Check for updates

OPEN ACCESS

EDITED BY Guangyu Zhu, Xi'an Jiaotong University, China

REVIEWED BY Xian-Tao Zeng, Wuhan University, China Tiancai Wen, China Academy of Chinese Medical Sciences, China

*CORRESPONDENCE Jun Lyu, ⊠ lyujun2020@jnu.edu.cn Jian-Rui Wei, ⊠ jianruiw2021@163.com

[†]These authors have contributed equally to this work

RECEIVED 09 March 2023 ACCEPTED 30 May 2023 PUBLISHED 04 July 2023

CITATION

Zhu J-L, Xu X-M, Yin H-Y, Wei J-R and Lyu J (2023), Development and validation of a nomogram for predicting hospitalization longer than 14 days in pediatric patients with ventricular septal defect—a study based on the PIC database. *Front. Physiol.* 14:1182719. doi: 10.3389/fphys.2023.1182719

COPYRIGHT

© 2023 Zhu, Xu, Yin, Wei and Lyu. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

Development and validation of a nomogram for predicting hospitalization longer than 14 days in pediatric patients with ventricular septal defect—a study based on the PIC database

Jia-Liang Zhu^{1,2†}, Xiao-Mei Xu^{1,2†}, Hai-Yan Yin¹, Jian-Rui Wei²* and Jun Lyu^{3,4}*

¹Department of Intensive Care Unit, The First Affiliated Hospital of Jinan University, Guangzhou, China, ²Guangzhou Women and Children's Medical Center, Guangzhou Medical University, Guangzhou, China, ³Department of Clinical Research, The First Affiliated Hospital of Jinan University, Guangzhou, China, ⁴Guangdong Provincial Key Laboratory of Traditional Chinese Medicine Informatization, Guangzhou, China

Background: Ventricular septal defect is a common congenital heart disease. As the disease progresses, the likelihood of lung infection and heart failure increases, leading to prolonged hospital stays and an increased likelihood of complications such as nosocomial infections. We aimed to develop a nomogram for predicting hospital stays over 14 days in pediatric patients with ventricular septal defect and to evaluate the predictive power of the nomogram. We hope that nomogram can provide clinicians with more information to identify high-risk groups as soon as possible and give early treatment to reduce hospital stay and complications.

Methods: The population of this study was pediatric patients with ventricular septal defect, and data were obtained from the Pediatric Intensive Care Database. The resulting event was a hospital stay longer than 14 days. Variables with a variance inflation factor (VIF) greater than 5 were excluded. Variables were selected using the least absolute shrinkage and selection operator (Lasso), and the selected variables were incorporated into logistic regression to construct a nomogram. The performance of the nomogram was assessed by using the area under the receiver operating characteristic curve (AUC), Decision Curve Analysis (DCA) and calibration curve. Finally, the importance of variables in the model is calculated based on the XGboost method.

Results: A total of 705 patients with ventricular septal defect were included in the study. After screening with VIF and Lasso, the variables finally included in the statistical analysis include: Brain Natriuretic Peptide, bicarbonate, fibrinogen, urea, alanine aminotransferase, blood oxygen saturation, systolic blood pressure, respiratory rate, heart rate. The AUC values of nomogram in the training

Abbreviations: ADA, Adenosine deaminase; ALT, Alanine aminotransferase; BNP, brain natriuretic peptide; DBP, Diastolic blood pressure; Hr, Heart rate; INR, International normalized ratio; Na, Serum sodium; PIC, Paediatric Intensive Care; PLT, Platelets; RCB, Red blood cells; Rr, Respiratory rate; SBP, Systolic blood pressure; Spo2, Blood oxygen saturation; T, Body temperature; Tbil, Total bilirubin; WBC, White blood cells.

cohort and validation cohort were 0.812 and 0.736, respectively. The results of the calibration curve and DCA also indicated that the nomogram had good performance and good clinical application value.

Conclusion: The nomogram established by BNP, bicarbonate, fibrinogen, urea, alanine aminotransferase, blood oxygen saturation, systolic blood pressure, respiratory rate, heart rate has good predictive performance and clinical applicability. The nomogram can effectively identify specific populations at risk for adverse outcomes.

KEYWORDS

ventricular septal defect, nomogram, PIC database, retrospective study, XGBoost

Introduction

Ventricular septal defect (VSD) is the most common form of congenital heart disease, accounting for approximately 40% of congenital heart disease cases (Penny and Vick, 2011). VSD leads to blood shunting, resulting in increased pulmonary blood circulation volume and pathological changes in the pulmonary blood vessels, which makes children with VSD particularly prone to pulmonary infection. As the disease progresses, when the pulmonary circulation pressure is higher than the systemic circulation pressure, the blood flows from the right to the left ventricle, which increases the preload of the left ventricle that easily leads to heart failure. The alveolar development of young infants is not perfect, the respiratory system is immature, and the synthesis function of alveolar type II epithelial cells is deficient, resulting in less alveolar surfactant production, and so the respiratory function is immature. Both pulmonary hypoplasia and infection may lead to a decrease in the arterial partial pressure of oxygen, which further leads to respiratory rate changes.

Occurrence of infection may be a risk factor for prolonged hospital stay in patients with VSD. Extended hospital stays may further increase the likelihood of nosocomial infections. Pulmonary infection may cause pulmonary interstitial edema, resulting in decreased pulmonary ventilation and hypoxemia, which ultimately leads to rapid breathing. The infection diagnoses of some cases are not clear, making it difficult to determine whether a hospital stay was prolonged. However, the respiratory rate (Rr) is an easy metric to monitor and its measurement accuracy is high, and hence it may have predictive value for the length of hospital stay. A study found that a higher respiratory rate on admission was associated with an increased risk of in-hospital mortality in patients admitted from nursing homes (Myint et al., 2011). We therefore set the outcome of the present study as a hospital stay longer than 14 days. We aimed to develop a nomogram to assess the risk of hospitalization for longer than 14 days in pediatric patients with VSD. We hope that clinicians can make timely adjustments to treatment regimens based on changes in risk factors in the nomogram model to reduce hospital stays in pediatric patients with VSD.

Methods

Data source

This is a single center retrospective study. Data for the study were extracted from the Paediatric Intensive Care (PIC) database. This is a

large, single-center, pediatric-specific database that includes the clinical data of all patients admitted to different ICUs at the Children's Hospital Zhejiang University School of Medicine during 2010–2018, and includes the admission information of 13,499 cases from among 12,881 different pediatric patients (Yang et al., 2020). The data include patient demographics, drug use, fluid balance, comprehensive laboratory test results, and microbiological information obtained throughout a hospital stay. It also includes vital sign information collected from the anesthesia information management system during surgery. According to the requirements of the database, we completed the relevant training courses and passed the examination to obtain an access certificate (no. 45848364). The private information of patients has been altered in the PIC database, and so this study did not require approval from the hospital ethics committee.

Participant selection

All of the included patients had been diagnosed with a ventricular septal defect.

Study metrics

We used Structured Query Language to extract metrics within 24 h of admission from the PIC database. These included the basic information, vital signs, and laboratory test indicators of the patients. Basic information included sex, marital status, language. Since the subjects of this study were infants and young children, we did not include the language and marital statuses. Vital-sign information included body temperature (T), heart rate (Hr), respiratory rate, systolic blood pressure (SBP), diastolic blood pressure (DBP), and blood oxygen saturation (SpO2). We also collected laboratory variables including adenosine deaminase (ADA), serum albumin, alanine aminotransferase (ALT), serum creatinine, total bilirubin (Tbil), globulin, hemoglobin, lymphocyte count, red blood cell count (RBC), white blood cell count (WBC), platelet count (PLT), eosinophils, D-dimer, fibrinogen, international normalized ratio (INR), serum sodium (Na), brain natriuretic peptide (BNP) and bicarbonate (Wu et al., 2021).

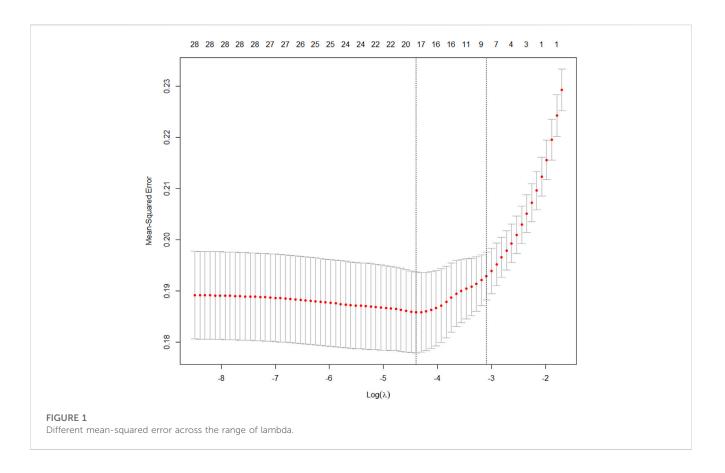
Research design

We used variance inflation factor (VIF) to assess whether there was collinearity between variables, and excluded variables with VIF

TABLE 1 Baseline characteristics.

Variables	Total (<i>n</i> = 705)	Training cohort ($n = 423$)	Verification cohort ($n = 282$)	<i>p</i> -value
Gender (n,%)				0.296
Male	362 (51%)	224 (53%)	138 (49%)	
Female	343 (49%)	199 (47%)	144 (51%)	
Body temperature (°C)	36.70 (36.40, 36.90)	36.70 (36.50, 36.90)	36.60 (36.40, 36.90)	0.070
Heart Rate(beats/min)	128 (114, 138)	128 (112, 136)	130 (116, 138)	0.093
Respiration rate (beats/min)	30 (28, 34)	30 (26, 34)	30 (28, 34)	0.133
DBP (mmHg)	54 (46, 62)	54 (46, 64)	55 (45, 62)	0.248
SBP (mmHg)	98 (89, 106)	98 (89, 106)	98 (88, 107)	0.651
Spo2 (%)	98 (97, 99)	98 (97, 99)	98 (97, 99)	0.862
Adenosine deaminase	14.0 (11.7, 16.8)	14.0 (11.7, 16.8)	14.1 (11.8, 17.1)	0.743
Albumin	43.9 (41.7, 46.1)	43.8 (41.6, 46.0)	44.0 (41.7, 46.2)	0.512
ALT	18 (13, 27)	18 (13, 27)	18 (13, 26)	0.663
AST	41 (33, 57)	41 (32, 56)	41 (34, 58)	0.391
Urea	3.79 (2.66, 4.70)	3.80 (2.64, 4.71)	3.75 (2.71, 4.67)	0.657
Creatinine (umol/L)	43 (39, 48)	43 (39, 48)	44 (38, 47)	0.775
Tbil (umol/L)	7 (5, 10)	7 (5, 10)	6 (5, 10)	0.120
Globulin (g/L)	21.0 (18.0, 23.8)	21.3 (18.1, 23.8)	20.6 (17.7, 23.7)	0.235
Hemoglobin (g/dL)	119 (110, 126)	119 (110, 126)	118 (110, 126)	0.981
Lymphocyte count (×109/L)	5.06 (3.82, 6.62)	4.87 (3.63, 6.46)	5.22 (4.13, 6.80)	0.019
PLT (×109/L)	333 (271, 399)	331 (272, 394)	335 (267, 401)	0.650
RBC (×1012/L)	4.42 (4.13, 4.67)	4.42 (4.14, 4.66)	4.42 (4.10, 4.69)	0.989
WBC (×109/L)	8.87 (7.30, 10.99)	8.78 (7.19, 10.98)	8.89 (7.65, 11.18)	0.173
Eosinophils (×109/L)	0.22 (0.13, 0.35)	0.23 (0.14, 0.36)	0.21 (0.13, 0.35)	0.460
D-dimer (mg/L)	0.21 (0.14, 0.39)	0.21 (0.15, 0.40)	0.21 (0.14, 0.38)	0.434
Fibrinogen (g/L)	1.97 (1.67, 2.32)	2.00 (1.65, 2.36)	1.96 (1.68, 2.23)	0.356
INR	0.98 (0.93, 1.03)	0.98 (0.93, 1.03)	0.98 (0.93, 1.03)	0.536
Bicarbonate (mmol/L)	22.10 (20.80, 23.30)	22.20 (20.80, 23.30)	22.00 (20.90, 23.30)	0.924
Na (mmol/L)	135 (133, 137)	135 (133, 137)	135 (133, 138)	0.801
BNP (pg/mL)	297 (106, 1,582)	283 (106, 1,492)	320 (108, 1,608)	0.858

values > 5. The existence of a variable with multicollinearity means that the corresponding information provided by this variable is redundant in the presence of other variables. The variables were incorporated into the least absolute shrinkage and selection operator model (Lasso). The LASSO regression method was used to reduce the dimensionality of variables with certain correlations, and variables with non-zero coefficient characteristics were selected. Calculate the cut-off point of continuous variables, and divide the continuous variables into two groups according to the cut-off point, one is above the cut-off point, and the other is below the cutoff value. The Odds ratio of the variables was calculated using a logistic regression model and represented by a forest plot. Eligible variables were incorporated into logistic regression to build the nomogram. The performance of the nomogram was assessed using the area under the receiver operating characteristic curve (AUC), with AUC >0.7 indicating good model performance. Calibration curves and Decision Curve Analysis (DCA) were used to evaluate the predictive and clinical performance of the nomogram. The feature importance of variables in the model is calculated based on the XGboost method. Restricted cubic splines (RCS) were used to explore the relationship between respiratory rate and hospital stay longer than 14 days.



Statistical analysis

Continuous variables were represented by medians and quartiles, and differences between groups were determined using analysis of variance. Categorical variables were expressed as frequency and percentage values, and differences between groups were determined using chi-square tests. We used R software (4.2.0) for statistical analysis. The R packages used in this study include gtsummary (Table 1), dplyr, glmnet, foreign, ISwR, car, rms, regplot, pROC, nricens, forestmodel, and tidymodels.

Results

We extracted data from the PIC database on 705 patients with a diagnosis of ventricular septal defect, 362 of whom were male. We randomly assigned 60% of patients to the training cohort and 40% to the validation cohort. Except for lymphocyte counts, baseline data for both cohorts were balanced (*p*-value >0.05). The baseline data of the two groups of patients are shown in Table 1. The VIF of all variables is less than 5, indicating that there is no multicollinearity among the variables. The VIF of the variable is shown in Supplementary Table S1; Figure 1 shows the different mean squared error within the range of log(lambda). Based on LASSO analysis, a total of 9 most likely non-zero coefficient characteristic variables were screened, including heart rate, respiratory rate, SBP, Spo2, urea, ALT, fibrinogen, bicarbonate, and BNP. The cutoff values for heart rate, respiratory rate, SBP, Spo2, urea, ALT, fibrinogen, bicarbonate and BNP were 123, 31, 89, 96, 23, 2.14,

1.56, 24.5, 270, respectively. The variables were incorporated into the logistic regression model to determine the Odds ratio, and the results were represented by a forest plot (Figure 2). The above 9 variables were incorporated into the logistic regression model, and a nomogram was established (Figure 3). The AUC of nomogram in the training cohort and validation cohort were 0.812 and 0.736, respectively, indicating that nomogram has good predictive performance (Figure 4). The calibration curve indicated that the nomogram had good consistency between the training and validation cohorts in predicting the length of hospital stay in patients with ventricular septal defect longer than 14 days (Figure 5). Figure 6 shows the DCA of the nomogram, showing a net benefits in both the training and validation cohorts. Figure 7 shows the importance of each variable, BNP is the most important variable in the nomogram.

Discussion

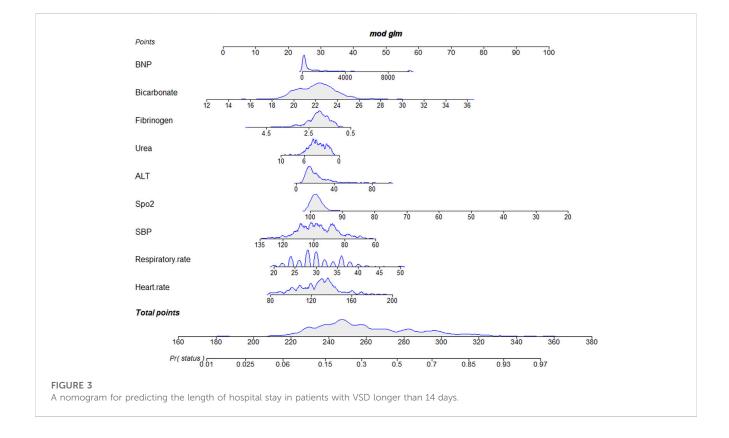
Ventricular septal defect is the most common congenital heart disease caused by hypoplasia of the ventricular septum during the embryonic stage. In the early stage of ventricular septal defect, blood is shunted from the left to the right ventricle, which leads to an increase in the pulmonary blood circulation volume that causes pulmonary interstitial edema, which is prone to pulmonary infection.

Previous studies have shown that risk factors for unplanned readmission to pediatric intensive care units include age, disease type, disease severity, source of admission, unplanned initial admission, time of initial discharge, and respiratory support at

Variable		Ν	Odds ratio		р
Heart.rate	Below 123	279	.	Reference	
	Over 123	426	·•	1.68 (1.09, 2.59)	0.019
Respiratory.rate	Below 31	431	.	Reference	
	Over 31	274	—	1.93 (1.25, 2.97)	0.003
SBP	Below 89	192	•	Reference	
	Over 89	513	·	0.82 (0.55, 1.20)	0.303
Spo2	Below 96	118	.	Reference	
	Over 96	587	·	0.65 (0.41, 1.02)	0.059
ALT	Below 23	487	•	Reference	
	Over 23	218	·•	1.54 (1.03, 2.29)	0.033
Urea	Below 2.14	114	.	Reference	
	Over 2.14	591	⊨i	0.87 (0.54, 1.41)	0.584
Fibrinogen	Below 1.56	130	.	Reference	
	Over 1.56	575	·	0.86 (0.54, 1.37)	0.529
Bicarbonate	Below 24.5	634		Reference	
	Over 24.5	71	· · · · · · · · · · · · · · · · · · ·	2.18 (1.24, 3.87)	0.007
BNP	Below 270	341		Reference	
	Over 270	364	·	2.12 (1.42, 3.15)	<0.001

FIGURE 2

Variables that are ultimately included in the nomogram. Abbreviation: Spo2, blood oxygen saturation; SBP, systolic blood pressure; ALT, alanine aminotransferase; BNP: brain natriuretic peptide.



discharge (Odetola et al., 2007; Linton et al., 2009; Bernard and Czaja, 2013; Czaja et al., 2013; Edwards et al., 2013; Mandell et al., 2015; Kaur et al., 2018). Most of these risk factors cannot be changed, but the length of hospital stay may be reduced with optimal treatment. This study identified risk factors for hospitalization longer than 14 days in children with ventricular septal defect,

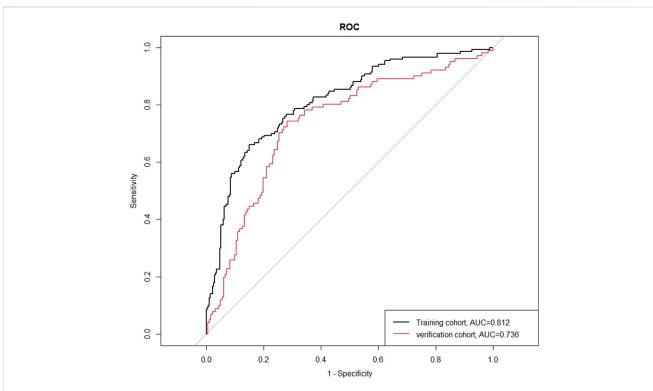
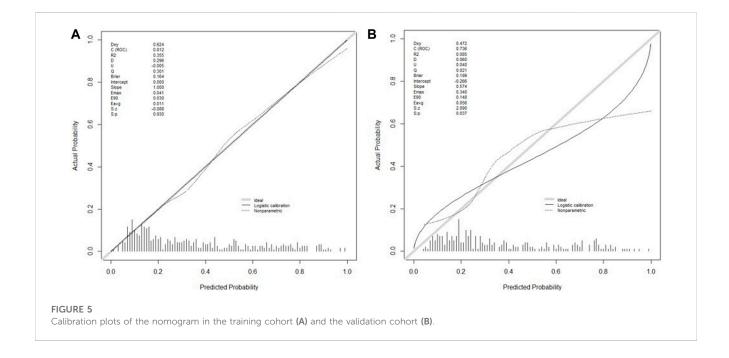


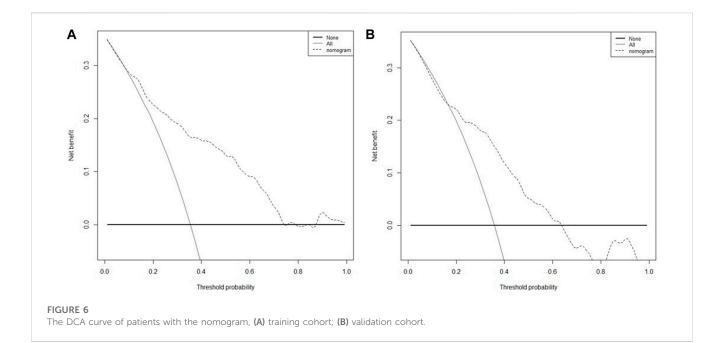
FIGURE 4

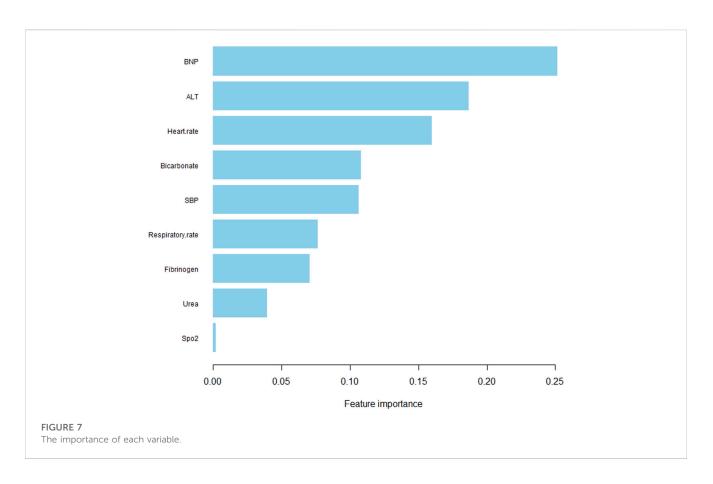
The ROC curves of the training cohort and the verification cohort. Abbreviations: ROC, receiver operating characteristic curve; AUC, area under the receiver operating characteristic curve.



including heart rate, respiratory rate, systolic blood pressure, oxygen saturation, ALT, urea, fibrinogen, bicarbonate, and BNP. The risk of hospitalization for more than 14 days was increased when heart rate >123 beats/min, respiratory rate >31 beats/min, ALT >23 U/L, bicarbonate >24.5 mmol/L, and BNP >240 pg/mL.

When pulmonary infection or heart failure occurs in patients with ventricular septum, pulmonary interstitial edema and pulmonary diffusion dysfunction may occur, resulting in hypoxemia and decreased SpO₂. The compensatory mechanism of the body deepens and accelerates the breathing rate, which





often worsens the illness, so the risk of a hospital stay longer than 14 days increases. Respiratory frequency also plays an important role in monitoring infectious and respiratory diseases. Studies have found that respiratory rate is also an important prognostic factor in pulmonary infection and heart failure (Fischl et al., 1981; Aujesky et al., 2006; Bhatia et al., 2006). The respiratory rate has also been used as an important predictive value in assessing the severity and early stages of exacerbation of acute diseases (McCartan et al., 2021). A study of 705,928 patients found that the respiratory rate on admission had a U-shaped relationship with mortality, with a respiratory rate of 20 breaths/min at admission having the lowest mortality rate (Strauß et al., 2014). Studies have found that the degree of pneumonia infection is related to duration, and so early identification or early antibiotics use can significantly improve the survival rate (Weinstein et al., 1997; Balakrishnan et al., 2000; Rivers et al., 2001; Houck et al., 2004). Therefore, monitoring respiratory rate for early identification of lung infections may improve outcomes. RCS revealed a nonlinear relationship between respiratory rate and a stay longer than 14 days. A respiratory rate exceeding 30 breaths/min increases the risk of a patient being hospitalized for longer than 14 days.

Fibrinogen is a plasma protein coagulation factor synthesized and secreted by hepatocytes, which can participate in hemostasis and thrombosis, regulate coagulation and fibrosis, and prevent the spread of infection and inflammation (Davalos and Akassoglou, 2012). Studies have shown that elevated fibrinogen levels in patients with stable coronary artery disease increase the risk of myocardial infarction and long-term death (Mjelva et al., 2018; Zhang et al., 2020). Studies have shown that in patients with acute aortic dissection, lower fibrinogen levels are associated with an increased risk of in-hospital mortality (Liu et al., 2018). Our study had similar findings that fibrinogen >1.56 g/L was associated with a reduced risk of hospitalization longer than 14 days. In the initial stage of infection, an increase in fibrinogen may be a common finding. Possible mechanisms include activation of the coagulation system by the release of tissue factor when the vascular endothelium is disrupted, which consumes large amounts of fibrinogen and reduces its concentration in the blood. Lower fibrinogen levels imply overactivation of the coagulation system, which may prolong hospital stays and increase the risk of death. In addition, when disseminated intravascular coagulation (DIC) occurs, resulting in a large consumption of coagulation factors, including fibrinogen, it may also lead to poor prognosis.

We hope that clinicians can use the nomogram from this study to accurately identify high-risk groups in pediatric patients with VSD, and provide early interventions to reduce hospital stays, which can help reduce the risk of in-hospital complications and improve outcomes.

The limitations of this study were as follows: First, it had a single-center retrospective design, and its conclusions are not representative of all juvenile patients with ventricular septal defect. Second, we excluded some variables due to high rates of missing data. Due to database limitations, we did not include age, trisomy 21 syndrome, and Eisenmenger syndrome in the analysis. In future studies we hope to include troponin and other variables that can reflect myocardial injury, which may increase the accuracy of the findings.

Conclusion

The nomogram we developed includes heart rate, respiratory rate, SBP, Spo2, ALT, Urea, fibrinogen, bicarbonate, BNP. The model has good predictive performance and clinical utility, and is convenient for clinicians to use. Therefore, the new model can help clinicians assess the risk of hospitalization for longer than 14 days in pediatric patients with VSD.

Data availability statement

Publicly available datasets were analyzed in this study. This data can be found here: http://pic.nbscn.org/.

Author contributions

J-LZ contributed to the study concept and study design. J-LZ and X-MX performed statistical analysis and data interpretation. H-YY performed literature research and data extraction. JL and J-RW were responsible for the quality control of data and algorithms. All authors contributed to the article and approved the submitted version.

Funding

This work was supported by the Scientific research project of Guangdong Bureau of Traditional Chinese Medicine (No. 20221106), the Clinical Frontier Technology Program of the First Affiliated Hospital of Jinan University, China (No. JNU1AF-CFTP-2022-a01235) and the Science and Technology Projects in Guangzhou, China (No. 202201020054, No. 2023A03J1032).

Acknowledgments

The authors would like to thank the PIC project of the Children's Hospital Affiliated to Zhejiang University School of Medicine.

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.

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fphys.2023.1182719/ full#supplementary-material

References

Aujesky, D., Roy, P. M., Le Manach, C. P., Verschuren, F., Meyer, G., Obrosky, D. S., et al. (2006). Validation of a model to predict adverse outcomes in patients with pulmonary embolism. *Eur. Heart J.* 27 (4), 476–481. doi:10.1093/eurheartj/ehi588

Balakrishnan, I., Crook, P., Morris, R., and Gillespie, S. H. (2000). Early predictors of mortality in pneumococcal bacteraemia. *J. Infect.* 40 (3), 256–261. doi:10.1053/jinf.2000. 0653

Bernard, A. M., and Czaja, A. S. (2013). Unplanned pediatric intensive care unit readmissions: A single-center experience. *J. Crit. Care* 28 (5), 625–633. doi:10.1016/j. jcrc.2013.02.008

Bhatia, R. S., Tu, J. V., Lee, D. S., Austin, P. C., Fang, J., Haouzi, A., et al. (2006). Outcome of heart failure with preserved ejection fraction in a population-based study. *N. Engl. J. Med.* 355 (3), 260–269. doi:10.1056/NEJMoa051530

Czaja, A. S., Hosokawa, P. W., and Henderson, W. G. (2013). Unscheduled readmissions to the PICU: Epidemiology, risk factors, and variation among centers. *Pediatr. Crit. Care Med.* 14 (6), 571–579. doi:10.1097/PCC.0b013e3182917a68

Davalos, D., and Akassoglou, K. (2012). Fibrinogen as a key regulator of inflammation in disease. *Semin. Immunopathol.* 34 (1), 43–62. doi:10.1007/s00281-011-0290-8

Edwards, J. D., Lucas, A. R., Stone, P. W., Boscardin, W. J., and Dudley, R. A. (2013). Frequency, risk factors, and outcomes of early unplanned readmissions to PICUs. *Crit. Care Med.* 41 (12), 2773–2783. doi:10.1097/CCM.0b013e31829eb970

Fischl, M. A., Pitchenik, A., and Gardner, L. B. (1981). An index predicting relapse and need for hospitalization in patients with acute bronchial asthma. *N. Engl. J. Med.* 305 (14), 783–789. doi:10.1056/NEJM198110013051402

Houck, P. M., Bratzler, D. W., Nsa, W., Ma, A., and Bartlett, J. G. (2004). Timing of antibiotic administration and outcomes for Medicare patients hospitalized with community-acquired pneumonia. *Arch. Intern Med.* 164 (6), 637–644. doi:10.1001/archinte.164.6.637

Kaur, H., Naessens, J. M., Hanson, A. C., Fryer, K., Nemergut, M. E., and Tripathi, S. (2018). Proper: Development of an early pediatric intensive care unit readmission risk prediction tool. *J. Intensive Care Med.* 33 (1), 29–36. doi:10.1177/0885066616665806

Linton, S., Grant, C., Pellegrini, J., and Davidson, A. (2009). The development of a clinical markers score to predict readmission to paediatric intensive care. *Intensive Crit. Care Nurs.* 25 (6), 283–293. doi:10.1016/j.iccn.2009.07.003

Liu, J., Sun, L. L., Wang, J., and Ji, G. (2018). The relationship between fibrinogen and in-hospital mortality in patients with type A acute aortic dissection. *Am. J. Emerg. Med.* 36 (5), 741–744. doi:10.1016/j.ajem.2017.10.001

Mandell, I. M., Bynum, F., Marshall, L., Bart, R., Gold, J. I., and Rubin, S. (2015). Pediatric Early Warning Score and unplanned readmission to the pediatric intensive care unit. J. Crit. Care 30 (5), 1090–1095. doi:10.1016/j.jcrc.2015.06.019 McCartan, T. A., Worrall, A. P., Conluain, R. Ó., Alaya, F., Mulvey, C., MacHale, E., et al. (2021). The effectiveness of continuous respiratory rate monitoring in predicting hypoxic and pyrexic events: A retrospective cohort study. *Physiol. Meas.* 42 (6), 065005. Published 2021 Jun 29. doi:10.1088/1361-6579/ac05d5

Mjelva, Ø. R., Svingen, G. F. T., Pedersen, E. K. R., Seifert, R., Kvaløy, J. T., Midttun, Ø., et al. (2018). Fibrinogen and neopterin is associated with future myocardial infarction and total mortality in patients with stable coronary artery disease. *Thromb. Haemost.* 118 (4), 778–790. doi:10.1055/s-0038-1629912

Myint, P. K., Trepte, N. J., Parker, R. A., Sultanzadeh, S. J., Patel, J., Mallinson, R., et al. (2011). Vital signs as determinants of immediate and longer term mortality outcome of patients admitted from nursing homes. *Aging Clin. Exp. Res.* 23 (4), 309–315. doi:10. 1007/BF03337756

Odetola, F. O., Clark, S. J., Dechert, R. E., and Shanley, T. P. (2007). Going back for more: An evaluation of clinical outcomes and characteristics of readmissions to a pediatric intensive care unit. *Pediatr. Crit. Care Med.* 8 (4), 343–347. doi:10.1097/01. PCC.0000269400.67463.AC

Penny, D. J., and Vick, G. W. (2011). Ventricular septal defect. Lancet 377 (9771), 1103-1112. doi:10.1016/S0140-6736(10)61339-6

Rivers, E., Nguyen, B., Havstad, S., Ressler, J., Muzzin, A., Knoblich, B., et al. (2001). Early goal-directed therapy in the treatment of severe sepsis and septic shock. *N. Engl. J. Med.* 345 (19), 1368–1377. doi:10.1056/NEJMoa010307

Strauß, R., Ewig, S., Richter, K., König, T., Heller, G., and Bauer, T. T. (2014). The prognostic significance of respiratory rate in patients with pneumonia: A retrospective analysis of data from 705,928 hospitalized patients in Germany from 2010-2012. *Dtsch. Arztebl Int.* 111 (29-30), 503–508. doi:10.3238/arztebl.2014.0503.v

Weinstein, M. P., Towns, M. L., Quartey, S. M., Mirrett, S., Reimer, L. G., Parmigiani, G., et al. (1997). The clinical significance of positive blood cultures in the 1990s: A prospective comprehensive evaluation of the microbiology, epidemiology, and outcome of bacteremia and fungemia in adults. *Clin. Infect. Dis.* 24 (4), 584–602. doi:10.1093/ clind/24.4.584

Wu, W. T., Li, Y. J., Feng, A. Z., Li, L., Huang, T., Xu, A. D., et al. (2021). Data mining in clinical big data: The frequently used databases, steps, and methodological models. *Mil. Med. Res.*;8(1):44. doi:10.1186/s40779-021-00338-z

Yang, J., Li, Y., Liu, Q., Feng, A., Wang, T., et al. (2020). Brief introduction of medical database and data mining technology in big data era. *J. Evid. Based Med.* 13 (1), 57–69. doi:10.1111/jebm.12373

Zhang, D. P., Mao, X. F., Wu, T. T., Chen, Y., Hou, X. G., Yang, Y., et al. (2020). The fibrinogen-to-albumin ratio is associated with outcomes in patients with coronary artery disease who underwent percutaneous coronary intervention. *Clin. Appl. Thromb. Hemost.* 26, 1076029620933008. doi:10.1177/1076029620933008