



A bidirectional mendelian-randomization analyses of genetically predicted circulating levels of systemic inflammatory regulators with risk of sepsis

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Abstract

Whether there is a causal relationship between circulating levels of systemic inflammatory regulators and sepsis remains unclear. To determine whether genetically predicted circulating levels of cytokines are associated with risk of sepsis, a bidirectional twosample Mendelian randomization (MR) analysis based on the a STROBE-compliant cross-sectional observational study was conducted utilizing gene-wide association study (GWAS) data. Selected with rigor, single-nucleotide polymorphisms served as instrumental variables for subsequent MR analysis. The preferred method for the MR analysis was the inverse-variance weighted approach. However, for comprehensive sensitivity analyses, 6 additional MR methods were employed. Cochrane's Q test was performed to examine heterogeneity. A leave-one-out method ensured the stability of MR results. Our findings suggest an inverse association between the levels of beta-nerve growth factor (BNGF) and the risk of sepsis development (OR = 0.769, 95% CI = 0.599-0.987, P = .039). In contrast, higher levels of TNF-related apoptosis-inducing ligand and vascular endothelial growth factor A (VEGF-A) are positively correlated with sepsis risk (OR = 1.094, 95% CI = 1.012-1.183, P = .025; OR = 1.182, 95% CI = 1.016-1.375, P = .031, respectively). Reverse MR Analysis indicated that sepsis risk is linked with lower circulating levels of adenosine deaminase and Interleukin-17A ($\beta = -0.043$, 95% CI = -0.085 to -0.002, P = .042; $\beta = -0.061$, 95% CI = -0.108 to -0.013, P = .012, respectively), and also with higher circulating levels of BNGF, delta/notchlike epidermal growth factor-related receptor, fibroblast growth factor 23, leukemia inhibitory factor, monocyte chemoattractant protein-1, and osteoprotegerin $(\beta = 0.056,\ 95\%\ CI = 0.015 - 0.096,\ P = .007;\ \beta = 0.137,\ 95\%\ CI = 0.035 - 0.240,\ P = .009;\ \beta = 0.118,\ 95\%\ CI = 0.020 - 0.216,$ P = .018; $\beta = 0.136$, 95% CI = 0.020–0.252, P = .022; $\beta = 0.143$, 95% CI = 0.043–0.242, P = .005; $\beta = 0.116$, 95% CI = 0.010–0.010 0.222, P = .031, respectively). Sum up, our study provides evidence supporting a bidirectional causal relationship between sepsis and genetically predicted circulating levels of systemic inflammatory regulators.

Abbreviations: ADAR1 = adenosine deaminase acting on RNA 1, BNGF = beta-nerve growth factor, CIs = confidence intervals, DNER = delta/notchlike epidermal growth factor-related receptor, FGF23 = fibroblast growth factor 23, GWAS = gene-wide association study, IL = interleukin, IV = instrumental variable, IVW = inverse variance weighted, LIF = leukemia inhibitory factor, MCP-1 = monocyte chemoattractant protein-1, MR = Mendelian randomization, OPG = osteoprotegerin, OPGL = osteoprotegerin ligand, ORs = odds ratios, SES = socio-economic status, SNPs = single nucleotide polymorphisms, TNF = tumor necrosis factor, TRAIL = TNF-related apoptosis-inducing ligand, VEGF-A = vascular endothelial growth factor A.

Keywords: genetic causal association, gwas, inflammatory modulators, mendelian randomization, sepsis, systemic inflammation

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All data generated or analyzed during this study are included in this published article [and its supplementary information files].

Each study incorporated in the GWAS used in the present study was approved by local research ethics committees or Institutional Review Boards, and all participants had given their informed consent.

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1. Introduction

Sepsis is commonly understood as a disorderly immune response of the host, provoked by an infection.^[1] The pathogenesis of sepsis intertwines with numerous pathophysiological changes, such as imbalances in the inflammatory response, irregularities in immune regulation, coagulation dysfunction, mitochondrial damage, and endoplasmic reticulum stress.^[2] Given that sepsis lacks specific clinical manifestations, diagnosis in its early stages can be challenging. A delay in detection often leads to swift disease progression and increased mortality rates.^[3]

Inflammatory factors refer to a variety of cytokines engaged in the inflammation process. These cytokines, small peptide proteins secreted by cells, serve essential functions in cell proliferation and differentiation, signaling pathways, and the regulation of immune-inflammatory responses.^[4] Recent studies^[5,6] underscore the crucial role of cytokines in sepsis onset and progression. In early sepsis, pro-inflammatory and anti-inflammatory cytokines coexist, battling for immune homeostasis. If the pro-inflammatory response gains the upper hand, it can trigger an immune imbalance, causing a cytokine storm^[7] and initiating a systemic inflammatory response syndrome with microcirculation dysfunction, escalating eventually to severe multiple organ dysfunction syndrome. [8] The levels of cytokines correlate closely with the severity of organ failure and may potentially act as biomarkers for sepsis.^[9] However, typical studies often restrict examination to a single cytokine, ignoring the potential influences of other cytokines on sepsis development. Further, substantial clinical trials are lacking that verify the sensitivity and specificity of certain cytokines for early sepsis identification.^[10]

Numerous genome-wide association studies (GWASs) have explored the links between genetic variations and diseases or phenotypes. [11] Mendelian randomization, a rigorously statistical method for causation detection, uses genetic variants with significant exposure associations as instrumental variables (IVs) to examine the causal link between the exposure and the outcome. [12] Two-sample Mendelian randomization (MR) analysis can provide causal estimates using single-nucleotide polymorphisms (SNPs) associated with exposure and outcome from independent GWAS studies. [13] MR studies are less likely to be affected by reverse causality and potential environmental-social confounders. MR serves to complement traditional epidemiological approaches by mitigating specific biases such

as confounding and reverse causation. Evaluating the associations between cytokine levels and sepsis risk involves not only identifying which cytokines are involved but also understanding the strength and implications of these associations. In our study, we applied MR to assess the causal relationships between genetically predicted cytokine levels and sepsis. The distinctions between weak, moderate, and strong associations are critical. Strong associations signify a robust and potentially causative link, suggesting these cytokines as strong candidates for therapeutic targets or biomarkers for early detection. In contrast, weaker associations, though less direct, should not be overlooked, as they might represent complex interactions within broader immunological pathways or highlight cytokines involved in secondary or tertiary response mechanisms. This study employs GWASs and bidirectional Mendelian randomization analyses to investigate the causal associations between circulating levels of systemic inflammatory regulators and sepsis risk, which provides evidence for a potential causal relationship and is expected to serve as the basis for further longitudinal or intervention studies..

2. Methods

2.1. Ethics statement

Each study incorporated in the GWAS used in the present study was approved by local research ethics committees or Institutional Review Boards, and all participants had given their informed consent.

2.2. Study design

The Mendelian Randomization analysis is predicated on 3 fundamental assumptions^[14]: The IVs used in the analysis exhibit a strong association with the variable of interest; The IVs are not associated with any confounding factors that could influence the relationship between exposure and outcome; and The IVs only affect the outcome via their influence on exposure. These assumptions are graphically depicted in Figure 1.

Figure 2 briefly portrays the bidirectional MR design. Genetic instruments for 91 systemic inflammatory regulators were sourced from the latest GWAS Catalog (https://www.ebi.ac.uk/

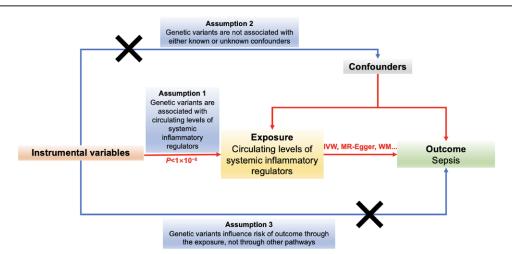


Figure 1. The Directed Acyclic Graph (DAG) representing the Mendelian Randomisation (MR) framework employed to investigate the causal relationship between the circulating levels of systemic inflammatory regulators and sepsis. The MR analysis is guided by 3 crucial instrumental variable assumptions: (1) The instrumental variables must exhibit an association with the circulating levels of systemic inflammatory regulators ($P < 5 \times 10^{-8}$). (2) The instrumental variables must not be associated with any potential confounders that could influence the relationship between the circulating levels of systemic inflammatory regulators and sepsis. (3) The instrumental variables should solely impact the risk of sepsis through their influence on the the circulating levels of systemic inflammatory regulators. The instrumental variables are represented by SNPs, and the MR analysis employs the IVW method, Wald ratio, MR-Egger, Weighted-Median, Simple Mode, Weighted Mode and MR-PRESSO methods to estimate causal relationships. IVW = inverse variance weighted, SNPs = single nucleotide polymorphisms.

Study design and workflow

- GWAS summary data for circulating levels of systemic inflammatory regulators (Exposure)
 GWAS summary data for sepsis (Outcome)
 - Extracted SNPs that associated with exposure (P < 5×10-8)
 Extracted SNPs that match with exposure
 - Harmonize effects size and alleles of SNPs on the exposure and outcome data
- Clumping process (r₂ threshold < 0.001, clumping distance = 10000kb)
- MR analysis, sensitivity analysis, reverse MR analysis ($P < 5 \times 10^{-6}$)

Figure 2. Flow chart of this study.

gwas/publications/37563310) and IEU Open GWAS (https://gwas.mrcieu.ac.uk/).^[15-17] Initially, genetic variants for each inflammatory regulator were chosen to deduce the causality from each regulator to sepsis. Subsequently, genetic variants linked with sepsis were used to deduce the causality from sepsis to inflammatory regulators. Lastly, estimates from the sepsis source were amalgamated using the meta-analysis method. All the studies incorporated in the original GWASs obtained approval from the appropriate institutional review board. The study's flowchart is delineated in Figure 2.

2.3. Data source for inflammatory regulators

This study draws upon a published large-scale GWAS meta-analysis, involving up to 91 inflammation-related proteins across 14,824 participants, ^[17] with some contributions from Chinese populations. Despite being predominantly European, the inclusion of these datasets enhances the study's scope by integrating findings across different ethnic backgrounds and geographic origins. Distributions of these 91 cytokines underwent normalization through a two-step inverse transformation. The cytokine distributions were initially normalized by inverse transformation. Then, inverse transformation was applied to residuals of the linear regression model of the transformed cytokines, considering age, gender, body mass index (BMI), and genetic principal components. ^[18] To amalgamate genetic associations across multiple cohorts, meta-analyses were conducted.

2.4. Data sources for sepsis

Summary-level data on sepsis was obtained from the GWAS catalog and the IEU Open GWAS project (https://gwas.mrcieu.ac.uk; https://www.ebi.ac.uk/gwas/).^[19] The use of multiple data sources was intended to enhance the credibility of the findings

due to high sensitivity and specificity. The chosen GWAS in the UK Biobank encompasses 11,643 sepsis cases and 474,841 controls (with a maximum 16% overlap with insomnia GWAS). Case definition adhered to the explicit sepsis criteria set out in the most recent Global Burden of Disease Study of Sepsis. [20] This European-centric composition reflects a substantial portion of our data, providing meaningful insights into genetic associations within this demographic. Sepsis admissions within the UK Biobank were identified using linked secondary care data coded by the International Classification of Diseases (ICD). ICD-10 codes A02, A39, A40, and A41 were utilized to diagnose sepsis, aligning with contemporary literature. [21] Cases were included if the code appeared in the primary or secondary diagnostic position in Hospital Episode Statistics (HES) data or comparable datasets in the devolved nations, as provided by the UK Biobank. Study participants were predominantly of European ancestry, including both genders. The GWAS catalog included 1573 cases and 454,775 controls as outcomes.^[22] Self-reported cases or cases presenting solely in primary care were excluded from the study.

2.5. The selection of instruments

Under bidirectional mendelian-randomization (BIMR), the genetic variants serving as instrumental variables for exposure X (Gx) and those serving as instrumental variables for outcome Y (Gy) were entirely distinct. In the initial stage, a forward MR analysis was conducted using a genome-wide threshold of significance ($P < 5 \times 10^{-8}$) to shield against the selection of false-positive instruments. Subsequently, conducting a reverse MR analysis, we studied the relationship of sepsis to the circulating levels of systemic inflammatory regulators. The positions of sepsis and the circulating levels of systemic inflammatory regulators were designated as exposure and outcome, respectively. As we did not screen sites when we imposed a threshold of 5×10^{-8} ,

the P value for exposure IVs was set at 5×10^{-6} to investigate the causative impact of sepsis on systemic inflammatory regulators. All SNPs in linkage disequilibrium $(r^2 < 0.001)$ within 10 Mb in the European 1000G reference panel $(r^2 < 0.001)$ were pruned, retaining the SNPs with the lowest P value as independent variables. Following the harmonization of the selected SNPs with outcome data, all 91 systemic inflammation regulators were chosen. To mitigate the presence of weak IVs, the F-statistics of the SNPs were averaged, considering IVs with F-statistics over 10 as strong. (r^2) This method helps to avoid the problem of weak instrumental variables, so as to improve the accuracy of causal effect estimation.

2.6. Statistical analysis

Table S1, Supplemental Digital Content, http://links.lww.com/MD/O687 encapsulates details of the MR analyses based on summary-level data from GWAS used in this BIMR analysis. The table outlines exposure numbers, exposure names, outcome numbers, outcome names, SNP names, and the chromosome and position of the SNP in the outcome and exposures data. Subsequently, it presents effect alleles, other alleles, beta values, standard error (SE) values, *P* values, sample size, gene frequencies, *R*² values, and *F* values in sequence.

A bidirectional two-sample MR procedure, utilizing summary association data, was employed to examine the causal relationship between inflammatory regulators and sepsis. We executed data harmonization meticulously to ensure the same allele corresponded to the effect of an SNP on the exposure and the outcome. For SNPs with varying effect alleles due to different strands, we adjusted the strand to align the effect allele in both datasets. However, harmonizing palindromic SNPs poses challenges due to the allele uniformity on both strands. To preclude ambiguity regarding whether exposure and outcome GWAS report the same effect allele, [25] such SNPs were deleted. In the primary analysis, a Wald ratio estimate was computed for each genetic variant, and these estimates were synthesized using the inverse variance weighted (IVW) procedure. The IVW with a multiplicative random effects strategy offers an efficient estimate, taking into account possible heterogeneity among the Wald ratio estimates from SNPs. [26] Therefore, if heterogeneity is observed, random-effects IVW models are applied; otherwise, fixed-effect IVW models are used. The Cochran Q test and I^2 evaluated the heterogeneity of the meta-analysis. Scatter plots were provided to illustrate the causal associations of systemic inflammatory regulators with sepsis. The effects in 91 cytokines are reported as changes in inverse normalized cytokine concentrations per effect allele dosage. The impact of 91 cytokines on sepsis are presented as odds ratios (ORs) with 95% confidence intervals (CIs) per 1 SD genetically predicted cytokine change. The influences of sepsis on systemic inflammatory regulators are reported as β coefficients with 95% CIs.^[13] Given the exploratory nature of this study and the high dimensionality of cytokine data (n = 91), we prioritized sensitivity analyses over strict multiple testing correction. While this approach increases the risk of Type I errors, it aligns with recent MR guidelines for hypothesis generation. [12] Future confirmatory studies should apply methods such as Benjamini-Hochberg correction to control false discovery rates.

2.7. Sensitivity analysis

The Inverse-Variance Weighted (IVW) method was employed as the primary method of analysis. When the number of instrumental variables was limited to one, we utilized the Wald ratio analysis method. Heterogeneity in IVW estimates was then scrutinized via the Cochran Q test. Horizontal pleiotropy was evaluated by the intercept *P* value of the MR-Egger. Additionally, MR-Egger, Weighted Median, Simple

Mode, Weighted Mode, and MR-PRESSO methods served as sensitivity analysis methods. The P value from the pleiotropy test was adopted to examine the presence of pleiotropy. If the obtained P value exceeded .05, this indicated a negligible risk of pleiotropy in the causal analysis, allowing it to be overlooked. To ascertain the consistency of the results, a leave-one-out analysis was implemented. MR-Egger analysis measures instrumental variable pleiotropy, wherein a non-zero intercept demonstrates that the IVW estimate may be skewed.[27] A weighted median, conversely, can deliver a consistent estimate for the causal effect even if up to half of the SNPs infringe on horizontal pleiotropy. [28] As this study's large sample size accommodated the exploration of numerous potential positives, we did not perform multiple testing correction for adjusting significance levels. All analyses were twosided and executed using the TwoSampleMR (version 0.5.6) packages in R software (version 3.6.3). The report adhered to the STROBE-MR statement.

3. Results

3.1. Causal effects of circulating levels of systemic inflammatory regulators on sepsis

We did not find any potential correlation between genetically predicted circulating levels of systemic inflammatory regulators in the GWAS Catalog data (All P > .05) (Fig. 3), but based on the IEU-open GWAS project, lower beta-nerve growth factor levels is inversely associated with decreased risks of sepsis (OR = 0.769, 95% CI = 0.599-0.987, P = .039), because there are too few sites of SNPs (n = 2), MR analysis other than IVW cannot be performed. And we also found that higher Tumor Necrosis Factor (TNF)-related apoptosis-inducing ligand levels and vascular endothelial growth factor A level are inversely associated with decreased risks of sepsis (OR = 1.094, 95% CI = 1.012-1.183, P = .025; OR = 1.182, 95% CI = 1.016–1.375, P = .031, respectively) (Fig. 4), the MR-Egger for them were OR = 1.135, 95% CI = 0.996-1294, P = .107, and OR = 1.522, 95% CI = 0.917-2.525, P = .246, respectively. Furthermore, Q values based on MR-Egger and IVW tests showed that there was no obvious heterogeneity (all P > .05).

3.2. Causal effects of sepsis on circulating levels of systemic inflammatory regulators

After MR Analysis of the data from the GWAS Catalog, we found that the lower circulating level of adenosine deaminase and Interieukin-17A (IL-17A) were related to an increased risk of sepsis ($\beta = -0.043$, 95% CI = -0.085 to -0.002, P = .042; $\beta = -0.061$, 95% CI = -0.108 to -0.013, P = .012, respectively) using IVW methods, and higher circulating level of beta-nerve growth factor was found to be related to an increased risk of sepsis ($\beta = 0.056$, 95% CI = 0.015–0.096, P = .007) (Fig. 5). MR analysis based on IEU-open GWAS project data showed that the higher circulating level of Delta/Notch-like EGFrelated receptor (DNER), Fibroblast growth factor 23, leukemia inhibitory factor, monocyte chemoattractant protein-1 and Osteoprotegerin were related to an increased risk of sepsis $(\beta = 0.137, 95\%)$ CI = 0.035-0.240, P = .009; $\beta = 0.118$, 95% CI = 0.020–0.216, P = .018; $\beta = 0.136$, 95% CI = 0.020– 0.252, P = .022; $\beta = 0.143$, 95% CI = 0.043–0.242, P = .005; $\beta = 0.116$, 95% CI = 0.010–0.222, P = .031, respectively) (Fig. 6). MR-Egger Intercept did not detect potential horizontal pleiotropy for them (OR = 1.008, 95% CI = 0.923-1.102, P = .858; OR = 0.099, 95% CI = 0.916-1.090, P = .991; OR = 0.909, 95% CI = 0.821-1.006, P = .106; OR = 0.927, 95% CI = 0.753-1.140, P = .486; OR = 1.165, 95% CI = 0.945-1.437, P = .178; OR = 1.129, 95% CI = 0.872–1.463, P = .375; OR = 1.168, 95% CI = 0.943-1.445, P = .180; OR = 0.974,

Exposure	Method	nSNP	OR(95%CI)		P-value
EIF4EBP1 levels	IVW	3	1.106(0.733-1.671)		0.631
Adenosine Deaminase levels	IVW	2	0.997(0.859-1.156)	⊢	0.964
beta-nerve growth factor levels	IVW	2	0.733(0.379-1.415)	—	0.354
Caspase 8 levels	Wald ratio	1	0.830(0.495-1.392)	⊢	0.480
Eotaxin levels	IVW	6	0.886(0.670-1.172)	—	0.398
C-C motif chemokine 19 levels	IVW	3	1.024(0.748-1.401)	———	0.883
C-C motif chemokine 20 levels	IVW	2	0.856(0.503-1.460)		0.569
C-C motif chemokine 23 levels	IVW	5	0.977(0.821-1.162)	⊢	0.792
C-C motif chemokine 25 levels	IVW	7	0.973(0.870-1.089)	⊢	0.639
C-C motif chemokine 28 levels	IVW	3	1.101(0.602-2.014)		0.755
C-C motif chemokine 4 levels	IVW	4	0.918(0.791-1.066)		0.264
Natural killer cell receptor 2B4 levels	IVW	9	1.140(0.915-1.421)		0.242
CD40L receptor levels	IVW	4	0.983(0.829-1.164)	⊢	0.839
T-cell surface glycoprotein CD5 levels	IVW	6	0.870(0.654-1.157)	——	0.338
T-cell surface glycoprotein CD6 isoform levels	IVW	4	1.109(0.977-1.259)	↓	0.110
CUB domain-containing protein 1 levels	IVW	4	1.184(0.935-1.498)	-	0.161
Macrophage colony-stimulating factor 1 levels	IVW	3	1.335(0.928-1.920)	· · · · · · · · · · · · · · · · · · ·	0.119
Cystatin D levels	IVW	10	1.031(0.875-1.214)	—	0.715
Fractalkine levels	IVW	4	0.886(0.557-1.407)		0.608
C-X-C motif chemokine 1 levels	IVW	2	0.900(0.690-1.174)		0.437
C-X-C motif chemokine 10 levels	IVW	6	1.120(0.848-1.479)	⊢	0.425
C-X-C motif chemokine 11 levels	IVW	5	0.928(0.705-1.221)	——	0.592
C-X-C motif chemokine 5 levels	IVW	6	1.047(0.881-1.243)	⊢ •	0.605
C-X-C motif chemokine 6 levels	IVW	3	0.968(0.839-1.118)	⊢	0.661
C-X-C motif chemokine 9 levels	IVW	4	0.964(0.567-1.640)		0.893
DNER levels	IVW	5	1.184(0.860-1.631)	⊢	0.299
Protein S100-A12 levels	IVW	3	1.008(0.633-1.605)		0.974
Fibroblast growth factor 19 levels	IVW	3	0.730(0.522-1.021)	<u> </u>	0.066
Fibroblast growth factor 21 levels	IVW	4	1.048(0.779-1.408)		0.758
Fibroblast growth factor 23 levels	IVW	2	1.069(0.617-1.853)	<u> </u>	0.812
Fibroblast growth factor 5 levels	IVW	5	0.956(0.832-1.099)	⊢ •	0.528
Fms-related tyrosine kinase 3 ligand levels	IVW	10	1.002(0.808-1.241)		0.988
GDNF levels	IVW	6	0.834(0.629-1.105)		0.206
Hepatocyte growth factor levels	IVW	2	1.222(0.759-1.967)	<u> </u>	0.409
Interleukin-10 levels	IVW	4	1.186(0.816-1.723)		0.371
Interleukin-10 receptor subunit beta levels	IVW	3	1.030(0.873-1.217)		0.724
Interleukin-12 subunit beta levels	IVW	13	1.010(0.849-1.202)	⊢	0.907

Exposure	Method	nSNP	OR(95%CI)		P-value
Interleukin-13 levels	Wald ratio	1	1.642(0.635-4.241)	•	0.306
Interleukin-15 receptor subunit alpha levels	IVW	4	1.043(0.862-1.261)	⊢•	0.664
Interleukin-17C levels	IVW	2	0.844(0.488-1.463)	· · · · · · · · · · · · · · · · · · ·	0.546
Interleukin-18 levels	IVW	4	0.989(0.762-1.284)	⊢	0.935
interleukin-18 receptor 1 levels	IVW	10	0.924(0.812-1.051)	⊢● -I	0.227
Interleukin-1-alpha levels	Wald ratio	1	0.810(0.483-1.360)	——	0.426
Interleukin-20 receptor subunit alpha levels	Wald ratio	1	0.689(0.285-1.664)	· · · · · · · · · · · · · · · · · · ·	0.407
Interleukin-6 levels	Wald ratio	1	0.846(0.555-1.290)	—	0.438
Interleukin-7 levels	Wald ratio	1	0.888(0.435-1.813)	-	0.745
Interleukin-8 levels	IVW	2	0.958(0.504-1.822)	•	0.897
Latency TGF beta 1 levels	IVW	2	1.205(0.775-1.873)		0.407
Leukemia inhibitory factor receptor levels	IVW	4	0.884(0.676-1.155)		0.366
Monocyte chemoattractant protein-1 levels	IVW	4	1.034(0.775-1.380)		0.818
Monocyte chemoattractant protein 2 levels	IVW	9	0.970(0.883-1.065)	H-B-H	0.526
Monocyte chemoattractant protein-3 levels	IVW	5	0.891(0.654-1.212)	——	0.462
Monocyte chemoattractant protein-4 levels	IVW	7	1.048(0.862-1.273)	——	0.641
Macrophage inflammatory protein 1a levels	IVW	6	1.052(0.878-1.261)	⊢•	0.580
Matrix metalloproteinase-1 levels	IVW	12	0.963(0.652-1.423)		0.850
Matrix metalloproteinase-10 levels	IVW	8	0.885(0.763-1.027)	⊢	0.107
Neurotrophin-3 levels	Wald ratio	1	0.615(0.257-1.471)		0.275
Osteoprotegerin levels	IVW	4	1.255(0.934-1.686)	· · · · · · · · · · · · · · · · · · ·	0.132
Oncostatin-M levels	IVW	4	0.906(0.598-1.371)	<u> </u>	0.639
Programmed cell death 1 ligand 1 levels	Wald ratio	1	1.496(0.837-2.674)	•	0.174
Stem cell factor levels	IVW	11	0.994(0.763-1.295)	⊢	0.965
SIR2-like protein 2 levels	Wald ratio	1	0.756(0.400-1.426)	—	0.387
Signaling lymphocytic activation molecule levels	IVW	6	1.108(0.830-1.481)		0.486
Sulfotransferase 1A1 levels	IVW	3	1.186(0.906-1.551)	<u> </u>	0.214
Transforming growth factor-alpha levels	IVW	2	0.621(0.329-1.173)		0.142
TNF-beta levels	IVW	6	1.057(0.921-1.212)	⊢i• →	0.431
TNFRSF9 levels	IVW	3	1.168(0.582-2.346)		0.662
TNFRSF14 levels	IVW	4	1.141(0.891-1.461)	—	0.296
TNF-related apoptosis-inducing ligand levels	IVW	8	1.007(0.847-1.196)	-	0.941
TNF-related activation-induced cytokine levels	IVW	7	1.013(0.813-1.263)	──	0.908
Thymic stromal lymphopoietin levels	Wald ratio	1	0.514(0.200-1.323)		0.168
TWEAK levels	IVW	6	0.883(0.684-1.140)		0.341
Urokinase-type plasminogen activator levels	IVW	9	0.921(0.727-1.167)		0.494
Vascular endothelial growth factor A levels	IVW	5	1.073(0.894-1.287)	⊢	0.449

Figure 3. The causal association between between sepsis and circulating inflammatory factors when exposures were circulating inflammatory factors based on GWAS catalog. Inverse-variance weighting was regarded as the major method in this study. P value for heterogeneity based on Cochran's Q statistic for IVW. CI = confidence internal, GWAS = gene-wide association study, IVW = inverse variance weighted, MR = mendelian randomization, nSNP = numbers of single nucleotide polymorphism, OR = odds ratio.

95% CI = 0.785-1.209, P = .816, respectively). Furthermore, there was also no obvious heterogeneity (all P > .05). Leave-one out studies were used for sensitivity analysis and demonstrated no influence of individual studies.

3.3. Sensitivity analyses

The results produced by MR-Egger, simple mode, weighted mode, weighted median, and MR-PRESSO methodologies provided consistent estimates regarding both the magnitude and direction

Exposure	Method	nSNP	OR(95%CI)		P-value
EIF4EBP1 levels	IVW	3	0.942(0.807-1.100)		0.452
Adenosine Deaminase levels	IVW	3	1.039(0.983-1.098)	i je 1	0.176
beta-nerve growth factor levels	IVW	2	0.769(0.599-0.987)	-	0.039
Caspase 8 levels	Wald ratio	1	1.122(0.921-1.368)	1	0.252
Eotaxin levels	IVW	5	0.925(0.826-1.037)	H-H	0.181
C-C motif chemokine 19 levels	IVW	4	1.050(0.907-1.215)	7	0.512
C-C motif chemokine 20 levels	IVW	2	0.975(0.795-1.195)		0.804
C-C motif chemokine 23 levels C-C motif chemokine 25 levels	IVW	4 7	0.988(0.901-1.083)	1	0.795
C-C motif chemokine 28 levels	IVW	4	1.017(0.974-1.062)		0.441 0.573
C-C motif chemokine 4 levels	IVW	4	1.064(0.857-1.320) 0.979(0.925-1.035)		0.451
Natural killer cell receptor 2B4 levels	IVW	8	0.989(0.908-1.077)	1	0.451
CD40L receptor levels	IVW	4	0.988(0.926-1.054)	I	0.719
T-cell surface glycoprotein CD5 levels	IVW	6	0.970(0.870-1.082)	<u></u>	0.719
T-cell surface glycoprotein CD3 levels T-cell surface glycoprotein CD6 isoform levels	IVW	4	1.006(0.959-1.056)		0.808
CUB domain-containing protein 1 levels	IVW	5	0.949(0.871-1.034)	1-01	0.229
Macrophage colony-stimulating factor 1 levels	IVW	2	1.010(0.874-1.167)		0.892
Cystatin D levels	IVW	9	0.964(0.915-1.017)	i•∳	0.182
Fractalkine levels	IVW	4	1.133(0.950-1.353)		0.166
C-X-C motif chemokine 1 levels	Wald ratio	1	0.990(0.892-1.098)		0.846
C-X-C motif chemokine 10 levels	IVW	4	0.959(0.844-1.090)		0.524
C-X-C motif chemokine 11 levels	IVW	4	0.942(0.832-1.068)		0.351
C-X-C motif chemokine 5 levels	IVW	6	1.010(0.946-1.079)		0.759
C-X-C motif chemokine 6 levels	IVW	3	0.968(0.917-1.022)		0.739
C-X-C motif chemokine 9 levels	IVW	3	0.865(0.722-1.036)		0.246
DNER levels	IVW	4	1.039(0.926-1.166)		0.511
Protein S100-A12 levels	IVW	3	1.011(0.876-1.167)		0.879
Fibroblast growth factor 19 levels	IVW	3	0.966(0.841-1.110)		0.625
Fibroblast growth factor 21 levels	IVW	3	0.972(0.838-1.128)		0.710
Fibroblast growth factor 23 levels	IVW	2	0.997(0.788-1.263)		0.983
Fibroblast growth factor 5 levels	IVW	4	1.060(0.928-1.210)		0.391
Fms-related tyrosine kinase 3 ligand levels	IVW	8	0.932(0.859-1.011)	101	0.091
GDNF levels	IVW	5	0.979(0.877-1.094)		0.711
Hepatocyte growth factor levels	IVW	2	1.046(0.873-1.252)		0.627
Interleukin-10 levels	IVW	4	0.981(0.851-1.131)		0.791
Interleukin-10 receptor subunit beta levels	IVW	2	1.051(0.986-1.121)	101	0.127
Interredicti 10 receptor suburit beta levels	1000		1.001(0.000 1.121)		
				À 4	ò
				0 1	2
Exposure	Method	nSNP	OR(95%CI)		P-value
Interleukin-12 subunit beta levels	IVW	13	1.010(0.960-1.063)	0 1	P-value 0.698
Interleukin-12 subunit beta levels Interleukin-13 levels	IVW Wald ratio	13 1	1.010(0.960-1.063) 1.176(0.818-1.691)		P-value 0.698 0.380
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels	IVW Wald ratio IVW	13 1 4	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096)		P-value 0.698 0.380 0.337
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels	IVW Wald ratio IVW IVW	13 1 4 2	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077)		P-value 0.698 0.380 0.337 0.102
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels Interleukin-18 levels	IVW Wald ratio IVW IVW	13 1 4 2 4	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126)		P-value 0.698 0.380 0.337 0.102 0.607
Interleukin–12 subunit beta levels Interleukin–13 levels Interleukin–15 receptor subunit alpha levels Interleukin–17C levels Interleukin–18 levels interleukin–18 receptor 1 levels	IVW Wald ratio IVW IVW IVW	13 1 4 2 4 8	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058)		P-value 0.698 0.380 0.337 0.102 0.607 0.892
Interleukin–12 subunit beta levels Interleukin–13 levels Interleukin–15 receptor subunit alpha levels Interleukin–15 receptor subunit alpha levels Interleukin–18 levels Interleukin–18 receptor 1 levels Interleukin–1–alpha levels	IVW Wald ratio IVW IVW IVW IVW	13 1 4 2 4 8 2	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814
Interleukin–12 subunit beta levels Interleukin–13 levels Interleukin–15 receptor subunit alpha levels Interleukin–17 levels Interleukin–18 levels Interleukin–18 receptor 1 levels Interleukin–1-alpha levels Interleukin–20 receptor subunit alpha levels	IVW Wald ratio IVW IVW IVW IVW IVW IVW	13 1 4 2 4 8 2	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 1.013(0.909-1.129) 1.225(0.875-1.716)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-1-alpha levels Interleukin-20 receptor subunit alpha levels Interleukin-6 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Wald ratio	13 1 4 2 4 8 2 1	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269
Interleukin–12 subunit beta levels Interleukin–13 levels Interleukin–15 receptor subunit alpha levels Interleukin–17 C levels Interleukin–18 levels Interleukin–18 receptor 1 levels Interleukin–1a pla levels Interleukin–20 receptor subunit alpha levels Interleukin–8 levels Interleukin–8 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Wald ratio	13 1 4 2 4 8 2 1 1	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.990(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297
Interleukin–12 subunit beta levels Interleukin–13 levels Interleukin–15 receptor subunit alpha levels Interleukin–17C levels Interleukin–18 levels Interleukin–18 receptor 1 levels Interleukin–1-alpha levels Interleukin–20 receptor subunit alpha levels Interleukin–6 levels Interleukin–6 levels Interleukin–6 levels Latency TGF beta 1 levels	IVW Wald ratio IVW IVW IVW IVW VWW IVW Wald ratio Wald ratio IVW IVW	13 1 4 2 4 8 2 1 1 1 2 2	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.687-1.121) 0.927(0.785-1.095)		P-value 0.698 0.380 0.337 0.102 0.807 0.892 0.814 0.237 0.269 0.297
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-5 levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Lettency TGF beta 1 levels Leukemia inhibitory factor receptor levels	IVW Wald ratio IVW IVW IVW IVW VWW IVW VWald ratio IVW IVW IVW	13 1 4 2 4 8 2 1 1 2 2	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Wald ratio IVW IVW IVW	13 1 4 2 4 8 2 1 1 1 2 2 4 3	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-17C levels Interleukin-17C levels Interleukin-18 receptor 1 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-2 receptor subunit alpha levels Interleukin-8 levels Interleukin-8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein 2 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Wald ratio IVW IVW IVW IVW IVW IVW IVW IVW IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976
Interleukin–12 subunit beta levels Interleukin–13 levels Interleukin–15 receptor subunit alpha levels Interleukin–15 receptor subunit alpha levels Interleukin–17C levels Interleukin–18 receptor 1 levels Interleukin–18 receptor 1 levels Interleukin–1alpha levels Interleukin–2 receptor subunit alpha levels Interleukin–8 levels Interleukin–8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein–1 levels Monocyte chemoattractant protein–3 levels	IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.996(0.916-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.687-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels Interleukin-18 levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-6 levels Letaency TGF beta 1 levels Letaency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-4 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.126) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.211) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.146) 0.994(0.909-1.087)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-4 levels Macrophage inflammatory protein 1a levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW VWald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.999(0.934-1.069)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-17 Clevels Interleukin-17 Clevels Interleukin-18 receptor 1 levels Interleukin-18 receptor 1 levels Interleukin-18 receptor 1 levels Interleukin-20 receptor subunit alpha levels Interleukin-20 receptor subunit alpha levels Interleukin-8 levels Interleukin-8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Vald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.990(0.994-1.118) 1.001(0.966-1.037) 1.019(0.906-1.046) 0.994(0.909-1.087) 0.999(0.934-1.069) 0.978(0.880-1.088)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-18 levels Interleukin-18 levels Interleukin-18 levels Interleukin-18 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-6 levels Interleukin-7 levels Interleukin-7 levels Interleukin-8 levels Letancy TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-4 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-1 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7 7 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.984-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.994(0.908-1.087) 0.999(0.934-1.069) 0.978(0.880-1.088)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-15 receptor subunit alpha levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-10 receptor subunit alpha levels Interleukin-10 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.126) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.096) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.146) 0.994(0.909-1.087) 0.999(0.934-1.069) 0.978(0.880-1.086) 0.978(0.880-1.086) 0.964(0.888-1.046) 0.813(0.585-1.132)		P-value 0.698 0.380 0.387 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17 Clevels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels Osteoprotegerin levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW VWald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7 7 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.088) 0.964(0.888-1.046) 0.813(0.585-1.132) 1.065(0.910-1.245)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-17 Clevels Interleukin-18 receptor subunit alpha levels Interleukin-18 receptor I levels Interleukin-18 receptor I levels Interleukin-18 receptor I levels Interleukin-20 receptor subunit alpha levels Interleukin-20 receptor subunit alpha levels Interleukin-8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels Osteoprotegerin levels Oncostatin-M levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Wald ratio VW IVW IVW IVW IVW IVW IVW IVW IVW IVW	13 1 4 2 4 8 2 1 1 2 2 2 4 3 9 5 7 7 7 11 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.994(0.994-1.118) 1.001(0.966-1.037) 1.019(0.906-1.037) 1.019(0.906-1.087) 0.999(0.934-1.069) 0.978(0.880-1.088) 0.994(0.888-1.046) 0.813(0.585-1.1324) 1.065(0.910-1.245) 0.888(0.740-1.065)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17 Clevels Interleukin-18 levels Interleukin-18 levels Interleukin-18 levels Interleukin-1-alpha levels Interleukin-20 receptor subunit alpha levels Interleukin-20 receptor subunit alpha levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Letaneny TGF beta 1 levels Letaneny TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-3 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels Osteoprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Vwald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7 7 7 11 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.126) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.994(0.909-1.087) 0.999(0.934-1.069) 0.978(0.880-1.086) 0.978(0.888-1.046) 0.813(0.585-1.132) 1.065(0.910-1.245) 0.888(0.740-1.065) 1.023(0.821-1.276)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-15 receptor subunit alpha levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Interleukin-10 receptor subunit alpha levels Interleukin-10 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels Osteoprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels Stem cell factor levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 1 2 2 4 3 9 5 7 7 7 11 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.126) 0.913(0.909-1.127) 0.878(0.887-1.121) 0.927(0.785-1.076) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.960-1.146) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.888-1.046) 0.813(0.585-1.132) 1.065(0.910-1.245) 0.888(0.740-1.276) 0.988(0.740-1.276) 0.103(0.821-1.276) 0.959(0.882-1.042)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17 Clevels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-1 levels Neurotrophin-3 levels Osceprotegerin levels Osceprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels SIR2-like protein 2 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 1 2 2 2 4 3 9 5 7 7 7 11 7 1 4 4 4 4 4 1 1 1 1 1 1 1 1	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.994(0.909-1.087) 0.803(0.821-1.245) 0.888(0.740-1.065) 1.023(0.821-1.276) 0.959(0.882-1.0424) 0.807(0.635-1.024)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325 0.078
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-15 receptor subunit alpha levels Interleukin-18 levels Interleukin-18 levels Interleukin-18 levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-6 levels Interleukin-7 levels Interleukin-7 levels Interleukin-8 levels Leutency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Neurotrophin-3 levels Osteoprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels Stem cell factor levels Sit2-like protein 2 levels Signaling lymphocytic activation molecule levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio VWald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 2 4 3 9 5 7 7 7 11 7 1 4 4 4 1 1 1 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.990(0.994-1.118) 1.001(0.966-1.037) 1.019(0.906-1.046) 0.999(0.934-1.069) 0.978(0.880-1.088) 0.994(0.888-1.046) 0.813(0.585-1.1324) 1.085(0.910-1.245) 0.888(0.740-1.065) 1.023(0.821-1.276) 0.959(0.882-1.024) 1.0807(0.835-1.024)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325 0.095
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17C levels Interleukin-18 levels Interleukin-18 levels Interleukin-18 levels Interleukin-1-alpha levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Interleukin-1 levels Interleukin-1 levels Interleukin-1 levels Interleukin-1 levels Interleukin-20 receptor subunit alpha levels Interleukin-1 levels Interleukin-2 levels Interleukin-3 levels Interleukin-3 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels Osteoprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels SIR2-like protein 2 levels Signaling lymphocytic activation molecule levels Sulfotransferase 1A1 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7 7 11 4 4 1 10 1 5 2	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.126) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.21) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.18) 1.001(0.966-1.146) 0.994(0.909-1.087) 0.999(0.934-1.069) 0.978(0.880-1.089) 0.978(0.880-1.084) 0.810(0.888-1.046) 0.813(0.585-1.132) 1.055(0.910-1.245) 0.882(0.740-1.065) 1.023(0.821-1.276) 0.959(0.882-1.042) 0.807(0.835-1.024) 1.000(0.871-1.147) 0.982(0.810-1.191)		P-value 0.698 0.380 0.387 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325 0.788 0.989
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17 Clevels Interleukin-18 levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-1-alpha levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-8 levels Interleukin-10 receptor subunit alpha levels Interleukin-10 receptor subunit alpha levels Interleukin-10 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-1 levels Matrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels Oncostatin-M levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels Stem cell factor levels Signaling lymphocytic activation molecule levels Transforming growth factor-alpha levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 2 2 4 3 9 5 7 7 11 7 1 4 4 1 10 1 5 2 2 2	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.126) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.096) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.994(0.909-1.087) 0.999(0.934-1.069) 0.978(0.888-1.046) 0.813(0.585-1.132) 1.065(0.910-1.245) 0.888(0.740-1.265) 1.023(0.821-1.276) 0.959(0.882-1.042) 0.807(0.635-1.024) 1.009(0.835-1.024) 1.009(0.871-1.147) 0.992(0.810-1.191)		P-value 0.698 0.380 0.387 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325 0.078 0.995 0.888
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17 levels Interleukin-18 levels Interleukin-18 receptor I levels Interleukin-18 receptor I levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Latency TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-2 levels Monocyte chemoattractant protein-3 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-1 levels Osteoprotegerin levels Osteoprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels SIR2-like protein 2 levels Signaling lymphocytic activation molecule levels Sulfotransferase 1A1 levels Transforming growth factor-alpha levels TNF-beta levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio IVW	13 1 4 2 4 8 2 1 1 1 2 2 4 3 9 5 7 7 11 7 1 4 4 1 10 1 5 2 7	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.894-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.994(0.909-1.087) 0.999(0.934-1.069) 0.978(0.880-1.088) 0.964(0.888-1.046) 0.813(0.585-1.132) 1.065(0.910-1.245) 0.888(0.740-1.065) 1.023(0.821-1.276) 0.959(0.882-1.042) 1.000(0.871-1.147) 0.982(0.810-1.171) 0.982(0.810-1.171) 0.983(0.570-1.626) 0.990(0.871-1.147)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325 0.078 0.995 0.855 0.888 0.460
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-15 receptor subunit alpha levels Interleukin-18 levels Interleukin-18 levels Interleukin-18 receptor 1 levels Interleukin-20 receptor subunit alpha levels Interleukin-3 levels Interleukin-6 levels Interleukin-6 levels Interleukin-6 levels Letancy TGF beta 1 levels Letancy TGF beta 1 levels Leukemia inhibitory factor receptor levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metallioproteinase-1 levels Matrix metallioproteinase-1 levels Neurotrophin-3 levels Osteoprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels Stem cell factor levels Signaling lymphocytic activation molecule levels Sulfotransferase 1A1 levels Transforming grwh factor-alpha levels TNF-beta levels TNF-beta levels TNFRSF9 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Wald ratio IVW	13 1 4 2 4 8 2 1 1 2 4 3 9 5 7 7 1 1 4 1 10 1 5 2 2 7 3	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.129) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.984-1.118) 1.001(0.966-1.037) 1.019(0.906-1.146) 0.994(0.909-1.087) 0.999(0.934-1.093) 0.978(0.880-1.088) 0.964(0.888-1.046) 0.813(0.585-1.132) 1.065(0.910-1.245) 0.880(0.740-1.065) 1.023(0.821-1.276) 0.999(0.938-1.042) 0.807(0.835-1.024) 1.000(0.871-1.147) 0.982(0.810-1.191) 0.963(0.570-1.266) 0.990(0.928-1.034) 1.140(0.987-1.316)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325 0.995 0.855 0.888 0.460 0.075
Interleukin-12 subunit beta levels Interleukin-13 levels Interleukin-15 receptor subunit alpha levels Interleukin-17 Clevels Interleukin-18 levels Interleukin-18 levels Interleukin-18 levels Interleukin-1-alpha levels Interleukin-1-alpha levels Interleukin-1-alpha levels Interleukin-6 levels Interleukin-8 levels Interleukin-8 levels Interleukin-1 levels Interleukin-1 levels Interleukin-1 levels Interleukin-1 levels Interleukin-1 levels Interleukin-20 receptor subunit alpha levels Interleukin-1 levels Interleukin-3 levels Interleukin-3 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Monocyte chemoattractant protein-1 levels Macrophage inflammatory protein 1a levels Matrix metalloproteinase-1 levels Matrix metalloproteinase-10 levels Neurotrophin-3 levels Osteoprotegerin levels Oncostatin-M levels Programmed cell death 1 ligand 1 levels Stem cell factor levels SiR2-like protein 2 levels Signaling lymphocytic activation molecule levels Sulfotransferase 1A1 levels Transforming growth factor-alpha levels TNFSF9 levels TNFRSF14 levels	IVW Wald ratio IVW IVW IVW IVW Wald ratio Wald ratio IVW	13 1 4 2 4 8 2 1 1 1 2 2 4 3 9 5 7 7 1 4 4 1 10 1 5 2 2 7 3 4	1.010(0.960-1.063) 1.176(0.818-1.691) 1.031(0.969-1.096) 0.690(0.443-1.077) 0.959(0.816-1.126) 0.996(0.937-1.058) 1.013(0.909-1.126) 1.225(0.875-1.716) 0.913(0.777-1.073) 0.878(0.887-1.121) 0.927(0.785-1.095) 1.052(0.951-1.164) 0.999(0.984-1.118) 1.001(0.966-1.146) 0.994(0.909-1.087) 0.999(0.984-1.087) 0.999(0.984-1.087) 0.999(0.888-1.046) 0.813(0.585-1.132) 1.055(0.910-1.245) 0.888(0.740-1.065) 1.023(0.821-1.276) 0.959(0.882-1.042) 1.000(0.871-1.147) 0.982(0.810-1.191) 0.983(0.570-1.626) 0.980(0.928-1.034) 1.140(0.987-1.316) 1.050(0.956-1.153)		P-value 0.698 0.380 0.337 0.102 0.607 0.892 0.814 0.237 0.269 0.297 0.374 0.323 0.991 0.976 0.758 0.898 0.981 0.684 0.375 0.220 0.433 0.199 0.839 0.325 0.078 0.995 0.888 0.460 0.075 0.313
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Figure 4. The causal association between between sepsis and circulating inflammatory factors when exposures were circulating inflammatory factors based on IEU-open GWAS project. Inverse-variance weighting was regarded as the major method in this study. *P* value for heterogeneity based on Cochran's Q statistic for IVW. CI = confidence internal, GWAS = gene-wide association study, IVW = inverse variance weighted, MR = mendelian randomization, nSNP = numbers of single nucleotide polymorphism, OR = odds ratio.

1.027(0.938-1.123)

1.182(1.016-1.375)

IVW

of causality. MR-Egger's regression intercept approach revealed no significant evidence of horizontal pleiotropy relevant to the susceptibility of sepsis due to circulating inflammatory factors with P > .05. The absence of outliers detected by MR-PRESSO

Urokinase-type plasminogen activator levels

Vascular endothelial growth factor A levels

posits a lack of substantive evidence to affirm the presence of heterogeneity in the research outcomes. Furthermore, no significant heterogeneity was observed from the conclusions extracted from Cochrane's Q Statistics, with all P > .05 (Figs. 7 and 8).

0.568

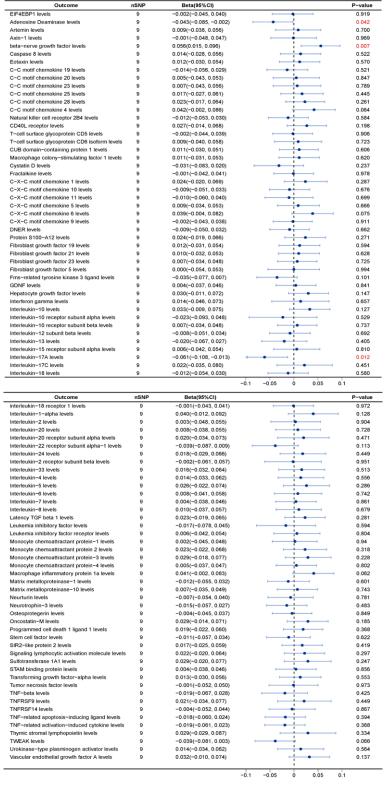


Figure 5. The causal association of sepsis on circulating levels of systemic inflammatory regulators based on GWAS catalog. Inverse-variance weighting was regarded as the major method in this study. P value for heterogeneity based on Cochran's Q statistic for IVW. CI = confidence internal, GWAS = gene-wide association study, IVW = inverse variance weighted, MR = mendelian randomization, nSNP = numbers of single nucleotide polymorphism, OR = odds ratio.

4. Discussion

In this bidirectional two-sample Mendelian randomization study, we employed pooled GWAS data from a European demographic to scrutinize the causal relationships between 91 biomarkers and sepsis. The results unequivocally established a causal relationship between the genetically predetermined levels of systemic inflammatory regulators and sepsis susceptibility.

Our analysis revealed an inverse relationship between increased beta-nerve growth factor (BNGF) levels and sepsis incidence. NGF, a polypeptide indispensable to normal neural

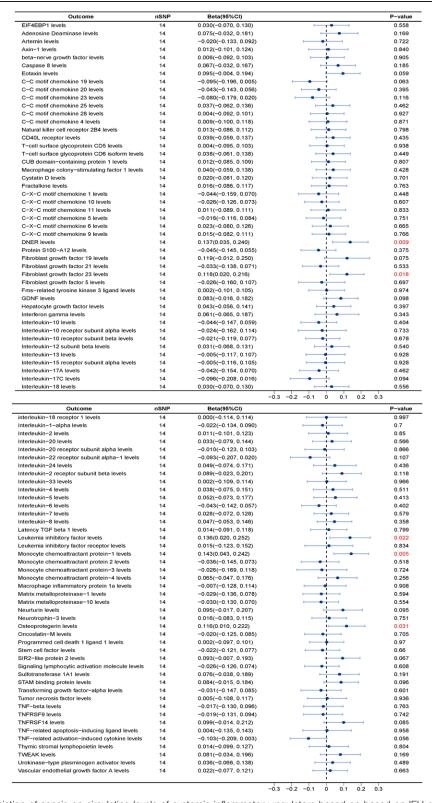
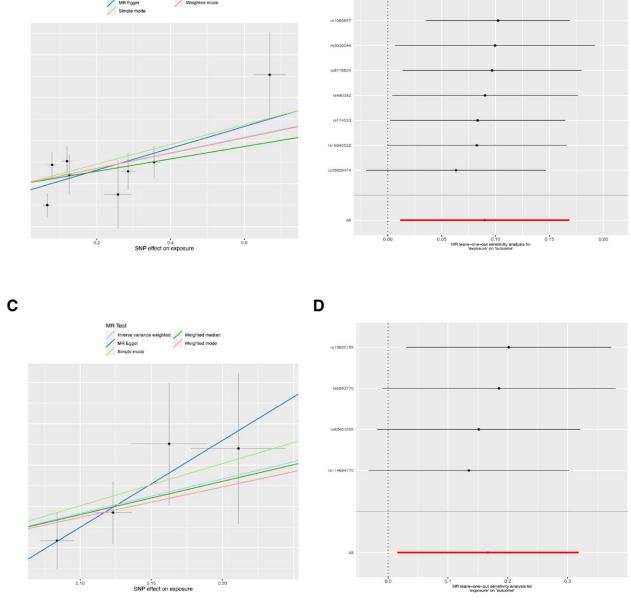


Figure 6. The causal association of sepsis on circulating levels of systemic inflammatory regulators based on based on IEU-open GWAS project. Inverse-variance weighting was regarded as the major method in this study. P value for heterogeneity based on Cochran's Q statistic for IVW. CI = confidence internal, GWAS = gene-wide association study, IVW = inverse variance weighted, MR = mendelian randomization, nSNP = numbers of single nucleotide polymorphism, OR = odds ratio.

development, fosters the sustenance and differentiation of sensory and sympathetic neurons in culture.^[29] Additionally, it has been associated with cognitive alterations following brain damage.^[30] Sepsis-induced dysfunction of the central nervous system

often results from a localized generation of inflammatory cytokines, cerebral microcirculation disruption, imbalances in neurotransmitters, and apoptosis. [31] Our study uncovers an inverse causality between B-NGF and sepsis, linking an augmented risk

Α



В

Figure 7. Scatter plots and Leave-one-out plots of circulating levels of systemic inflammatory regulators related SNPs with the risk of sepsis when outcome was spesis. Red lines in leave-one-out plots represent estimations from the IVW test. (A) and (B): Genetic association of TNF-related apoptosis-inducing ligand levels related SNPs and sepsis. (C) and (D): Genetic association of vascular endothelial growth factor A level related SNPs and sepsis. IVW = inverse variance weighted, SNPs = single nucleotide polymorphisms.

of sepsis with decreased circulating BNGF levels, which is also consistent with previous research results. Previous research has unearthed a positive correlation between expansive involutive changes, heightened distribution, and increased NGF immunoreactivity in the mast cells derived from autopsy thymus specimens in pediatric cases of sepsis. [32] Animal studies further demonstrate that elevated NGF levels catalyze cellular activity and expedite bone tissue regeneration.[33] Reports have underscored mitochondrial impairment and reduced NGF levels in the hippocampus of septic diabetic patients. [34] These findings indirectly substantiate our conclusion that genetically predicted higher β-NGF levels may escalate sepsis risk, potentially attributable to its association with excessive inflammatory responses and cellular damage during sepsis. Despite the clear role of NGF in the interplay between the immune and nervous systems, a comprehensive understanding of NGF's influence on immune cells awaits further exploration.^[35] While our MR analysis suggests a protective role of BNGF, experimental studies are needed to dissect its dual effects – for instance, whether BNGF modulates neuroinflammation directly or via cross-talk with peripheral immune cells during sepsis.^[35]

TNF-related apoptosis-inducing ligand (TRAIL), a cytokine part of the TNF superfamily, triggers apoptosis in transformed or tumor cells upon binding to death receptors 4 or 5, thus playing a pivotal role in sepsis' immune responses. [36,37] TRAIL's multifaceted nature in sepsis involves inducing apoptosis in tissue-invading neutrophils, facilitating organ protection against sepsis-induced damage. [38–40] Our investigation found an increasing trend in sepsis development corresponding to elevated circulating TRAIL levels. Importantly, another study reported an association between plasma TRAIL and worsening prognosis in patients with sepsis. [41] However, a deviation was noted in

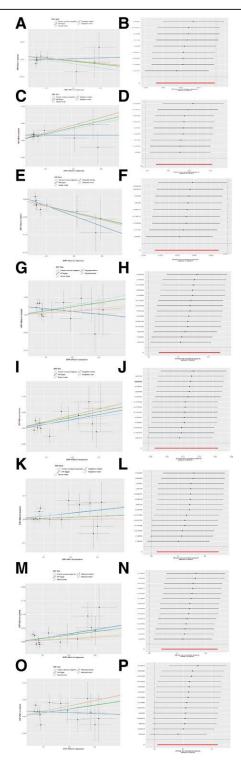


Figure 8. Scatter plots and Leave-one-out plots of circulating levels of systemic inflammatory regulators related SNPs with the risk of sepsis when outcome were circulating levels of systemic inflammatory regulators. Red lines in leave-one-out plots represent estimations from the IVW test. (A) and (B): Genetic association of adenosine dearninase levels related SNPs and sepsis. (C) and (D): Genetic association of Interieukin-17A (IL-17A) level related SNPs and sepsis. (E) and (F): Genetic association of beta-nerve growth factor level related SNPs and sepsis. (G) and (J): Genetic association of DNER level related SNPs and sepsis. (I) and (J): Genetic association of Fibroblast growth factor 23 level related SNPs and sepsis. (K) and (L): Genetic association of Leukemia inhibitory factor level related SNPs and sepsis. (M) and (N): Genetic association of Monocyte chemoattractant protein-1 level related SNPs and sepsis. (O) and (P): Genetic association of Osteoprotegerin level related SNPs and sepsis. IVW inverse variance weighted, SNPs = single nucleotide polymorphisms.

another recent study where researchers identified a relationship between lower TRAIL levels and septic shock and organ dysfunction across 3 independent Intensive Care Unit cohorts. [42] Likewise, Beyer et all [43] reported that while exogenous and endogenous TRAIL was protective in early sepsis, endogenous TRAIL appeared to have a harmful effect in the later stages. Such contrasting viewpoints emphasize the necessity for a more profound understanding of TRAIL's underlying mechanisms in sepsis progression. This notion underscores the inherent complexity of sepsis and fuels the ongoing pursuit to unravel its multifarious aspects. Further research is needed to determine the precise role of TRAIL in sepsis and its potential as a therapeutic target.

Our research determined that increased plasma levels of vascular endothelial growth factor A (VEGF-A) are associated with a heightened risk of sepsis. Sepsis-induced inflammatory responses, complement activation, and coagulation characteristics can trigger severe endothelial dysfunction.[44] This dysfunction could subsequently provoke disorders in hemostasis and vascular reactivity, and tissue edema, [45] thereby exacerbating sepsis severity. VEGF-A impacts several endothelial cell properties, inclusive of Nitrous Oxide and prostacyclin production. [46] It also can induce the production and release of multiple cytokines, such as TNF, NF-κB, IL-4, IL-6, Monocyte chemotactic protein-1 (MCP-1), among others. [47,48] Some studies have noticed an upsurge in VEGF-A and a downswing in VEGF-B in septic children compared to controls. [49-51] One study in adults with sepsis linked elevated serum VEGF to worsening tissue perfusion and oxygenation, and significant influence on tissue fluid accumulation.[52] Additionally, a clinical study revealed septic children exhibiting specific higher concentration levels of IL-6, vascular endothelial growth factor (VEGF), and its soluble decoy receptor II.[53] A multicenter observational clinical trial suggested that dogs with naturally occurring sepsis and organ dysfunction had higher average concentrations of endothelial activation and inflammation biomarkers compared to their healthy counterparts.^[54] These research endeavors underscore the pronounced elevation of VEGF, particularly VEGF-A, in sepsis and its proinflammatory role. These insights lend credence to its potential utility as a therapeutic target in managing sepsis. Due to the limitations of Mendelian randomization research, future studies should further explore potential mechanisms, mechanistic studies using sepsis models (e.g., cecal ligation and puncture in rodents) are essential to validate these genetic associations and explore therapeutic modulation of key cytokines like TRAIL and VEGF-A.

Adenosine deaminase acting on RNA 1 (ADAR1), a doublestranded RNA-editing enzyme responsible for converting adenosine (A) to inosine (I), plays a significant role in regulating immune responses. [55] A recent study unveiled ADAR1's hitherto unknown protective effect in maintaining intestinal homeostasis. [56] In line with this, Zhao et al [57] demonstrated the regulatory roles of adenosine deaminase in sepsis pathology. Additionally, ADAR1 has been reported to attenuate inflammation and organ damage via the ADAR1-Mir-30a-SOCS3 axis in a mouse model, thereby serving as a protective agent in sepsis.^[58] An in vitro study further established ADAR1's role in mitigating IL-1β-induced endothelial activation to prevent sepsis exacerbation.^[59] Our study found an inverse correlation between adenosine deaminase and sepsis. However, given the multifaceted nature of downstream targeted molecules of RNA-editing enzyme ADAR1 and miRNA, the potential for other regulatory pathways influencing sepsis' evolution and progression cannot be ruled out. Therefore, further research is warranted to elucidate the precise relationship between adenosine deaminase and sepsis, it should focus on identifying novel ADAR1-dependent pathways and exploring the therapeutic potential of targeting ADAR1 to modulate immune responses in sepsis.

Interleukin (IL)-17A, part of the IL-17 family, may initially reduce the inflammatory response, [60] but subsequently has been reported to interact with specific inflammatory cytokines to amplify inflammation.^[61] IL-17A has been implicated in spearheading neutrophil recruitment, host defense, and inflammation, thereby instigating tissue damage and aiding sepsis progression. This has been evidenced by increased IL-17A levels observed in the plasma and tissues of septic animal models. [62,63] Zhao et al [64] have noted a significant increase in IL-17 signaling pathwayrelated genes in the blood samples of septic patients compared to an age-matched healthy control group. Similarly, another clinical study^[65] pointed to a link between heightened IL-17A levels and acute kidney injury in septic patients, further associated with greater renal damage and mortality. Additionally, Mikacenic et al^[66] suggested that an uptick in circulating IL-17A levels might signal the onset of Acute Respiratory Distress Syndrome. However, contradictorily, our inverse MR analysis indicated that IL-17A levels inversely corresponded with sepsis incidence. In line with our findings, animal studies^[67] have shown that exogenous IL-17A can mitigate the harmful inflammatory response, enhancing the survival rate of septic mice. This apparent discrepancy emphasizes the need for further welldesigned trials to comprehensively explore the role of IL-17A in sepsis.

DNER is a transmembrane protein that potentially serves as oncogenic or antioncogenic factors by modulating cellular proliferation, invasion, and metastasis. [68] However, few studies have explored its link to sepsis. Our research identified a potential positive relationship between DNER and sepsis. A cohort study noted that DNER was downregulated in patients with severe infection. [69] These findings suggest that DNER could be involved in lymphopoiesis and apoptosis under pathological conditions. Additionally, an animal experiment highlighted that DNER plays a crucial role in mediating itouch cell-cell interactions and maintaining glucose homeostasis.^[70] The possibility of DNER influencing the progression and remission of sepsis through these mechanisms merits further investigation. Future research should focus on elucidating the molecular pathways through which DNER affects sepsis progression and identifying potential therapeutic targets.

In conditions such as sepsis and autoimmune diseases, inflammation is frequently correlated with an upsurge in the fibroblast growth factor (FGF23).[71] As a bone-derived hormone, FGF23 is involved in a positive feedback loop with inflammation - FGF23 stimulates the production of pro-inflammatory cytokines, which in turn induce more FGF23 production.^[72] Several studies have drawn a connection between elevated levels of circulating FGF23 and the activation of inflammatory cells in the liver. [73,74] Inflammatory cytokines are directly responsible for inducing FGF23 production in bones and osteoblast/ osteocyte lines. Sepsis patients often exhibit transient hypophosphatemia, indicating regulation of FGF23 levels by proinflammatory factors.^[75] It has been shown that pro-inflammatory stimuli can amplify the secretion of FGF23 by osteocytes. [76] In severe conditions, such as septic shock and myocardial injury, an escalation in inflammation is often accompanied by an abnormal increase in circulating FGF23 levels.[77] However, the biological significance of elevated FGF23 in the context of sepsis remains undefined. Nevertheless, due to its key role in correcting 1,25(OH)2D deficiency, [78] FGF23 may present therapeutic possibilities for enhancing survival rates in sepsis patients. Further research is warranted to explore this hypothesis and elucidate FGF23's functions in sepsis.

We identified a positive correlation between leukemia inhibitory factor (LIF) and sepsis. LIF, an interleukin-6 cytokine family member, mediates a variety of central nervous system (CNS) responses to inflammatory stimuli.^[79,80] In patients with sepsis and septic shock, elevated circulating LIF levels were observed^[81,82] and these levels were found to correlate with disease severity.^[83] In animal models,^[84] administering LIF was

shown to mitigate the severity of sepsis and septic shock caused by live Escherichia coli infection. Another animal experiment revealed that endogenous LIF enhances the expression of acute phase proteins and the production of IL-10, thus reducing the synthesis and release of TNF- α , providing some protection against sepsis. These findings suggest that changes in LIF are part of the host response to tissue damage induced by endotoxin and sepsis, offering new insights for future exploration of underlying mechanisms. Future research should focus on elucidating the specific pathways through which LIF exerts its protective effects in sepsis and exploring its potential as a therapeutic agent.

MCP-1, also known as chemokine ligand 2, is a proinflammatory chemokine involved in the recruitment and activation of monocytes and macrophages.[86] Our reverse MR analysis observed a positive correlation between MCP-1 and sepsis mortality. Previous studies[87,88] delineated a relationship between plasma levels of inflammatory cytokines, MCP-1, and outcomes in adults and children with sepsis. Inhibiting MCP-1 or specific MCP-1 antagonists curbed the release of TNF-α, IL-1β, and IL-6 from macrophages.[89] Another clinical study[90] associated the MCP-1/chemokine ligand 2 polymorphisms rs1024611 and rs2857656 with sepsis susceptibility and development. MCP-1 triggers the conversion of blood monocytes from anti-inflammatory IL-10 producing cells to pro-inflammatory TNF-α/IL-6 secreting cells,^[91] and its blockers have demonstrated protective effects in animal models of sepsis. [92] Given these results, we propose an anti-MCP-1 strategy [93] for managing sepsis and endotoxin levels, as it might have significant therapeutic implications. Future research should focus on developing and testing MCP-1 inhibitors in clinical trials to determine their efficacy in reducing sepsis mortality.

Osteoprotegerin (OPG) is a soluble protein that, as consistent with our findings, exhibited higher serum levels in some patients with sepsis or septic shock.^[94] High levels of OPG were linked to poor sepsis prognosis,^[95] which may be partly attributed to OPG's ability to augment inflammation by inhibiting the receptor activator of NF-kappaB.^[96] The Osteoprotegerin Ligand (OPGL), a member of the tumor necrosis factor ligand superfamily, has been implicated in T-cell and dendritic cell interactions. Rat experiments have corroborated this, with disease development being suppressed in monocyte/macrophagemediated conditions following administration of receptor fusion proteins that block OPGL activity.^[97] Our study provides additional genomic evidence for the link between osteoprotegerin and sepsis.

Our study highlights the potential of certain cytokines as biomarkers that could transform the way sepsis is diagnosed and managed. The identified associations between cytokine levels and sepsis risk offer a promising avenue for developing new diagnostic tools. By integrating these cytokine profiles into routine clinical assessments, healthcare providers can achieve earlier detection of sepsis, enabling timely interventions that could significantly improve patient outcomes. Additionally, the cytokines identified in our study, such as BNGF, TRAIL, and VEGF-A, may serve as therapeutic targets. Modulating their levels could reduce sepsis risk or severity, presenting a novel therapeutic strategy. This approach requires further investigation through clinical trials to validate efficacy and safety in diverse patient populations. To translate these findings into clinical practice, a few pathways could be considered. Firstly, collaborative efforts between researchers, clinicians, and industry partners are essential for the development and validation of cytokine-based diagnostic assays. Standardization of these assays across healthcare settings will be crucial to ensure reliability and reproducibility. Secondly, pilot studies and clinical trials should be initiated to explore therapeutic interventions targeting specific cytokines, focusing on optimizing treatment regimens and patient selection criteria. By advancing these strategies, the insights gained from this study can contribute to more personalized and effective management of sepsis, ultimately reducing the burden of this complex and often life-threatening condition.

While this study provides valuable insights into the potential

causal relationship between systemic inflammatory regulators

and sepsis, several limitations must be acknowledged. First, our reliance on summary statistics, rather than individuallevel data, limited our ability to conduct more granular analyses, such as delineating the different etiologies of sepsis or exploring non-linear genetic associations. In our research, the use of summary level GWAS data can improve the statistical efficacy, and it is relatively easy to obtain summary level GWAS data, which can save time and resources. However, the disadvantage is that it depends on the effectiveness of GWAS results reported by other research teams. If there are deviations or errors in these data, it may affect the results of MR analysis, and the use of individual level data can carry out more comprehensive analysis, such as nonlinear MR analysis or analysis of specific subgroups (such as only in smokers). The use of aggregated data limits these more in-depth analyses. In addition, the double sample design requires that 2 samples represent similar basic populations. If the age, sex or other characteristics of the 2 samples are different, the reliability of the instrumental variables may be affected, and there may be overlapping cases in the 2 samples, which will also make the MR estimation biased in the direction of observation correlation, especially when the correlation between genetic and risk factors is not strong. These limitations prompt us to improve in future research. Our inability to stratify sepsis cases by infection source (e.g., pulmonary vs abdominal) or clinical severity scores (e.g., SOFA, APACHE II) due to data availability constraints represents a critical limitation. Future studies with access to granular clinical metadata should prioritize stratified MR analyses to explore subtype-specific causal relationships. Future research could benefit from access to raw data, which would enable more intricate analyses and potentially uncover additional insights. Moreover, the potential for horizontal pleiotropy, where genetic variants influence the outcome through pathways other than the exposure of interest, presents a concern. Although we employed several sensitivity analyses and included a broad array of genetic variants as instrumental variables, identifying and accounting for pleiotropic effects remains a challenge and limits causal inference. Another significant limitation is our inability to dissect subtype-specific effects of sepsis. This is critical given the heterogeneous nature of sepsis, influenced by diverse sources of infection, variations in host genetics, and comorbidities. Subtype-specific analyses could provide more targeted insights and are an important direction for future studies. The population stratification bias is another concern due to the predominance of European ancestry in the GWAS data. This overrepresentation may limit the findings' generalizability to other ethnic groups, highlighting the need for more diverse genetic data in future studies to ensure broader applicability of the results. Future multicenter and multinational collaborations that would include diverse populations to validate our findings across different ethnic backgrounds, particularly in regions with high sepsis burden (e.g., Asia and Africa). Additionally, our focus on circulating cytokine levels, while informative, potentially overlooks significant contributions from local tissue cytokine environments, which may more accurately reflect the inflammatory processes associated with sepsis. [98] Integrated approaches that consider both systemic and local inflammatory responses could further elucidate the complex nature of cytokine involvement in sepsis. Lastly, socio-economic status (SES)[99] is a confounding factor not adequately addressed in this study. SES can affect baseline cytokine levels and sepsis risk through various pathways, yet our study did not find direct associations between SES, the exposures, and the outcomes.[100] Future MR studies should consider stratifying by SES to better understand its potential moderating effects.

In conclusion, addressing these limitations in future research efforts will be crucial to strengthen the understanding of the causal pathways linking systemic inflammatory regulators to sepsis and enhance the generalizability and clinical relevance of such findings across diverse populations. Our findings highlight several cytokines (e.g., TRAIL, VEGF-A) as potential biomarkers for sepsis risk stratification. Integrating these biomarkers into existing diagnostic frameworks (e.g., qSOFA) could improve early detection in highrisk populations. Furthermore, therapies targeting MCP-1 or OPG pathways – already under investigation in autoimmune diseases – may be repurposed for sepsis management. Collaborative efforts between geneticists, clinicians, and pharmaceutical developers are critical to translate these insights into clinical trials.

5. Conclusions

In conclusion, our study found that decreased levels of BNGF, TRAIL, and VEGF-A were inversely associated with sepsis risk. On the other hand, lower circulating levels of adenosine deaminase and IL-17A, coupled with higher circulating levels of BNGF, DNER, FGF23, LIF, MCP-1, and OPG were linked to a heightened risk of sepsis. These findings underscore the pivotal role of cytokines in the pathogenesis of sepsis. Consequently, regulating these inflammatory factors and intervening therapeutically might be a promising strategy for both future treatment and prevention of sepsis. However, further studies are required to confirm whether these biomarkers can indeed be harnessed for sepsis prevention or treatment. Our findings contribute foundational knowledge to the field of sepsis research, emphasizing the need for comprehensive exploration of cytokine interactions. By addressing these limitations and translating these findings into clinical applications, future efforts can enhance the efficacy of sepsis management, paving the way for personalized diagnostic and therapeutic strategies.

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References

- [1] Liu D, Huang SY, Sun JH, et al. Sepsis-induced immunosuppression: mechanisms, diagnosis and current treatment options. Mil Med Res. 2022-9-56
- [2] Huang M, Cai S, Su J. The Pathogenesis of Sepsis and Potential Therapeutic Targets. Int J Mol Sci. 2019;20:5376.

- [3] Cecconi M, Evans L, Levy M, Rhodes A. Sepsis and septic shock. Lancet. 2018;392:75–87.
- [4] Yoshimura A, Ito M, Chikuma S, Akanuma T, Nakatsukasa H. Negative regulation of cytokine signaling in immunity. Cold Spring Harb Perspect Biol. 2018;10:a028571.
- [5] Doganyigit Z, Eroglu E, Akyuz E. Inflammatory mediators of cytokines and chemokines in sepsis: from bench to bedside. Hum Exp Toxicol. 2022;41:9603271221078871.
- [6] Gharamti AA, Samara O, Monzon A, et al. Proinflammatory cytokines levels in sepsis and healthy volunteers, and tumor necrosis factoralpha associated sepsis mortality: a systematic review and meta-analysis. Cytokine. 2022;158:156006.
- [7] Fajgenbaum DC, June CH. Cytokine Storm. N Engl J Med. 2020;383:2255–73.
- [8] Vu HH, Moellmer SA, McCarty OJT, Puy C. New mechanisms and therapeutic approaches to regulate vascular permeability in systemic inflammation. Curr Opin Hematol. 2025;32:130–7.
- [9] Grondman I, Pirvu A, Riza A, Ioana M, Netea MG. Biomarkers of inflammation and the etiology of sepsis. Biochem Soc Trans. 2020;48:1–14.
- [10] Frimpong A, Owusu EDA, Amponsah JA, et al. Cytokines as potential biomarkers for differential diagnosis of sepsis and other non-septic disease conditions. Front Cell Infect Microbiol. 2022;12:901433.
- [11] Visscher PM, Wray NR, Zhang Q, et al. 10 Years of GWAS Discovery: biology, function, and translation. Am J Hum Genet. 2017;101:5–22.
- [12] Burgess S, Timpson NJ, Ebrahim S, Davey Smith G. Mendelian randomization: where are we now and where are we going? Int J Epidemiol. 2015;44:379–88.
- [13] Bowden J, Holmes MV. Meta-analysis and Mendelian randomization: a review. Res Synth Methods. 2019;10:486–96.
- [14] Birney E. Mendelian Randomization. Cold Spring Harb Perspect Med. 2022;12:a041302.
- [15] Buniello A, MacArthur JAL, Cerezo M, et al. The NHGRI-EBI GWAS Catalog of published genome-wide association studies, targeted arrays and summary statistics 2019. Nucleic Acids Res. 2019;47:D1005–12.
- [16] Zhao JH, Stacey D, Eriksson N, et al; Estonian Biobank Research Team. Genetics of circulating inflammatory proteins identifies drivers of immune-mediated disease risk and therapeutic targets [published correction appears in Nat Immunol. 2023 Sep 7;:]. Nat Immunol. 2023;24:1540–51.
- [17] Matzaraki V, Le KTT, Jaeger M, et al. Inflammatory protein profiles in plasma of candidaemia patients and the contribution of host genetics to their variability. Front Immunol. 2021;12:662171.
- [18] Abecasis GR, Altshuler D, Auton A, et al; 1000 Genomes Project Consortium. A map of human genome variation from population-scale sequencing [published correction appears in Nature. 2011 May 26;473(7348):544. Xue, Yali [added]; Cartwright, Reed A [added]; Altshuler, David L [corrected to Altshuler, David]; Kebbel, Andrew [corrected to Keebler, Jonathan]; Koko-Gonzales, Paula [corrected to Kokko-Gonzales, Paula]; Nickerson, Debbie A [corrected to Nickerson, Debo]. Nature. 2010;467:1061–73.
- [19] Bycroft C, Freeman C, Petkova D, et al. The UK Biobank resource with deep phenotyping and genomic data. Nature. 2018;562:203–9.
- [20] Li J, Xiang L, Chen X, et al. Global, regional, and national burden of neonatal sepsis and other neonatal infections, 1990-2019: findings from the Global Burden of Disease Study 2019. Eur J Pediatr. 2023;182:2335–43.
- [21] Zekavat SM, Lin S-H, Bick AG, et al; Biobank Japan Project. Hematopoietic mosaic chromosomal alterations increase the risk for diverse types of infection. Nat Med. 2021;27:1012–24.
- [22] Jiang L, Zheng Z, Fang H, Yang J. A generalized linear mixed model association tool for biobank-scale data. Nat Genet. 2021;53:1616–21.
- [23] Georgakis MK, de Lemos JA, Ayers C, et al. Association of circulating monocyte chemoattractant protein-1 levels with cardiovascular mortality: a meta-analysis of population-based studies. JAMA Cardiol. 2021;6:587–92.
- [24] Bowden J, Del Greco MF, Minelli C, Davey Smith G, Sheehan NA, Thompson JR. Assessing the suitability of summary data for two-sample Mendelian randomization analyses using MR-Egger regression: the role of the I² statistic. Int J Epidemiol. 2016;45:1961–74.
- [25] Hartwig FP, Davies NM, Hemani G, Davey Smith G. Two-sample Mendelian randomization: avoiding the downsides of a powerful, widely applicable but potentially fallible technique. Int J Epidemiol. 2016;45:1717–26.
- [26] Bowden J, Del Greco M F, Minelli C, Davey Smith G, Sheehan N, Thompson J. A framework for the investigation of pleiotropy in two-sample summary data Mendelian randomization. Stat Med. 2017;36:1783–802.

- [27] Burgess S, Thompson SG. Interpreting findings from Mendelian randomization using the MR-Egger method [published correction appears in Eur J Epidemiol. 2017 Jun 29;:]. Eur J Epidemiol. 2017;32:377–89.
- [28] Bowden J, Davey Smith G, Haycock PC, Burgess S. Consistent estimation in mendelian randomization with some invalid instruments using a weighted median estimator. Genet Epidemiol. 2016;40:304–14.
- [29] Chesa PG, Rettig WJ, Thomson TM, Old LJ, Melamed MR. Immunohistochemical analysis of nerve growth factor receptor expression in normal and malignant human tissues. J Histochem Cytochem. 1988;36:383–9.
- [30] Wahlstrom G, Archer T, Larkfors L. Nerve growth factor (NGF plays) in rat brain following long-term barbital treatment: relation to convulsions and cognitive function. Neurosci Lett. 1992;137:65–8.
- [31] Comim CM, Silva NC, Mina F, et al; Evaluation of NCS-1, DARPP-32. and neurotrophins in hippocampus and prefrontal cortex in rats submitted to sepsis. Synapse. 2014;68:474–9.
- [32] Marinova T, Philipov S, Aloe L. Nerve growth factor immunoreactivity of mast cells in acute involuted human thymus. Inflammation. 2007;30:38–43.
- [33] Jiao Y, Liu Y, Li X, et al. Cortical perfection promotions bone regeneration by enhancing nerve growth factor secretion. Biochem Biophysics Res Commun. 2025;755:151562.
- [34] de Souza Stork S, Hubner M, Biehl E, et al. Diabetes exacerbates sepsis-induced neuroinflammation and brain mitochondrial dysfunction. Inflammation. 2022;45:2352–67.
- [35] Minnone G, De Benedetti F, Bracci-Laudiero L. NGF and its receptors in the regulation of inflammatory response. Int J Mol Sci. 2017;18:1028.
- [36] Condotta SA, Cabrera-Perez J, Badovinac VP, Griffith TS. T-cell-mediated immunity and the role of TRAIL in sepsis-induced immuno-suppression. Crit Rev Immunol. 2013;33:23–40.
- [37] Janssen EM, Droin NM, Lemmens EE, et al. CD4+ T-cell help controls CD8+ T-cell memory via TRAIL-mediated activation-induced cell death. Nature. 2005;434:88–93.
- [38] Berg AK, Hahn EM, Speichinger-Hillenberg F, et al. The Impact of TRAIL on the immunological milieu during the early stage of abdominal sepsis. Cancers (Basel). 2023;15:1773.
- [39] Chen YF, Chen GY, Chang CH, et al. TRAIL encapsulated to polypeptidecrosslinked nanogel exhibits increased anti-inflammatory activities in Klebsiella pneumoniae-induced sepsis treatment. Mater Sci Eng C Mater Biol Appl. 2019;102:85–95.
- [40] Beyer K, Poetschke C, Partecke LI, et al. TRAIL induces neutrophil apoptosis and dampens sepsis-induced organ injury in murine colon ascendens stent peritonitis. PLoS One. 2014;9:e97451.
- [41] Tian Y, Tao T, Zhu J, et al. Soluble tumor necrosis factor related apoptosis inducing ligand level as a predictor of severity of sepsis and the risk of mortality in septic patients. PLoS One. 2013;8:e82204.
- [42] Schenck EJ, Ma KC, Price DR, et al. Circulating cell death biomarker TRAIL is associated with increased organ dysfunction in sepsis. JCI Insight, 2019;4:e127143.
- [43] Beyer K, Stollhof L, Poetschke C, et al. TNF-related apoptosisinducing ligand deficiency enhances survival in murine colon ascendens stent peritonitis. J Inflamm Res. 2016;9:103–13.
- [44] Wiesinger A, Peters W, Chappell D, et al. Nanomechanics of the endothelial glycocalyx in experimental sepsis. PLoS One. 2013;8:e80905.
- [45] Chelazzi C, Villa G, Mancinelli P, De Gaudio AR, Adembri C. Glycocalyx and sepsis-induced alterations in vascular permeability. Crit Care. 2015;19:26. in Chinese.
- [46] Simoes JS, Rodrigues RF, Zavan B, Emidio RMP, Soncini R, Boralli VB. Endotoxin-induced sepsis on ceftriaxone-treated rats' ventilatory mechanics and pharmacokinetics. Antibiotics (Basel). 2024;13:83.
- [47] Gorenjak V, Vance DR, Petrelis AM, et al. Peripheral blood mononuclear cells extracts VEGF protein levels and VEGF mRNA: associations with inflammatory molecules in a healthy population [published correction appears in PLoS One. 2019 Oct 24; 14(10):e0224591]. PLoS One. 2019;14:e0220902.
- [48] Jing Y, Ding M, Fu J, Xiao Y, Chen X, Zhang Q. Neutrophil extracellular trap from Kawasaki disease alter the biologic responses of PBMC. Biosci Rep. 2020;40:bSR20200928.doi.
- [49] Cakmakecis S, Sari N, Sonmez C. The role of proangiogenic cytokines in predicting sepsis in febrile neutropenic children with cancer. Turk J Pediatr. 2024;66:90–8.
- [50] Rashid A, Brusletto BS, Al-Obeidat F, et al. A transcriptomic appreciation of childhood meningococcal and polymicrobial sepsis from A pro-inflammatory and trajectorial perspective, A role for vascular endothelial growth factor A and B modulation? Shock. 2023;60:503–16.
- [51] Janec P, Mojžíšek M, Panek M, Haluzik M, Živny J, Janota J. Earlyonset neonatal sepsis: inflammatory biomarkers and MicroRNA as

- potential diagnostic tools in preterm newborns. Folia Biol (Praha). 2023;69:173-80.
- [52] Lin CK, Tsai YH, Kao KC, et al. Serum vascular endothelial growth factor affects tissue fluid accumulation and is associated with deteriorating tissue perfusion and oxygenation in severe sepsis: a prospective observational study. Eur J Med Res. 2023;28:155.
- [53] Salgado DM, Rivera GM, Pinto WA, et al. Unique immune blood markers between severe dengue and sepsis in children. Pediatr Infect Dis J. 2023;42:792–800.
- [54] Gaudette S, Smart L, Woodward AP, et al. Biomarkers of endothelial activation and inflammation in dogs with organ dysfunction secondary to sepsis. Front Vet Sci. 2023;10:1127099.
- [55] Ota H, Sakurai M, Gupta R, et al. ADAR1 forms a complex with Dicer to promote microRNA processing and RNA-induced gene silencing. Cell. 2013;153:575–89.
- [56] Liu S, Xie J, Zhao B, et al. ADAR1 prevents small intestinal injury from inflammation in a murine model of sepsis [published correction appears in Cytokine. 2023 Jul;167:156225]. Cytokine. 2018; 104:30–7.
- [57] Zhao X, Xie J, Duan C, et al. ADAR1 protects pulmonary macrophages from sepsis-induced pyroptosis and lung injury through miR-21/A20 signaling. Int J Biol Sci. 2024;20:464–85.
- [58] Shangxun Z, Junjie L, Wei Z, et al. ADAR1 alleviates inflammation in a murine sepsis model via the ADAR1-miR-30a-SOCS3 Axis. Mediators Inflamm. 2020;2020:9607535.
- [59] Chen Y, Peng H, Zhou S, Zhuang Y. ADAR1 is targeted by miR-143 to regulate IL-1β-induced endothelial activation through the NFκB pathway. Int J Biochem Cell Biol. 2017;89:25–33.
- [60] Wang Y, Zhang Y, Shou S, Jin H. The role of IL-17 in acute kidney injury. Int Immunopharmacol. 2023;119:110307.
- [61] Ge Y, Huang M, Yao YM. Biology of Interleukin-17 and its pathophysiological significance in sepsis. Front Immunol. 2020;11:1558.
- [62] Han Y, Li X, Gao S, et al. Interleukin 17 is an important pathogenicity gene in pediatric sepsis. J Cell Biochem. 2019;120:3664–71.
- [63] Ahmed Ali M, Mikhael ES, Abdelkader A, et al. Interleukin-17 as a predictor of sepsis in polytrauma patients: a prospective cohort study. Eur J Trauma Emerg Surg. 2018;44:621–6.
- [64] Zhao H, Li Y, Sun G, Cheng M, Ding X, Wang K. Single-cell transcriptional gene signature analysis identifies IL-17 signaling pathway as the key pathway in sepsis. Immunobiology. 2023;228:152763.
- [65] Jin H, Wei W, Zhao Y, et al. The roles of interleukin-17A in risk stratification and prognosis of patients with sepsis-associated acute kidney injury. Kidney Res Clin Pract. 2023;42:742–50.
- [66] Mikacenic C, Hansen EE, Radella F, Gharib SA, Stapleton RD, Wurfel MM. Interleukin-17A is associated with alveolar inflammation and poor outcomes in acute respiratory distress syndrome. Crit Care Med. 2016;44:496–502.
- [67] Liang Y, Guan C, Meng H, Xie W, Meng X, Qu Y. Effects of interleukin-17A on liver and kidney injury and prognosis in septic mice. Zhonghua Wei Zhong Bing Ji Jiu Yi Xue. 2023;35:592–7.
- [68] Liang Y, Luo H, Zhang H, Dong Y, Bao Y. Oncogene Delta/Notch-Like EGF-related receptor promotes cell proliferation, invasion, and migration in hepatocellular carcinoma and predicts a poor prognosis. Cancer Biother Radiopharm. 2018;33:380–6.
- [69] Ricaño-Ponce I, Riza AL, DE Nooijer AH, et al. Characterization of sepsis inflammatory endotypes using circulatory proteins in patients with severe infection: a prospective cohort study. BMC Infect Dis. 2022;22:778.
- [70] Ruiz-Otero N, Kuruvilla R. Role of Delta/Notch-like EGF-related receptor in blood glucose homeostasis. Front Endocrinol (Lausanne). 2023;14:1161085.
- [71] El-Hodhod MA, Hamdy AM, Abbas AA, Moftah SG, Ramadan AA. Fibroblast growth factor 23 contributes to diminished bone mineral density in childhood inflammatory bowel disease. BMC Gastroenterol. 2012;12:44.
- [72] Bayer J, Vaghela R, Drechsler S, et al. The bone is the major source of high circulating intact fibroblast growth factor-23 in acute murine polymicrobial sepsis induced by cecum ligation puncture. PLoS One. 2021;16:e0251317.
- [73] Kendrick J, Cheung AK, Kaufman JS, et al; HOST Investigators. FGF-23 associates with death, cardiovascular events," and initiation of chronic dialysis. J Am Soc Nephrol. 2011;22:1913–22.
- [74] Fajol A. Fetuin-A, fibroblast growth factor 23 and inflammation in critically ill patients with sepsis. Metabol Open. 2021;12:100129.
- [75] Schnedl C, Fahrleitner-Pammer A, Pietschmann P, et al. FGF23 in Acute and Chronic Illness. Dis Markers. 2015;2015;358086.
- [76] Ito N, Wijenayaka AR, Prideaux M, et al. Regulation of FGF23 expression in IDG-SW3 osteocytes and human bone by proinflammatory stimuli. Mol Cell Endocrinol. 2015;399:208–18.

- [77] Yang Z, Wang J, Ma J, et al. Fibroblast growth factor 23 during septic shock and myocardial injury in ICU patients. Heliyon. 2024;10:e27939.
- [78] Li CH, Tang X, Wasnik S, et al. Mechanistic study of the cause of decreased blood 1, 25-dihydroxyvitamin D in sepsis. BMC Infect Dis. 2019;19:1020.
- [79] Pan W, Yu C, Hsuchou H, Zhang Y, Kastin AJ. Neuroinflammation relilif entry into brain: role of TNF. Am J Physiol Cell Physiol. 2008;294:C1436–42.
- [80] Su VY, Chiou SH, Chen WC, et al. Induced pluripotent stem cell-derived conditioned medium promotes endogenous leukemia inhibitory factor to attenuate endotoxin-induced acute lung injury. Int J Mol Sci. 2021;22:5554.
- [81] Waring P, Wycherley K, Cary D, Nicola N, Metcalf D. Leukemia inhibitory factor levels are elevated in septic shock and various inflammatory body fluids. J Clin Invest. 1992;90:2031–7.
- [82] Dai Q, Morita Y, Huang Y, et al. Modulation of Human Neutrophil Peptides on P. aeruginosa killing, epithelial cell inflammation and mesenchymal stromal cell secretome profiles. J Inflamm Res. 2019:12:335–43.
- [83] Waring PM, Waring LJ, Metcalf D. Circulating leukemia inhibitory factor levels correlate with disease severity in meningococcemia. J Infect Dis. 1994;170:1224–8.
- [84] Waring PM, Waring LJ, Billington T, Metcalf D. Leukemia infection factor protection against experimental lethal Escherichia coli septic shock in mice. Proc Natl Acad Sci USA. 1995;92:1337–41.
- [85] Weber MA, Schnyder-Candrian S, Schnyder B, et al. Endogenous leukemia inhibitory factor attenuates endotoxin response. Lab Invest. 2005;85:276–84.
- [86] Peng Y, Wang Q, Jin F, Tao T, Qin Q. Assessment of urine CCL2 as a potential diagnostic biomarker for acute kidney injury and septic acute kidney injury in intensive care unit patients. Ren Fail. 2024;46:2313171.
- [87] Song D, Zheng X. Serum monocyte chemotactic protein 1 and soluble mannose receptor aid predictive diagnosis of pediatric sepsis. Am J Transl Res. 2024;16:964–72.
- [88] Duan Y, Liu M, Wang J, Wei B. Association between Plasma Levels of Monocyte Chemoattractant Protein-1 (MCP-1) and 28-Day Mortality in elderly patients with sepsis: a retrospective single-center study. Med Sci Monit. 2024;30:e942079.
- [89] Li X. The association between MCP-1, Li x. VEGF polymorphisms and their serum levels in patients with diabetic foot ulcer. Medicine (Baltimore). 2018;97:e10959.
- [90] He J, Chen Y, Lin Y, et al. Association study of MCP-1 promoter polymorphisms with the susceptibility and progression of sepsis. PLoS One. 2017;12:e0176781.
- [91] Sierra-Filardi E, Nieto C, Domínguez-Soto A, et al. CCL2 shapes macrophage polarization by GM-CSF and M-CSF: identification of CCL2/CCR2-dependent gene expression profile. J Immunol. 2014;192:3858–67.
- [92] Purohit S, Sharma A, Hopkins D, et al. Large-scale discovery and validation studies demonstrate significant reductions in circulating levels of IL8, IL-1Ra, MCP-1, and MIP-1β in Patients With Type 1 Diabetes. J Clin Endocrinol Metab. 2015;100:E1179–87.
- [93] Ramnath RD, Ng SW, Guglielmotti A, Bhatia M. Role of MCP-1 in endotoxemia and sepsis. Int Immunopharmacol. 2008;8:810–8.
- [94] Schaalan M, Mohamed W. Predictive ability of circulating osteoprotegerin as a novel biomarker for early detection of acute kidney injury induced by sepsis. Eur Cytokine Netw. 2017;28:52–62.
- [95] Kemperman H, Schrijver IT, Roest M, Kesecioglu J, van Solinge WW, de Lange DW. Osteoprotegerin is higher in sepsis than in noninfectious SIRS and predicts 30-day mortality of SIRS patients in the intensive care. J Appl Lab Med. 2019;3:559–68.
- [96] Mota RF, Cavalcanti de Araujo PH, Cezine MER, et al. RANKL Impairs the TLR4 pathway by increasing TRAF6 and RANK interaction in macrophages. Biomed Res Int. 2022;2022:7740079.
- [97] Seshasayee D, Wang H, Lee WP, et al. A novel in vivo role for osteoprotegerin ligand in activation of monocyte effector function and inflammatory response. J Biol Chem. 2004;279:30202–9.
- [98] Wang Z, Zhang L, Lu B, Sun H, Zhong S. Causal relationships between circulating inflammatory cytokines and diabetic neuropathy: a Mendelian Randomization study. Cytokine. 2024;177:156548.
- [99] Andreu-Bernabeu A, González-Peñas J, Arango C, Díaz-Caneja CM. Socioeconomic status and severe mental disorders: a bidirectional multivariable Mendelian randomisation study. BMJ Ment Health. 2023;26:e300821.
- [100] Wang J, Yang M, Tian Y, et al. Causal associations between common musculoskeletal disorders and dementia: a Mendelian randomization study. Front Aging Neurosci. 2023;15:1253791.