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Association between anion gap/calcium ratio and 30-day all-cause mortality in sepsis patients with diabetes mellitus

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We aimed to determine the association between anion gap-to-calcium ratio (ACR) and 30-day mortality in sepsis patients with diabetes mellitus (DM). Data for sepsis patients diagnosed with DM was extracted from Medical Information Mart for Intensive Care Database IV. After screening, 4429 eligible subjects were included in our study finally. The receiver operating characteristic (ROC) curve was used to determine the cut-off value. According to the ROC curve analysis, the ACR demonstrated a higher area under the curve (AUC) of 0.622 compared to AG (0.598). Multivariable logistic regression with inverse probability of treatment weighting (IPTW) based on propensity score were used to detect the association between ACR and 30-day mortality. Our results showed that the patients with the high level of ACR had a higher risk of death within 30 days compared with those with low level of ACR (odds ratio 1.342, 95% confidence interval 1.180–1.526, $P < 0.001$). In a word, our results suggest that ACR may be an independent prognostic indicator for death with 30 days in critically ill patients with sepsis and DM.

Keywords Sepsis, Diabetes mellitus (DM), Anion gap, Calcium, Mortality

Diabetes mellitus (DM) is a tremendous health problem worldwide. It is reported that more than 529 million people are living with diabetes worldwide today, and this number is estimated to rise to 1.31 billion by 2050¹. Because of a compromised immune system, population with DM often have weakened defenses against pathogens^{2–7}. Approximately 2.6 million people with DM develop sepsis each year. Of these, approximately 15% die as the disease progresses⁸. Therefore, it is necessary to strengthen the supervision and management of sepsis patients with DM, and timely identify the population with a higher risk of death, and ultimately reduce the overall mortality of this group.

Accumulated evidences have revealed that some laboratory indicators are associated with the risk of death in sepsis patients with DM. In 2022, Xin et al. reported that the level of platelet was negatively associated with major adverse kidney events within 30 days in sepsis patients with DM⁹. In 2023, a large-scale cohort study showed that the high level of red blood cell distribution width was also associated with mortality in sepsis patients with DM¹⁰. However, above these indicators are also associated with prognosis for other diseases, with limited specificity in predicting the prognosis of critically ill sepsis patients with DM^{11–14}. Therefore, it is necessary to further screen more reliable indicators to identify these septic patients with a high risk of death at an early stage. Several studies recent reported that the level of anion gap (AG) can be used as a prognostic indicator for sepsis patients^{15,16}, and it was associated with short-term mortality among sepsis patients with DM¹⁰. Moreover, hypocalcemia is common in critically ill patients¹⁷, and the level of blood calcium is also closely related to insulin secretion^{18,19}. Therefore, it is speculated that a composite of AG and blood calcium level may be an indicators for predicting the prognosis of sepsis patients with DM. However, no published studies have investigated the association between a composite of AG and blood calcium and death in sepsis patients with DM. The aim of this paper was to investigate the association between a composite indicator of AG and calcium and 30-day mortality in critical ill sepsis patients with DM from the intensive care unit (ICU).

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Results

Clinical characteristics of included patients

The initial search identified 65,366 ICU admissions from the MIMIC-IV database. After further screening, we identified 4429 patients who met the inclusion criteria to be included in the final analysis. The flow of selection process is shown in Fig. 1.

The baseline characteristics between survivors and non-survivors were showed in Table 1. Compared with the survivor group, patients in the non-survival group presented higher levels of AG, ACR, Glu, BUN, WBC, RDW, Cr, heart rate, respiratory rate, SAPS II scores and SOFA scores, and lower levels of body mass index and temperature. The patients in non-survival group were older, were more likely to be treated with mechanical ventilation, and had more comorbidities such as coronary artery disease, valvular disease, malignant cancer, and septic shock than those in survival group.

Associations between ACR and 30-day mortality

The significant different variables, including age, BMI, SOFA, SAPS II, ventilation, ACR, BUN, RDW, Cr, WBC, glucose, coronary artery disease, malignant cancer, and septic shock, were used in multivariate logistic regression

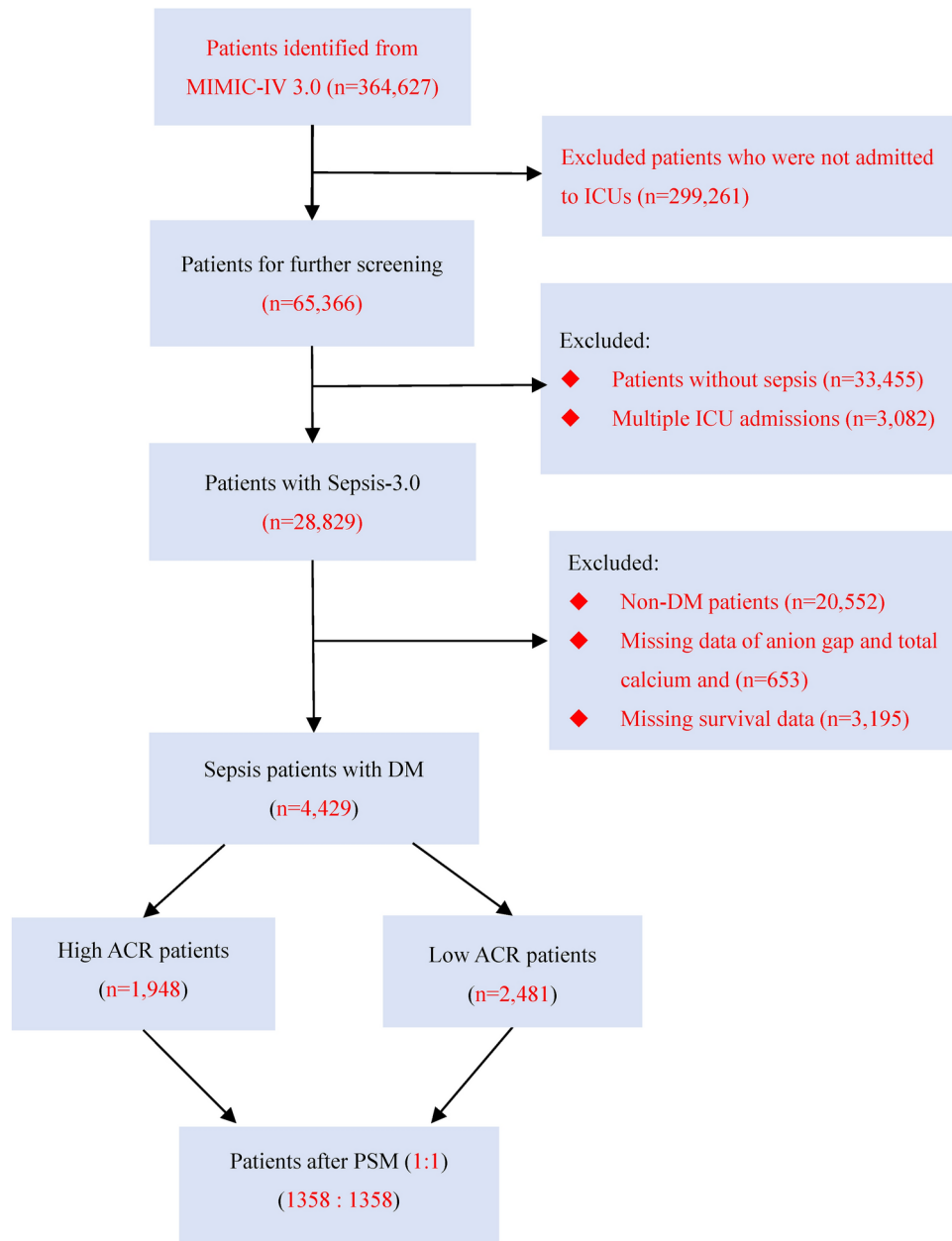


Fig. 1. Flowchart of patient selection. Abbreviations: MIMIC-IV 3.0, Medical Information Mart for Intensive Care IV, Version: 3.0; ICU, intensive care unit; DM, diabetes mellitus. ACR, anion gap-to-calcium ratio; PSM, propensity score matching.

Variables	Survivors group (2612)	Non-survivors group (1817)	P-value
Age [mean (SD)], years	70.2 ± 12.7	73.4 ± 12.3	< 0.001
Gender, male, N (%)	1535 (58.8)	1043 (57.4)	0.365
BMI [mean (SD)], kg/m ²	30.5 ± 8.0	29.3 ± 7.5	< 0.001
SOFA score [mean (SD)]	6 ± 3	8 ± 4	< 0.001
SAPS II score [mean (SD)]	43 ± 12	50 ± 14	< 0.001
Ventilation, N (%)	1176 (45.0)	1057 (23.3)	< 0.001
Laboratory tests			
Calcium [mean (SD)], mg/dL	8.4 ± 0.8	8.3 ± 0.8	0.2673
Anion gap [mean (SD)], mmol/L	15 ± 4	18 ± 5	< 0.001
ACR [mean (SD)]	1.83 ± 0.54	2.09 ± 0.69	< 0.001
Glu [mean (SD)], mg/dL	175 ± 72	185 ± 80	0.001
Lac [mean (SD)], mmol/L	2.2 ± 1.5	2.5 ± 1.5	0.051
PLT × 10 ⁹ /L [mean (SD)]	204 ± 110	199 ± 111	0.143
INR [mean (SD)]	1.8 ± 1.4	1.7 ± 1.1	0.254
BUN [mean (SD)], mg/dL	37 ± 26	44 ± 29	< 0.001
WBC × 10 ⁹ /L [mean (SD)]	13.2 ± 11.5	14.2 ± 9.6	0.003
RDW [mean (SD)], %	16.0 ± 2.4	16.5 ± 2.8	< 0.001
RBC × 10 ¹² /L [mean (SD)]	3.4 ± 0.7	3.4 ± 0.8	0.795
HCT [mean (SD)], %	30.9 ± 6.0	31.2 ± 6.2	0.211
Hb [mean (SD)], g/L	10.0 ± 2.0	10.0 ± 2.1	0.711
Cr [mean (SD)], mg/dL	2.0 ± 1.8	2.2 ± 1.8	< 0.001
Vital signs			
Heart rate [mean (SD)], times/min	90 ± 20	92 ± 21	0.004
MAP [mean (SD)], mmHg	80 ± 18	81 ± 19	0.993
Resp rate [mean (SD)], times/min	20 ± 6	21 ± 7	0.001
Temp [mean (SD)], °C	36.9 ± 0.7	36.8 ± 0.8	< 0.001
SpO ₂ [mean (SD)], %	96 ± 4	96 ± 4	0.120
Comorbidity, N (%)			
Coronary artery disease	603 (23.1)	481 (26.5)	0.010
Congestive heart failure	1213 (46.4)	894 (49.2)	0.070
Valvular disease	338 (12.9)	325 (17.9)	0.029
Hypertension	923 (35.3)	606 (33.4)	0.171
COPD	376 (14.4)	251 (13.8)	0.585
Malignant cancer	414 (15.8)	373 (20.5)	0.001
Renal failure	1027 (39.3)	725 (39.9)	0.696
Liver disease	268 (10.3)	211 (11.6)	0.836
Rheumatoid arthritis	56 (2.1)	30 (1.7)	0.244
Septic shock	645 (24.7)	696 (38.3)	< 0.001

Table 1. Comparisons of clinical characteristics between survivors and non-survivors. N, number; SD, standard deviation; BMI, body mass index; SOFA, Sequential Organ Failure Assessment; SAPS II, Simplified Acute Physiology Score; ACR, anion gap-to-calcium ratio; Glu, glucose; Lac, lactate; PLT, platelets; INR, international normalized ratio; BUN, blood urea nitrogen; WBC, white blood cell; RDW, red blood cell distribution width; RBC, red blood cell; HCT, hematocrit; Hb, hemoglobin; Cr, creatinine; MAP, mean arterial pressure; Resp rate, respiratory rate; Temp, temperature; SpO₂, blood oxygen saturation; Ventilation, the patient needs assisted ventilation on the first day; COPD, chronic obstructive pulmonary disease.

analyses. The adjusted results showed that ACR (OR 1.342, 95% CI 1.180–1.526, $P < 0.001$) was independent predictors for death within 30 days from any reason in sepsis patients with DM (Table 2).

Receiver operating characteristic (ROC) curve was used to evaluate the predictive performance of ACR, SOFA, AG, and Ca for 30-day mortality all-cause mortality of sepsis patients with DM, and ACR has better discrimination and accuracy compared to the other variables (Fig. 2). The areas under ROC of ACR was 0.622 (95% CI: 0.605–0.640), with a sensitivity of 0.501 and a specificity of 0.704. All patients were next divided into two groups according to the cut-off value of ACR: low-ACR group (ACR < 1.89, 2481 patients) and high-ACR group (ACR ≥ 1.89, 1948 patients). Kaplan–Meier analysis was performed between the two groups. As shown in Fig. 3A, the survival curve of the high-ACR group was significantly lower than that of the low-ACR group (log-rank test, $P < 0.001$).

Variables	OR	OR (95% CI)		P value
		Lower	Upper	
Age, years	1.019	1.013	1.026	<0.001
BMI, kg/m ²	0.984	0.975	0.993	0.001
SOFA score	1.028	1.002	1.054	0.037
SAPS II score	1.024	1.017	1.031	<0.001
ACR	1.342	1.180	1.526	<0.001
BUN, mg/dL	1.001	0.998	1.003	0.665
Cr, mg/dL	0.916	0.872	0.962	0.010
RDW, %	1.068	1.040	1.096	<0.001
Glu, mg/dL	1.001	1.000	1.002	0.004
WBC×10 ⁹ /L	0.997	0.991	1.003	0.255
Ventilation	1.532	1.334	1.758	0.001
Coronary artery disease	1.125	0.969	1.306	0.122
Malignant cancer	1.317	0.998	1.558	0.056
Septic shock	1.372	0.925	2.034	0.116

Table 2. Multivariate regression results of 30-day mortality for septic patients with diabetes mellitus. OR, odds ratio; CI, confidence interval; BMI, body mass index; SOFA, Sequential Organ Failure Assessment; SAPS II, Simplified Acute Physiology Score; ACR, anion gap-to-calcium ratio; BUN, blood urea nitrogen; RDW, red blood cell distribution width; Cr, creatinine; Glu, glucose; WBC, white blood cell; Ventilation, the patient needs assisted ventilation on the first day of admission.

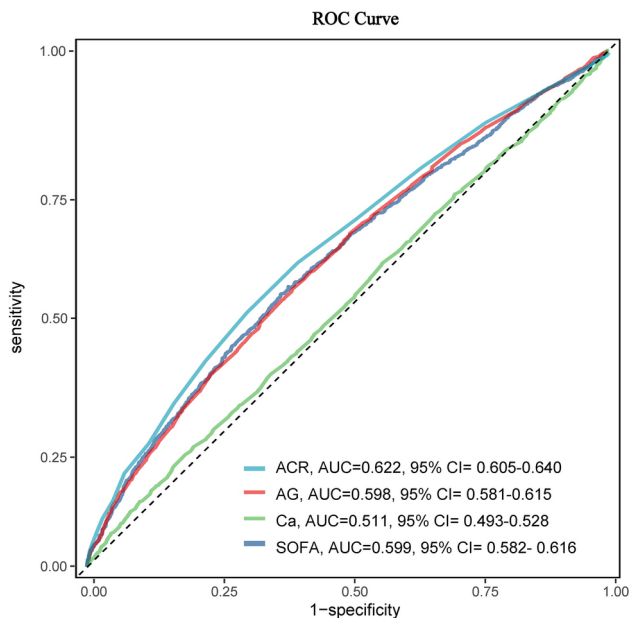


Fig. 2. ROC curve of ACR, SOFA, AG and Ca for predicting the all-cause mortality in the overall study population. The optimal cut-off value of ACR for the 30-day all-cause mortality was 1.89, with a sensitivity of 0.501 and a specificity of 0.704. Abbreviations: ROC, receiver operating characteristic; ACR, anion gap-to-calcium ratio; AUC, area under curve; AG, anion gap; Ca, calcium; SOFA, Sequential Organ Failure Assessment.

Propensity score analysis

To reduce the influence of confounding bias, the PS analyses were next performed in our studies. After PS (1:1), 1358 patients with high-ACR and 1358 patients with low-ACR were included in the final analysis. All covariates were evenly distributed across the two groups after PS analyses (Table 3). Multivariable logistic regression analysis was used to adjust for remaining confounding factors, the results showed that the survival probability 30-day of sepsis patients with DM was significantly higher in the low-ACR than those in high-ACR group (OR 1.622, 95% CI 1.391–1.891, $P < 0.001$). Multivariable analysis with IPW according to the propensity score also yielded similar results (OR 1.382, 95% CI 1.226–1.558, $P < 0.001$) (Table 4, Fig. 3B).

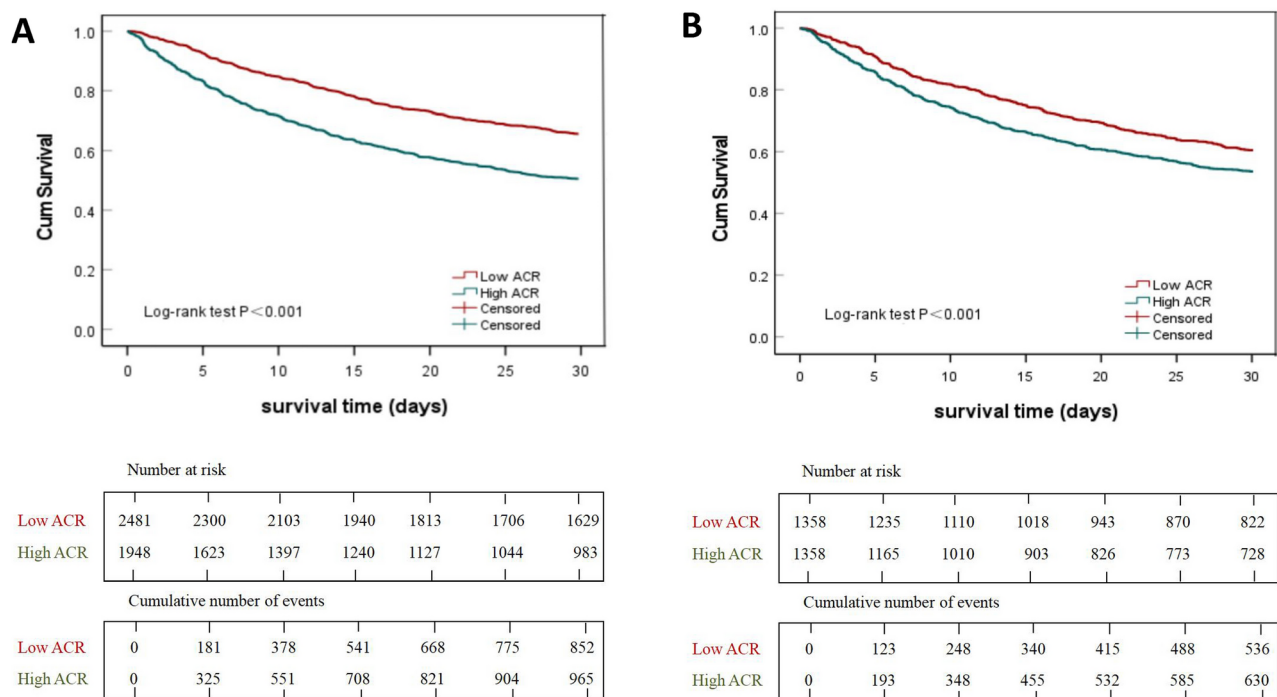


Fig. 3. Kaplan–Meier survival curves of septic patients with DM at 30-day, with high (green, $ACR \geq 1.89$) and low (red, $ACR < 1.89$) ACR. (**A** and **B**) were reflected the results before and after propensity score matching, respectively. Abbreviations: DM, diabetes mellitus; ACR, anion gap-to-calcium ratio.

The subgroup analyses

Subgroup analysis demonstrated a significant correlation between ACR and a worse prognosis in subgroups of gender, age, comorbidity, and invasive mechanical ventilation. When stratified analysis was performed based on whether septic shock was present, the results showed that patients with high level of ACR have a higher risk of death within 30 days than those with low level of ACR in both patients with or without septic shock (Table S5).

Discussion

Our study observed that the ACR was significantly associated with the risk of death within 30 days in sepsis patients with DM from ICU. Importantly, these results were also robust in a series of sensitivity analyses and showed predictive value of ACR for adverse clinical prognosis. Therefore, our study suggested that ACR may offer clinicians a valuable insights for guiding interventions in this high-risk population.

Sepsis is the life-threatening organ dysfunction caused by the dysfunction of body regulation and infection²⁰. Due to its high fatality rate, sepsis has also been identified by the World Health Organization as a global health priority^{20–22}. Currently, Sequential Organ Failure Assessment (SOFA), Simplified Acute Physiological Score II (SAPS II), and other scoring systems are commonly used to evaluate the severity of the condition and the risk of death for patients with sepsis^{23,24}. However, these methods are not only cumbersome in practice, but also lack specificity for sepsis patients with different characteristics. Therefore, it is necessary to further develop some new simple and effective prediction indicators. Interestingly, in this study, we found for the first time that ACR levels were associated with the risk of death in sepsis patients with DM. Our work extends the understanding established in several previous studies. In 2018, Mohr et al. found that anion gap was associated with the risk of death in patients with sepsis²⁵. In 2022, Li et al. reported that the level of blood calcium were also associated with the risk of death in patients with severe infections²⁶. However, these studies did not further investigate the association between a composite indicator of anion gap and calcium (ACR) and the risk of death in patients with sepsis. Our work suggest that ACR may be a promising indicator in predicting risk of death in sepsis patients with DM.

Septic shock is one of the important causes of mortality²⁷. Our work showed that sepsis patients with high level of ACR have a higher risk of death within 30 days than those with low level of ACR regardless of septic shock, which are in general agreement with several previously published studies. In 2016, Ganesh K et al. reported that high level of AG was associated with an increased risk of death from any cause in patients with septic shock²⁸. In 2017, another cohort study by performed by He et al. also found that sepsis patients with high level of AG had a higher risk of long-term death than those with low level of AG, regardless of whether they had concurrent septic shock²⁹. However, these published studies only looked at the association between AG and the risk of death in sepsis, and our study is the first to report that ACR levels are also associated with the prognosis among sepsis patients with or without septic shock.

Variables	Original cohort			Matched cohort		
	Low ACR (2481)	High ACR (1948)	SMD	Low ACR (1358)	High ACR (1358)	SMD
Age [mean (SD)], years	71.8 ± 12.8	71.2 ± 12.8	-0.042	72.0 ± 12.2	72.1 ± 12.6	0.010
Gender, male, N(%)	1441 (58.1)	1137 (58.4)	0.006	806 (59.4)	763 (56.2)	0.016
BMI [mean (SD)], kg/m ²	30.1 ± 7.8	29.9 ± 7.8	-0.027	29.6 ± 7.5	29.7 ± 7.5	0.031
SOFA score [mean (SD)]	6 ± 3	8 ± 4	0.598	7 ± 3	7 ± 3	0.047
SAPS II score [mean (SD)]	42 ± 13	50 ± 14	0.547	46 ± 13	47 ± 13	0.054
Ventilation, N(%)	1221 (49.2)	1012 (52.0)	0.055	669 (49.3)	683 (50.3)	0.013
Laboratory tests						
Cr [mean (SD)], mg/dL	1.5 ± 1.1	2.9 ± 2.2	0.617	1.9 ± 1.3	2.2 ± 1.8	0.121
Glu [mean (SD)], mg/dL	171.7 ± 68.0	188.8 ± 83.1	0.206	184.3 ± 77.1	184.0 ± 74.5	0.024
WBC × 10 ⁹ /L [mean (SD)]	12.7 ± 11.0	14.8 ± 10.4	0.198	13.9 ± 13.8	14.1 ± 10.0	0.002
RDW [mean (SD)], %	16.0 ± 2.5	16.5 ± 2.6	0.193	16.3 ± 2.6	16.3 ± 2.5	-0.009
RBC × 10 ¹² /L [mean (SD)]	3.5 ± 0.7	3.3 ± 0.7	-0.171	3.4 ± 0.7	3.4 ± 0.7	0.005
BUN [mean (SD)], mg/dL	32.9 ± 21.1	49.4 ± 31.1	0.531	39.7 ± 23.5	41.6 ± 26.2	0.045
Vital signs						
Heart rate [mean (SD)], times/min	89.7 ± 20.2	93.1 ± 21.5	0.155	92.9 ± 21.5	91.9 ± 20.9	0.002
Resp rate [mean (SD)], times/min	20.1 ± 6.4	21.1 ± 6.4	0.156	20.9 ± 6.5	20.9 ± 6.3	0.029
Temp [mean (SD)], °C	36.9 ± 0.7	36.8 ± 0.8	-0.092	36.8 ± 0.7	36.8 ± 0.8	-0.005
Comorbidity, N (%)						
Coronary artery disease	567 (22.9)	517 (26.5)	0.098	339 (25.0)	346 (25.5)	0.006
Congestive heart failure	1100 (44.3)	1007 (51.7)	0.147	657 (48.4)	681 (50.1)	0.027
COPD	1100 (44.3)	1007 (51.7)	-0.037	190 (14.0)	191 (14.1)	0.002
Malignant cancer	456 (18.4)	331 (17.0)	0.083	248 (18.3)	243 (17.9)	0.018
Renal failure	860 (34.7)	892 (45.8)	0.223	549 (40.4)	572 (42.1)	0.034
Septic shock	580 (23.4)	761 (39.1)	0.321	432 (31.8)	443 (32.6)	0.032

Table 3. Comparison of baseline characteristics between the original cohort and the matched cohort by using propensity score analysis. N, number; SD, standard deviation; SMD, standardized mean difference; ACR, anion gap-to-calcium ratio; BMI, body mass index; SOFA, Sequential Organ Failure Assessment; SAPS II, Simplified Acute Physiology Score; Cr, creatinine; Glu, glucose; WBC, white blood cell; RDW, red blood cell distribution width; RBC, red blood cell; BUN, blood urea nitrogen; Resp rate, respiratory rate; Temp, temperature; Ventilation, the patient needs assisted ventilation on the first day; COPD, chronic obstructive pulmonary disease.

Our study has several limitations. First, this is a data study based on a single center, and practice observed in this cohort may not be representative of other settings. Second, the cohort was generated from retrospective data, and therefore confounding factors may contribute to the unreliability of the results, and omission of unmeasured confounders may result in a biased estimation. Indeed, we used PSM to control for measured confounders, but there may be unmeasured or unknown confounders that were not included in the matching process. The results of PSM also depend on the quality of the input data, and errors or biases in the collection of single-center data may affect the calculation and matching process of propensity scores. Third, our analysis was limited to all-cause mortality, mainly because critically ill patients are often complicated by multiple diseases and it is difficult to select a specific cause of death for such patients. However, all-cause mortality is also an objective and useful endpoint that has been widely used in various clinical studies. Fourth, although the AUROC was greater than 0.5, it was less than 0.7 (0.622), indicating scope for improvement. The specificity for ACR suggests that it is effective at identifying the sepsis patients with DM who died in the short term. However, the relatively lower sensitivity also demonstrates the necessity for future studies with larger populations to validate the clinical utility of ACR as a prognostic indicator. Finally, this work also included some variables with missing information. However, we have removed the variables with more than 20% missing data and processed the remaining variables using multiple imputation to minimize bias.

Despite these limitations, our findings also have strengths. First, this study is the first to investigate the association between ACR and death in sepsis patients with DM, and confirmed that the level of ACR was associated with death within 30 days in critically ill sepsis patients with DM. Second, we used several sensitivity analyses to assess the robustness of the findings, which further improve the reliability of our findings. Third, our study provides a new perspective for predicting the prognosis of patients with sepsis and DM. If these results are further validated in future studies, which have a potential to improve strategies for assessing the condition of sepsis patients with DM.

Our results suggested that ACR, as a simple and practical clinical parameter, may be an independent prognostic indicator of death with 30 days in critically ill sepsis patients with DM. But these results still need to be further validated in future studies.

Analysis	30-day mortality	P value
No. of events/no. of patients at risk (%)		<0.001
Low ACR	852/2481 (34.3)	–
High ACR	965/1948 (49.5)	–
Crude analysis-OR (95% CI)	1.877 (1.662, 2.119)	<0.001
Multivariable analysis-OR (95% CI)*	1.342 (1.180, 1.526)	<0.001
Propensity-score analyses-OR (95% CI)		
With matching‡	1.622 (1.391, 1.891)	<0.001
With propensity score IPTW†	1.382 (1.226, 1.558)	<0.001

Table 4. Association of ACR and 30-day mortality in Septic patients with DM by using in the crude analysis, multivariable analysis, and propensity-score analyses. OR, odds ratio; CI, confidence interval; IPTW, inverse probability-of-treatment weighting; PSM, propensity score matching. *Shown is the odds ratio from the multivariable logistic regression model with additional adjustment for age, body mass index, SOFA score, SAPS II score, anion gap-to-calcium ratio, blood urea nitrogen, red blood cell distribution width, creatinine, glucose, white blood cell, the patient needs assisted ventilation on the first day of admission, coronary artery disease, metastatic cancer and septic shock. The analysis included all 4429 patients. The “event” shown is 30-day mortality; ‡Shown is the odds ratio from a multivariable logistic regression model with the same covariates with matching according to the propensity score. The analysis included 2716 patients (1358 in low-ACR group and 1358 in high-ACR group). †Shown is the odds ratio from the multivariable logistic regression model with 30-day mortality as the binary response with the remained unbalanced covariates with inverse probability-of-treatment weighting according to the propensity score. The analysis included all the patients.

Methods

Study setting

This is a longitudinal, single-center, retrospective cohort study and followed the reporting guidelines of the STrengthening the Reporting of OBservational studies in Epidemiology (STROBE) statement³⁰. The publicly available database of Medical Information Mart for Intensive Care Database IV (MIMIC IV, version 3.0) was employed for this work. MIMIC IV was a publicly available real-world clinical database, and maintained by Beth Israel Deaconess Medical Center (BIDMC, Boston, MA, USA) and Massachusetts Institute of Technology (MIT, Cambridge, MA, USA)³¹. We completed the “Data or Specimens Only Research” course of Collaborative Institutional Training Initiative Program before extracted the MIMIC-IV clinical data. Institutional review board (IRB) approval from Chengdu First People’s Hospital was exempted because this work was an analysis of the publicly available database with pre-existing IRB approval (Certification Number: 62955778).

The dictionary of codes for the International Classification of Diseases and Ninth Revision (ICD-9) codes dictionary were used to screen and extract sepsis, septic shock, and diabetes mellitus from the MIMIC-IV database. Definitions of sepsis was defined by a condition with life-threatening organ dysfunction caused by a dysregulated host response to infection. Septic shock was defined by the presence of hypotension requiring vasopressor support in patients with sepsis²⁰.

Study population

All patients diagnosed with both sepsis and diabetes mellitus in ICU were included in this study. Detailed inclusion criteria were patients: (1) with sepsis or septic shock; (2) with diabetes mellitus; (3) age \geq 18 years; (4) with data of anion gap and blood calcium; (5) admitted to the ICU.

Data collection and variable extraction

The categorical and continuous variables were extracted from the first day of ICU admission. Categorical variables included gender, ventilation, comorbidities (coronary artery disease, congestive heart failure, valvular disease, chronic obstructive pulmonary disease, hypertension, malignant cancer, renal failure, liver disease, rheumatoid arthritis and septic shock). Continuous variables included the age at admission, body mass index (BMI), sequential organ failure assessment (SOFA), simplified acute physiology score II (SAPS II), AG, blood calcium, glucose, lactate, platelets, international normalized ratio (INR), blood urea nitrogen (BUN), white blood cell (WBC), RDW, red blood cell (RBC), hematocrit, hemoglobin, creatinine (Cr), mean arterial pressure, respiratory rate, temperature, and blood oxygen saturation. Anion gap-to-calcium ratio (ACR) is the ratio of anion gap and calcium. The primary endpoint was the 30-day mortality, which was defined as the status of patient survival within 30 days from admission.

Statistical analysis

The data were analyzed using the statistical software packages R 4.1.2 (<http://www.R-project.org>, The R Foundation) and Statistical Package for the Social Sciences, version 27 (SPSS 27), which is a comprehensive software program designed for data analysis, reporting, and predictive analytics. We filled in the variables with missing values (<20%) by the multiple interpolation method. Categorical variables were expressed as the number of the population with the percentage (%). Continuous variables which were non-normally distributed were expressed as median and interquartile range (IQR), and normally distributed variables were expressed

as the mean and standard deviation (Mean \pm SD). Receiver operating characteristic (ROC) curve analysis was conducted to evaluate the predictive performance of ACR for 30-day mortality. The corresponding sensitivity and specificity were calculated using the cut-off value of ACR, which was determined by maximising the Youden index.

The enrolled patients were next divided into high- and low-ACR groups based on the obtained cut-off values. PS matching (using standard caliper value of 0.2) was used to account for the baseline differences between high and low groups³². Patients with high-ACR were matched to those with low-ACR by nearest neighbor matching. Standardized mean difference (SMD) was calculated before and after matching to examine whether the PSM reduced the differences in pretreatment covariates between low- and high-ACR groups. Multivariable logistic regression model was further used to adjust for residual imbalance by including parameters with $P < 0.1$ and potential confounders judged by clinical expertise. Inverse probability weighting (IPW) is used to assess the reliability of the results. In the IPW analysis, the predicted probabilities from the propensity-score model were used to calculate the stabilized inverse-probability-weighting weight³³. A P value less than 0.05 was considered to be statistically significant.

Several prespecified subgroup analyses were performed by restricting (1) the age of patients at admission (≤ 65 and > 65 years); (2) gender; (3) whether patients received mechanical ventilation (yes or no); (4) the type of comorbidity.

Data availability

MIMIC-IV v3.0 (Medical Information Mart for Intensive Care IV, version: 3.0) was accessed on July 23, 2024 from <https://physionet.org/content/mimiciv/3.0>.

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Author contributions

C.J., P.L., L.K., L.M. and Y.N.B. conceived and designed research; C.J., L.K. and L.M. collected data and conducted research; Y.N.B., G.J. and L.J.J. analyzed and interpreted data; C.J. wrote the initial paper; All authors revised the paper; C.J. had primary responsibility for final content. All authors reviewed the manuscript.

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Declarations

Competing interests

The authors declare no competing interests.

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