancer (30). In this regard, HH could only be termed as a risk factor rather than an etiology like viral hepatitis for liver cancer. In addition, liver iron metabolism signatures were related to survival, disease status, and prognosis in patients with hepatocellular carcinoma. These findings suggested an important role of iron in the development and progression of liver cancer.

The limitations of our study should be noted here. First, our results were based on genetic data from European populations, which limited the possibility of extrapolation to other populations. Second, the genetic summary data of PLC were derived from a GWAS with a small case size and an unbalanced case-control ratio, which might limit the statistical power and introduce variations into the MR estimates.

In summary, our study found that there might be no causal relationship between iron status and liver cancer risk. The previously observed links might be confounded by underlying factors and require further validation.

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Footnote

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Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

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