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Association between glucose-to-lymphocyte ratio and in-hospital mortality in intensive care patients with sepsis: A retrospective observational study based on Medical Information Mart for Intensive Care IV

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Background: This study aimed to evaluate the association between the glucose-to-lymphocyte ratio (GLR) and in-hospital mortality in intensive care unit (ICUs) patients with sepsis.

Methods: This is a retrospective cohort study. Patients with sepsis from the Medical Information Mart for Intensive Care-IV (MIMIC-IV) database had their baseline data and in-hospital prognosis retrieved. Multivariable Cox regression analyses were applied to calculate adjusted hazard ratios (HR) with 95% confidence intervals (CI). Survival curves were plotted, and subgroup analyses were stratified by relevant covariates. To address the non-linearity relationship, curve fitting and a threshold effect analysis were performed.

Results: Of the 23,901 patients, 10,118 patients with sepsis were included. The overall in-hospital mortality rate was 17.1% (1,726/10,118). Adjusted for confounding factors in the multivariable Cox regression analysis models, when GLR was used as a categorical variable, patients in the highest GLR quartile had increased in-hospital mortality compared to patients in the lowest GLR quartile (HR = 1.26, 95% CI: 1.15–1.38). When GLR was used as a continuous variable, each unit increase in GLR was associated with a 2% increase in the prevalence of in-hospital mortality (adjusted HR = 1.02, 95% CI: 1.01–1.03, $p = 0.001$). Stratified analyses indicated that the correlation between the GLR and in-hospital mortality was stable. The non-linear relationship between GLR and in-hospital mortality was explored in a dose-dependent manner. In-hospital mortality increased by 67% (aHR = 1.67, 95%

CI: 1.45–1.92) for every unit GLR increase. When GLR was beyond 1.68, in-hospital mortality did not significantly change (aHR: 1.04, 95% CI: 0.92–1.18).

Conclusion: There is a non-linear relationship between GLR and in-hospital mortality in intensive care patients with sepsis. A higher GLR in ICU patients is associated with in-hospital mortality in the United States. However, further research is needed to confirm the findings.

KEYWORDS

glucose-to-lymphocyte ratio, sepsis, MIMIC-IV, in-hospital mortality, non-linearity, intensive care unit

Background

Sepsis is a serious public health concern worldwide. Sepsis is a life-threatening organ dysfunction caused by dysregulated host systemic inflammation and immune response to infection (1, 2). Despite advances in the recognition and management of clinical sepsis (3), morbidity and mortality remain high (4, 5), with sepsis-related deaths accounting for 19.7% of global deaths (6). To date, the exact mechanism of sepsis remains unclear but is widely hypothesized.

Many clinical studies consider sepsis to be a host-mediated systemic inflammatory response to infection, and evidence of dysregulated immune cell activation and host response has been observed in patients with severe sepsis (7, 8). In addition, some systemic inflammatory biomarkers have been reported to be associated with sepsis and poor prognosis, including neutrophil-lymphocyte ratio (NLR) (9–11), platelet-lymphocyte ratio (PLR) (12), lymphocyte-monocyte ratio (LMR) (13), and red cell distribution width (RDW) (14–16). The loss and dysfunction of immune cells are considered the main factors for secondary infections and poor outcomes in patients with sepsis. Therefore, alterations in immune cell number and function may be related to mortality in patients with sepsis (17). Lymphocytes are one of the primary effector cells involved in the systemic inflammatory response of sepsis. Extensive lymphocyte apoptosis is a key contributor to the development of the immunosuppressive phase of sepsis (18). Their profound role in immunosurveillance, which may protect the host from sepsis development and impaired immune system, has been reported to be associated with poor prognosis in patients with sepsis (9). Consequently, lymphocyte count indicating the state of the immune system appears to predict the outcomes of patients with sepsis (18).

In addition, numerous studies have demonstrated an association between failure to control hyperglycemia and adverse outcomes in patients in the intensive care unit (ICU), including death, nosocomial infection, wound complications, prolonged ICU stay, and an increased incidence of critical illness

neuropathy (19). Acute hyperglycemia is an independent risk factor for in-hospital mortality in critically ill patients with sepsis (20).

The imbalance between these two indicators is reflected in the changes in the glucose-to-lymphocyte ratio (GLR). In this case, increased GLR indicates an imbalance in glucose regulation and immune responses (21). This imbalance leads to organ failure, metabolic problems, immune deficiencies, and oxygen supply and demand mismatch, all leading to death (22). There is growing evidence that elevated glucose levels and decreased lymphocyte counts are strongly associated with sepsis severity (11, 23). GLR may reflect the synergistic effect of hyperglycemia and immune dysfunction in critically ill patients (24). In addition, an increased GLR has been associated with poor prognosis in a range of disease cases, such as gallbladder cancer (25), pancreatic cancer (26), acute pancreatitis (27), and acute kidney injury (24). However, previous studies have not evaluated the prognostic relationship between biomarkers combined with glucose and lymphocyte counts in patients with sepsis. This study sought to assess the relationship between the GLR and hospital outcomes in patients with sepsis, an index that includes both glucose levels and systemic inflammation and may provide a new basis and reference for the clinical management of sepsis.

Materials and methods

Data source

We enrolled patients with sepsis from the MIMIC-IV (Medical Information Mart for Intensive Care IV, version 1.0) (28) database of the Massachusetts Institute of Technology (MIT). More than 70,000 adult patients were admitted to the intensive care unit (ICU) of Beth Israel Deaconess Medical Center in Boston between 2008 and 2019. Informed consent was waived because the data were obtained from publicly available sources. One author, Shaoyan Cai, obtained full

access to the database and completed the data extraction (certification number 46658933). Strengthening the Reporting of Observational Studies in Epidemiology guidelines (29) was used to conduct this study.

Participants

Patients aged >18 years who fulfilled the Sepsis-3 criteria (1) were eligible for our study. Sepsis was defined as an increase of ≥ 2 points in the sequential organ failure assessment (SOFA) score, plus documented or suspected infection (1, 30).

Septic shock was defined as (ICD) code 78552 (9th revision) and ICD code R6521 (10th revision). The diagnosis of diabetes was based on ICD-9. If patients were admitted to the ICU more than once, we only adopted the date of their first ICU admission (31).

Variates

Variables considered confounders of sepsis outcomes based on existing literature and clinical judgment were included (23, 32), except glucose and lymphocytes count because of their collinearity with GLR.

Demographic and admission information: age, sex, ethnicity, insurance, weight, Charlson comorbidity index (CCI), and severity at admission, as measured by the Acute Physiology Score (APS) III score and SOFA score.

Vital signs: Heart rate, mean arterial pressure (MAP), and SPO₂ at ICU admission.

Interventions: Mechanical ventilation, renal replacement treatment (RRT), and vasopressor agent use during the first 24 h of ICU admission.

Laboratory results: Glucose, lymphocyte count, hemoglobin, white blood cell (WBC) count, platelet count, neutrophil count, lactate, and pH.

GLR was calculated using the serum blood glucose (mmol/L)/lymphocyte count ($\times 10^9/L$).

If the above data were tested multiple times within 24 h, we chose the first set of parameters.

Outcome

The outcome was in-hospital mortality, which is defined as survival status at hospital discharge. Patients without any outcome information were excluded from the final cohort.

Statistical analysis

Descriptive analysis was performed for categorical variables to assess the significance of differences between groups stratified by GLR quartiles (<0.43; 0.43–0.78; 0.78–1.56; ≥ 1.56) using the Kruskal–Wallis test or one-way analysis of variance. Baseline characteristic data are presented as proportions (%) and were compared using chi-square tests for categorical variables. Normally distributed continuous data are presented as mean \pm standard deviation (SD) and compared using Student's *t*-test between groups, while skewed distribution data are presented as the median and interquartile range (IQR) and compared using the Wilcoxon rank-sum test.

A multivariate Cox proportional hazard model was used to assess the independent association between the GLR and in-hospital mortality. We constructed three models: Model 1, adjusted only for age and sex. Model 2 was additionally adjusted for ethnicity, weight, MAP, hazard ratio (HR), SPO₂,

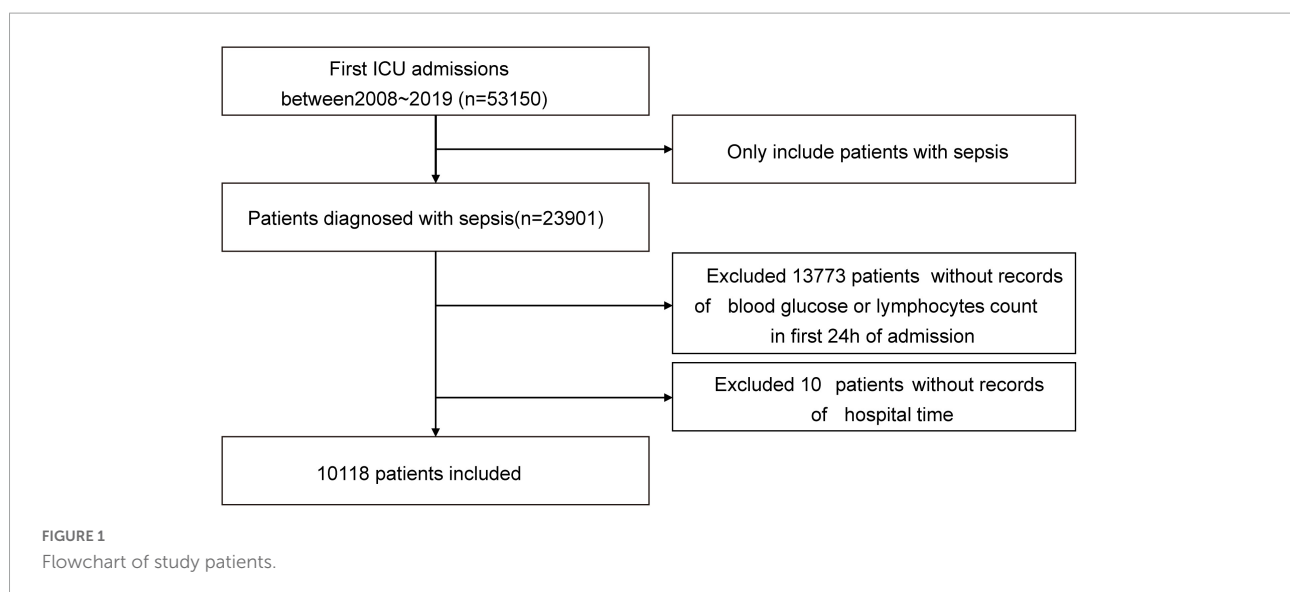


TABLE 1 Baseline characteristics of participants and outcome parameters.

Variables	All patients	Q1	Q2	Q3	Q4	P-value
		(GLR < 0.43)	(0.43 ≤ GLR < 0.78)	(0.78 ≤ GLR < 1.56)	(GLR ≥ 1.56)	
N	10118	2447	2576	2546	2549	
Age(year)	65.8 ± 16.3	64.7 ± 16.0	65.1 ± 16.3	66.1 ± 17.0	67.3 ± 15.8	< 0.001
Female, n (%)	4262 (42.1)	1004 (41)	991 (38.5)	1139 (44.7)	1128 (44.3)	< 0.001
Ethnicity, white, n (%)	6643 (65.7)	1630 (66.6)	1724 (66.9)	1640 (64.4)	1649 (64.7)	0.188
Insurance, Medicaid, n (%)	5602 (55.4)	1429 (58.4)	1476 (57.3)	1378 (54.1)	1319 (51.7)	< 0.001
weight(kg)	83.7 ± 23.7	83.0 ± 21.7	84.2 ± 23.4	83.9 ± 24.3	83.8 ± 25.2	0.295
Vital Signs						
Heart rate (bpm)	87.8 ± 16.3	85.1 ± 15.3	86.7 ± 15.3	89.1 ± 16.8	90.2 ± 17.3	< 0.001
MAP (mmHg)	75.8 ± 9.9	75.6 ± 9.2	75.8 ± 9.8	76.1 ± 10.1	75.9 ± 10.5	0.246
SPO2 (%)	96.8 ± 2.6	97.2 ± 2.4	96.9 ± 2.6	96.7 ± 2.3	96.4 ± 2.9	< 0.001
Laboratory results						
Hemoglobin (g/L)	10.4 ± 1.9	10.1 ± 1.7	10.4 ± 1.8	10.5 ± 2.0	10.6 ± 2.1	< 0.001
Platelet (× 1012)	171.0 (121.5, 239.0)	145.5 (109.5, 197.5)	167.0 (124.5, 231.1)	186.5 (128.5, 256.6)	195.0 (132.5, 272.5)	< 0.001
WBC(× 109/L)	12.5 (8.9, 16.9)	9.7 (6.6, 13.1)	11.9 (8.7, 15.3)	13.3 (9.7, 17.7)	15.7 (11.7, 21.5)	< 0.001
Neutrophil (× 109/L)	9.9 (6.6, 14.1)	6.7 (4.3, 9.4)	9.6 (6.8, 12.5)	10.8 (7.8, 14.7)	13.8 (9.8, 18.6)	< 0.001
Lactate (mmol/L)	2.6 ± 2.2	2.3 ± 1.8	2.4 ± 1.9	2.7 ± 2.4	3.1 ± 2.4	< 0.001
pH	7.4 ± 0.1	7.4 ± 0.1	7.4 ± 0.1	7.3 ± 0.1	7.3 ± 0.1	< 0.001
Glucose (mmol/L)	7.2 (6.1, 9.1)	6.2 (5.4, 7.1)	6.8 (5.9, 7.9)	7.7 (6.5, 9.6)	9.3 (7.3, 12.2)	< 0.001
Lymphocytes(× 109/L)	9.7 (5.3, 16.0)	21.4 (17.4, 27.7)	12.0 (10.0, 14.5)	7.3 (5.9, 9.2)	3.4 (2.1, 5.0)	< 0.001
GLR	0.8 (0.4, 1.6)	0.3 (0.2, 0.4)	0.6 (0.5, 0.7)	1.1 (0.9, 1.3)	2.6 (2.0, 3.9)	< 0.001
Score system, points						
CCI	5.8 ± 2.9	5.3 ± 2.8	5.5 ± 2.9	5.9 ± 3.0	6.3 ± 3.0	< 0.001
APS III score	58.0 ± 27.6	49.2 ± 25.5	53.3 ± 25.8	61.0 ± 27.2	68.3 ± 27.9	< 0.001
SOFA score	3.9 ± 2.2	3.8 ± 2.0	3.7 ± 2.0	3.9 ± 2.3	4.1 ± 2.4	< 0.001
Interventions						
Ventilator use, n (%)	5202 (51.4)	1259 (51.5)	1363 (52.9)	1284 (50.4)	1296 (50.8)	0.304
Diabetes, n (%)	3058 (30.2)	617 (25.2)	675 (26.2)	793 (31.1)	973 (38.2)	< 0.001
RRT, n (%)	565 (5.6)	85 (3.5)	95 (3.7)	166 (6.5)	219 (8.6)	< 0.002
Vasopressin use, n (%)	996 (9.8)	148 (6)	204 (7.9)	263 (10.3)	381 (14.9)	< 0.001
death, n (%)	1726 (17.1)	227 (9.3)	326 (12.7)	479 (18.8)	694 (27.2)	< 0.001

Data are presented as the mean ± standard deviation (SD), median (IQR) for skewed variables, and numbers (proportions) for categorical variables.

bpm, beats per minute; MAP, mean arterial pressure; WBC, white blood count; GLR, glucose-to-lymphocyte ratio; CCI, Charlson comorbidity index; APS III, Acute Physiology Score III; SOFA, Sequential Organ Failure Assessment; RRT, renal replacement treatment.

hemoglobin, platelet (PLT), WBC, lactate, and pH. Model 3 was additionally adjusted for SOFA score, APS III score, ventilator use, diabetes, CCI, vasopressin usage, and neutrophil count. In all models, linear trends were tested using GLR quartiles as categorical variables by assigning the median values of the quartiles to the variable.

A Cox proportional hazards regression model was used to assess the non-linear relationship between GLR and the outcome of sepsis. Based on the curve fitting (restricted cubic spline), we conducted a two-piecewise linear regression model to identify threshold effects, if a non-linear correlation was observed. Threshold levels of GLR were determined using a recursive method, and a maximum likelihood model was yielded.

A sensitivity analysis was performed to ensure the robustness of the data analysis. GLR was transformed into a categorical variable and a *p*-value for the trend was calculated. The purpose of this test was to validate the results of treating the GLR as a continuous variable and to determine the possibility of non-linearity.

Hospital survival was assessed using Kaplan–Meier survival curves according to GLR quartiles and evaluated using the log-rank test.

Stratified and interaction analyses were applied based on sex (male or female), age (<65 or ≥65 years), diabetes (yes or no), ventilator use (yes or no), and RRT use (yes or no). Subgroup analyses were adjusted for relevant covariates (age, sex, ethnicity, weight, MAP, HR, SPO2, hemoglobin, PLT, WBC,

TABLE 2 Multivariable Cox regression to assess the association of GLR with in-hospital mortality.

Variable	Unadjusted		Model 1		Model 2		Model 3	
	HR_95CI%	P-value	HR_95CI%	P-value	HR_95CI%	P-value	HR_95CI%	P-value
GLR	1.11 (1.1~1.12)	< 0.001	1.11 (1.1~1.12)	< 0.001	1.06 (1.05~1.07)	< 0.001	1.02 (1.01~1.03)	0.004
GLR4								
Q1(GLR < 0.43)	1(Ref)		1(Ref)		1(Ref)		1(Ref)	
Q2(0.43 ≤ GLR < 0.78)	1.23 (1.12~1.36)	< 0.001	1.23 (1.12~1.36)	< 0.001	1.18 (1.07~1.3)	0.001	1.2 (1.08~1.32)	0.001
Q3(0.78 ≤ GLR < 1.56)	1.6 (1.46~1.76)	< 0.001	1.57 (1.43~1.72)	< 0.001	1.34 (1.22~1.47)	< 0.001	1.23 (1.12~1.35)	< 0.001
Q4(GLR ≥ 1.56)	2.33 (2.14~2.55)	< 0.001	2.25 (2.06~2.46)	< 0.001	1.6(1.46~1.75)	< 0.001	1.3(1.185~1.43)	< 0.001
P for trend.test		< 0.001		<0.001		< 0.001		<0.001

GLR, glucose-to-lymphocyte ratio.

Model 1 = Adjust for (Age + sex).

Model 2 = Model 1 + (ethnicity + weight + MAP + HR + SPO2 + hemoglobin + PLT + WBC + lactate + pH).

Model 3 = Model 2 + (SOFA score + APS III + ventilator use + diabetes + CCI + vasopressin use + neutrophil).

lactate, pH, SOFA score, APS III, ventilator use, diabetes, CCI, and vasopressin use).

The percentages of covariates with missing data were less than 30% for all analyses. The missing values of the covariates were imputed *via* multiple imputations. We created and analyzed three datasets together. To assess the robustness of the findings, we applied sensitivity analysis of patients after excluding missing data from the study ([Supplementary Table 1](#)).

Data analyses were performed using packages R 4.1.2 (The R Foundation)¹ software and Free Statistics software versions 1.5. *P*-values < 0.05 were considered significant.

Results

Population

In total, 23,901 patients were identified according to the sepsis-3 criterion. Of these, 13,343 patients without GLR values and in-hospital time were excluded, and 10,118 with sepsis were included in the final cohort ([Figure 1](#) shows a flow chart).

Baseline characteristics

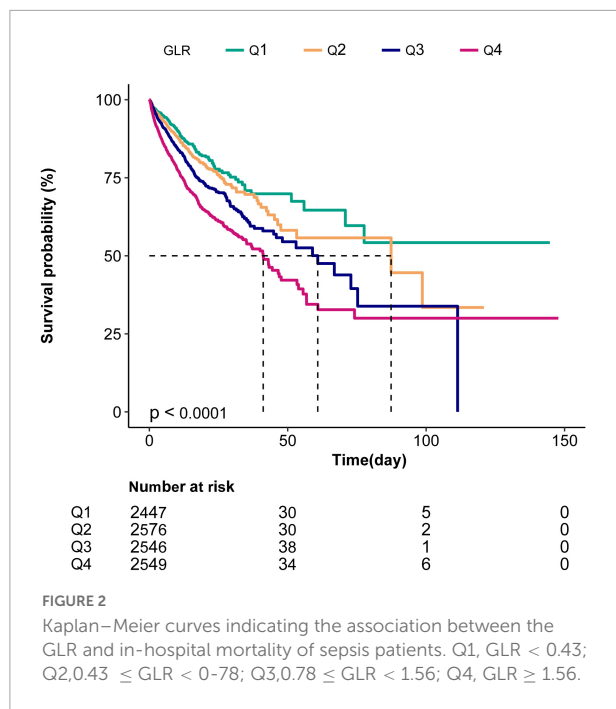
The basic demographic characteristics of all selected patients are summarized in [Table 1](#), stratified by GLR quartile. In general, the age of all participants was 65.8 ± 16.3 years old, and approximately 42.1% were female. The in-hospital mortality rate was 20.1% (480/2,383). Participants in the highest group of GLR (Q4) had higher values for age, heart rate, hemoglobin, platelet, WBC, neutrophil, lactate, glucose, APS III score, CCI,

SOFA score, and were more likely to have diabetes, RRT, vasopressin use, and death than those in the other groups. The opposite patterns were observed for SPO2, pH, lymphocytes, and insurance for medical aid.

Multivariable Cox regression analysis

In this study, we constructed three models to analyze the independent effects of the GLR on in-hospital mortality (multivariate Cox regression model; [Table 2](#)). The effect sizes (HRs) and 95% confidence intervals were listed. We observed that the HRs were robust between the unadjusted and adjusted models in all three models (*p* < 0.05). In the unadjusted model, the effect size of GLR for in-hospital mortality means that a difference of one unit of GLR is associated with an in-hospital mortality difference increased by 11% (HR = 1.11, 95% CI: 1.10–1.12). In the minimum-adjusted model (Model 1), with an increase in the GLR of one unit, the in-hospital mortality difference increased by 11% (HR = 1.11, 95% CI 1.1–1.12). In the fully adjusted model (Model 3) (adjusted covariates of age, sex, ethnicity, weight, MAP, HR, SPO2, hemoglobin, PLT, WBC, lactate, pH, SOFA score, APS III, ventilator use, diabetes, CCI, vasopressin usage, and neutrophil count) for each additional unit of GLR, in-hospital mortality difference increased by 2% (HR = 1.02, 95% CI 1.01–1.03). For further sensitivity analysis, the continuous variable GLR was converted into a categorical variable (quartile of GLR), of which the first category GLR (Q1) was used as a baseline reference. Patients in the highest GLR quartile had increased in-hospital mortality compared to patients in the lowest GLR quartile (HR = 1.26, 95% CI 1.15–1.38). The *P* for the trend in the fully adjusted model for GLR as a categorical variable was the result when GLR was a continuous variable. Moreover, the trend for effect size in the different GLR groups was equidistant.

¹ <http://www.R-project.org>



Kaplan–Meier curves

The Kaplan–Meier curve demonstrated that the in-hospital survival of the highest GLR quantile (Q4) patients was the lowest of all groups, which declined with declining baseline GLR (log-rank test: $p < 0.0001$; **Figure 2**).

Subgroup analysis

Subgroup analyses indicated no significant interaction in the subgroup analysis (all p -values for interaction were > 0.05 ; **Figure 3**).

The analyses of the non-linear relationship

Restricted cubic spline (**Figure 4**) showed that the relationship between GLR and in-hospital mortality was non-linear after adjusting for related confounding factors. Because the P for the log-likelihood ratio test was < 0.05 , we chose the two-piecewise Cox proportional hazard model for fitting the association between GLR and in-hospital mortality. By the two-piecewise Cox proportional hazard model and recursive algorithm, we calculated the inflection point was 1.68. It was shown that stronger positive association between GLR and in-hospital mortality within the inflection point of 1.68. In-hospital mortality increased by 67% (aHR = 1.67, 95% CI: 1.45–1.92)

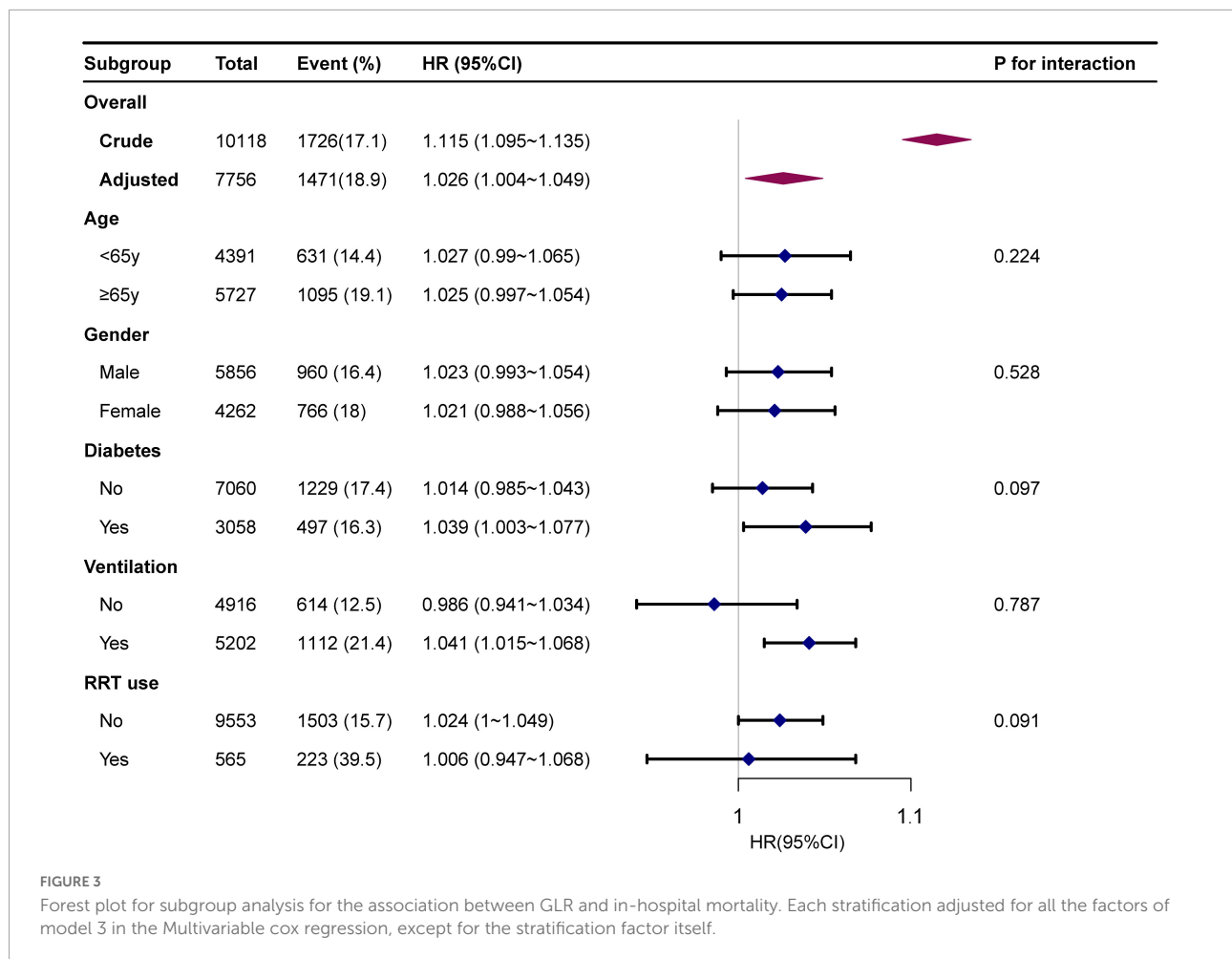
for every unit GLR increase. When GLR was beyond 1.68, in-hospital mortality did not significantly change (aHR: 1.04, 95% CI: 0.92–1.18; **Table 3**).

Discussion

This study evaluated the association of GLR, a combination of blood glucose levels and lymphocyte count, with in-hospital mortality after adjusting for the variables in a population-based analysis. Our findings indicate that an elevated GLR is associated with higher in-hospital mortality. Furthermore, as a continuous or categorical variable, GLR was positively associated with in-hospital mortality in intensive care patients with sepsis in the United States. Besides, the inflection point of GLR was 1.68, and we found the trend of HR on the two sides of the inflection point was not consistent. The result suggested a turning point effect on the independent association between GLR and in-hospital mortality.

Sepsis is characterized by systemic and organ-specific metabolic changes. Altered oxygen consumption, elevated circulating substrate levels, impaired glucose and lipid oxidation, and mitochondrial dysfunction are associated with organ dysfunction and adverse outcomes in animal models and patients (33). Sepsis can lead to a loss of glucose homeostasis, and the resulting hyperglycemia adversely affects immune function and metabolism, leading to poor outcomes (34, 35). The mechanisms that lead to glucose dysregulation are complex. Elevated blood glucose levels tend to reduce membrane fluidity, which impedes polymorphonuclear leukocyte (PMN) function, leading to reduced phagocytosis, intracellular killing, suboptimal migration, and chemotaxis (20, 36, 37). In addition, the neuroendocrine stress response can increase adrenal cortex secretion by 10 times, including excessive glycogenolysis, gluconeogenesis, and insulin resistance (38).

A low lymphocyte count may also be associated with a shortened survival time in sepsis (22). Clinical studies have shown that lymphocyte counts in the blood decrease during sepsis and remain low for up to 28 days (39, 40). Although the absolute lymphocyte counts of sepsis survivors and non-survivors were severely decreased at the onset of sepsis, lymphocyte counts recovered in survivors, while absolute lymphocyte counts remained persistently low in non-survivors (41). Various anti-inflammatory cytokines released into the bloodstream can induce immunosuppression and lead to massive lymphocyte apoptosis (42). Lymphopenia is a common marker of sepsis-induced immunosuppression, as it prevents microbial clearance and induces severe infections, which are the leading causes of sepsis-related death (39). Apoptosis-induced lymphocytopenia often occurs in sepsis and severe injuries, including major surgery, burns, and trauma. As active lymphocytes migrate to inflammatory areas, lymphocyte apoptosis increases (43). This process begins immediately after



the potential damage occurs. The severity and duration of lymphocytopenia are associated with poor clinical outcomes. The severity and duration of lymphopenia are associated with poor clinical outcomes. Extensive apoptosis of lymphocytes occurs in lymphoid (lymph nodes, thymus, and spleen) and other organs (44) leading to impaired immune cell activity, which is a key contributor to the development of the immunosuppressive phase of sepsis and plays a direct or indirect role in injury-induced immune paralysis (45).

The exact mechanism underlying the association between elevated GLR levels and poor prognosis in patients with sepsis is unclear. Recently, several researchers have been interested in biomarkers that combine blood glucose levels and inflammatory indicator lymphocytes to predict the prognosis of certain diseases. Navarro suggested that the preoperative GLR was an independent predictor of overall survival (OS) and disease-free survival (DFS) after surgery for T2 gallbladder cancer. This is the first report of the predictive value of GLR (25). As an easily available biomarker, Chen et al. reported that GLR was an independent predictor of in-hospital mortality in critically ill patients with acute pancreatitis. They combined

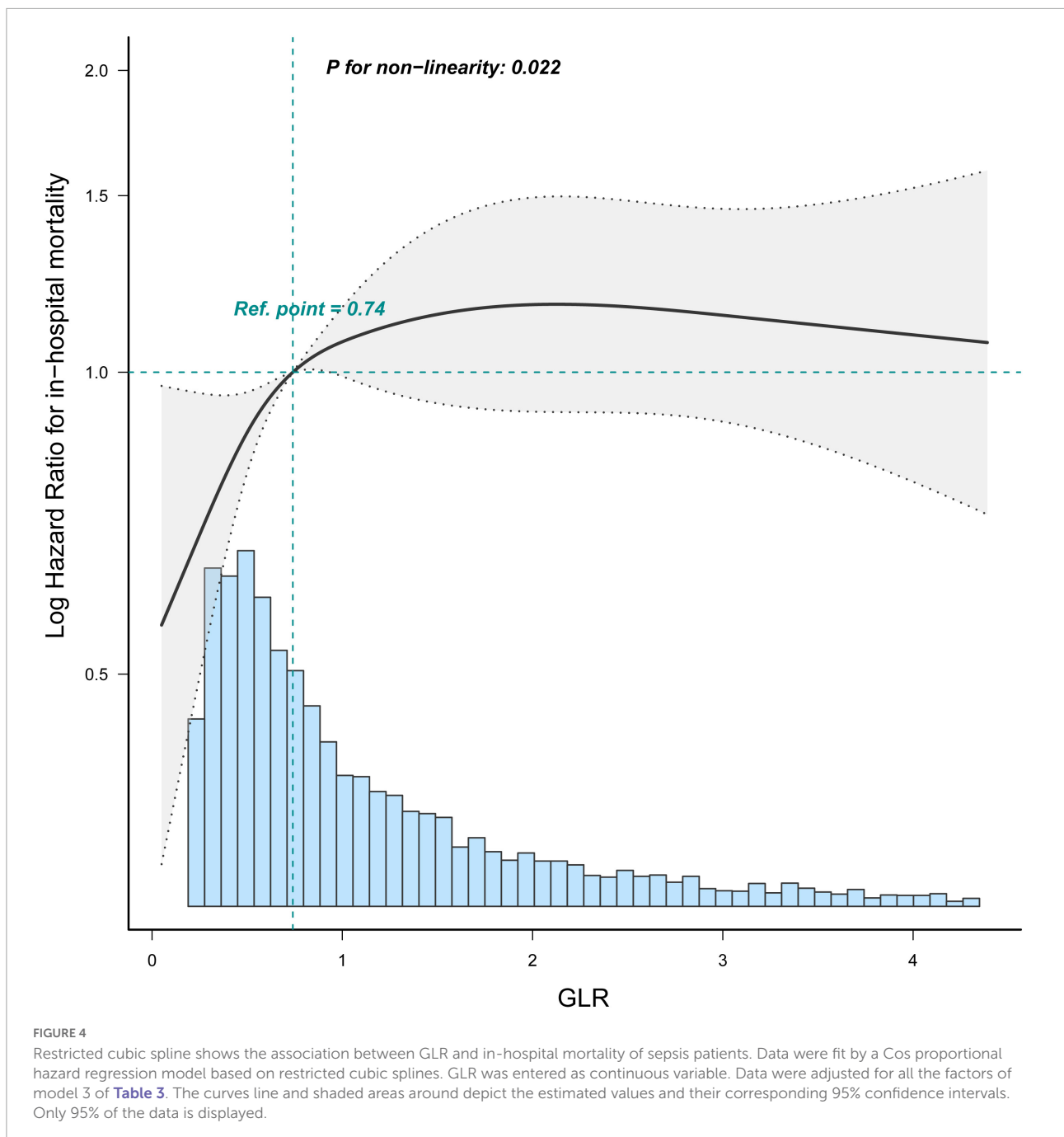
TABLE 3 Threshold effect analysis of the relationship between GLR and in-hospital mortality of patients with sepsis.

Threshold of GLR	HR 95CI%	P-value
< 1.68	1.67 (1.45,1.92)	< 0.001
≥ 1.68	1.04 (0.92,1.18)	0.5223
Likelihood Ratio test	-	< 0.001

Data were adjusted for all the factors of Model 3 of Table 2.

GLR with other clinical characteristics of acute pancreatitis to construct nomograms with favorable predictive performance for in-hospital mortality (27). Two other studies showed that GLR is an independent predictor of prognosis in patients with pancreatic cancer (26, 46). Preoperative GLR was also a promising predictor of acute kidney injury after cardiac surgery in ICU patients (24). Therefore, it is worth considering that GLR may reflect a synergistic effect of immunocompromise and hyperglycemia in sepsis.

To the best of our knowledge, this is the first report of an independent association between GLR and in-hospital mortality in ICU patients with sepsis. This study could help establish



diagnostic or predictive models of in-hospital sepsis mortality in future research.

Our study has several strengths. First, the study used real-world data for a large and diverse population. Second, strict statistical adjustment was used to minimize susceptibility to potential residual confounders in this retrospective observational study. Third, we considered the target independent variables as both continuous and categorical variables. With this approach, contingency in the data analysis was reduced, and the robustness of the results was enhanced.

Fourth, the non-linear processing of the study is a major improvement compared to former studies. Finally, the effect modifier factor analysis improved data usage and yielded more robust results in different subgroups.

There are some noteworthy limitations to this study. First, in the MIMIC-IV database, we could not obtain data on procalcitonin and organ functions, and other residual confounders potentially exist, as in all retrospective analyses. Some patients with sepsis were excluded from our study because of the lack of necessary data, which may have led to bias in

the study results. Second, the influence of antibiotic use on results was not considered. We believe that this is an important subject and will be the objective of our future research. Third, our research subjects were intensive care patients with sepsis. Therefore, the universality and extrapolation of research are lacking. Moreover, GLR values changed dynamically during hospitalization. However, the GLR value used in the present study was not calculated based on the date of the onset of sepsis but on the first day of admission to the ICU or hospital. Therefore, this may have caused bias in the results. Finally, it was a retrospective study based on the MIMIC-IV database; therefore, our study was a *post hoc* analysis of the MIMIC-IV database, the level of evidence was not strong enough, and further high-quality prospective studies are needed to validate the relationship between GLR and sepsis prognosis.

Conclusion

There was a non-linear relationship between GLR and in-hospital mortality in intensive care patients with sepsis. A higher GLR in ICU patients is associated with in-hospital mortality.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found below: <https://mimic.physionet.org/>.

Ethics statement

Ethical review and approval was not required for the study on human participants in accordance with the local legislation and institutional requirements. Written informed consent for participation was not required for this study in accordance with the national legislation and the institutional requirements.

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Author contributions

SC: study design and manuscript writing. CM: modified the manuscript. LeZ and YW: data collection. JC: data interpretation. ZF: statistical analysis. LiZ and CG: project administration. All authors have approved the manuscript and agreed to the submitted version.

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Conflict of interest

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

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Supplementary material

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fmed.2022.922280/full#supplementary-material>

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