

# A pragmatic lab-based tool for risk assessment in cardiac critical care: data from the Critical Care Cardiology Trials Network (CCCTN) **Registry**

 ${\sf S}$ iddharth M. Patel  $\bm \Theta^{\sf 1}$ , Jacob C. Jentzer $^2$ , Carlos L. Alviar $^3$ , Vivian M. Baird-Zars $^1$ , Gregory W. Barsness<sup>2</sup>, David D. Berg<sup>1</sup>, Erin A. Bohula<sup>1</sup>, Lori B. Daniels<sup>4</sup>, Andrew P. DeFilippis<sup>5</sup>, Ellen C. Keeley<sup>6</sup>, Michael C. Kontos<sup>7</sup>, Patrick R. Lawler<sup>8</sup>, P. Elliott Miller<sup>9</sup>, Jeong-Gun Park **®** <sup>1</sup>, Robert O. Roswell<sup>10</sup>, Michael A. Solomon<sup>11</sup>, Sean van Diepen @  $^{\text{12,13}},$  Jason N. Katz $^{\text{14}},$  and David A. Morrow @  $^{\text{1}*}$ ; on behalf of the CCCTN Investigators

<sup>1</sup>Levine Cardiac Intensive Care Unit, TIMI Study Group, Cardiovascular Division, Department of Medicine, Brigham and Women's Hospital and Harvard Medical School, 60 Fenwood Rd, Suite 7022, Boston, MA 02115, USA; <sup>2</sup>Department of Cardiovascular Medicine, Mayo Clinic, Rochester, MN, USA; <sup>3</sup>Division of Cardiology, Department of Medicine, NYU Langone Medical Center, New York, NY, USA; <sup>4</sup>Sulpizio Cardiovascular Center, University of California San Diego, La Jolla, CA, USA; <sup>5</sup>Division of Cardiology Department of Medicine, Vanderbilt University, Nashville, TN, USA; <sup>6</sup>Division of Cardiology, Department of Medicine, University of Florida, Gainesville, FL, USA; <sup>7</sup>Division of Cardiology, Department of Internal Medicine, Virginia Commonwealth University, Richmond, VA, USA; <sup>8</sup>Peter Munk Cardiac Centre at Toronto General Hospital, Division o*f* Cardiology and Interdepartmental Division of Critical Care Medicine, University of Toronto, Toronto, Ontario, Canada; <sup>9</sup>Section of Cardiovascular Medicine, Department of Internal Medicine, Yale University School of Medicine, New Haven, CT, USA; <sup>10</sup>Lenox Hospital, Northwell Health, New York, NY, USA; <sup>11</sup>Critical Care Medicine Department, National Institutes of Health Clinical Center and Cardiovascular Branch, National Heart, Lung, and Blood Institute of the National Institutes of Health, Bethesda, MD, USA;  $12$ Division of Cardiology, Department of Medicine, University of Alberta, Edmonton, AB, Canada;  $13$ Division of Cardiology, Department of Critical Care, University of Alberta, Edmonton, AB, Canada; and 14Division of Cardiology, Department of Medicine, Duke University, Durham, NC, USA

Received 12 November 2021; revised 11 December 2021; editorial decision 23 January 2022; accepted 24 January 2022; online publish-ahead-of-print 3 February 2022



\* Corresponding author. Tel: +1 617 278 0181, Fax: +1 617 734 7329, Email: dmorrow@partners.org

Published on behalf of the European Society of Cardiology. All rights reserved. © The Author(s) 2022. For permissions, please email: journals.permissions@oup.com.

<span id="page-1-0"></span>

Key points

- Laboratory measures which are often obtained routinely at the time of patient admission to cardiac intensive care units (CICUs) are associated with risk of in-hospital mortality.
- Among these, lower pH, higher lactate, lower estimated glomerular filtration rate, and lower platelets are the top four independent lab predictors of in-hospital mortality across a broad range of CICU patients.
- A simple integer risk score comprised of only these four independent lab predictors identifies an over 20-fold gradient of risk for inhospital mortality, with consistently good performance and excellent calibration in an external validation cohort.

# Introduction

Patients admitted to cardiac intensive care units (CICUs) present with a wide variety of diagnoses and mortality risk.<sup>1–6</sup> While tools validated in general ICUs, such as the Sequential Organ Failure Assessment (SOFA) score, offer good discrimination in the CICU, they require serial data, rely on a large number of variables, and are sub-optimally calibrated. $6-10$  Emerging CICU-specific risk tools using data from the in-hospital course robustly discriminate mortality risk.<sup>11</sup> Much of this discriminatory capacity derives from laboratory data obtained at presentation. We hypothesized that a pragmatic approach using only limited laboratory data would offer acceptable performance for rapid initial risk stratification among CICU admissions. We sought to derive a simple integer risk score for mortality using routinely collected lab markers which could be applied at time of admission.

# **Methods**

#### Study population

The Critical Care Cardiology Trials Network (CCCTN) is a network of tertiary CICUs in North America coordinated by the TIMI Study Group (Boston, MA, USA). Methods for the CCCTN Registry are published.<sup>3</sup> This analysis encompassed three annual collection campaigns (2017– 2020) of all consecutive CICU admissions during each site's ( $n = 25$ ) collection period. The first and second campaigns were derivation and validation cohorts, respectively. The third campaign was used for a sensitivity analysis based on timing of pH ascertainment. External validation was performed in a previously reported CICU population (Mayo Clinic, 2007– 2015, Rochester, MN, USA).<sup>11</sup>

#### Statistical analysis

Candidate lab indicators were prospectively selected for clinical relevance: haemoglobin, creatinine, lactate, alanine aminotransferase, aspartate aminotransferase, bilirubin, glucose, platelets, and pH (venous or arterial). pH values collected in the initial two campaigns reflected the lowest or 'worst' values. In the third campaign, initial and 'worst' pH values were separately collected. Continuous variables were logtransformed where appropriate.

Multivariable logistic regression was performed using forward stepwise-selection with  $\alpha \leq 0.1$  for inclusion and  $\alpha \leq 0.05$  for selection to identify the top independent predictors in the derivation cohort. For ease of use, continuous variables were categorized based on clinically relevant pre-specified cut-offs. Points were allocated for each independent predictor with simple weighting guided by beta-coefficients. Labs not measured (or missing) were assigned 0 points. Discrimination was assessed using the C-index and contrasted with the SOFA score.<sup>10</sup> Calibration was assessed in the validation cohorts by comparing observed mortality rates by risk category with predicted rates in the derivation cohort (Pearson r). P-values were two-sided. Statistics were performed with SAS v9.4.

# **Results**

# Patient demographics and intensive care unit indications

A total of 4352 admissions comprised the derivation cohort ([Table 1](#page-2-0)). The most common primary diagnoses were acute coronary syndrome (28.1%) and heart failure (13.5%). Cardiac intensive care unit indications included shock in 26.0% and cardiac arrest in 10.0%.

#### Lab indicators

. .

All nine candidate lab indicators from the derivation cohort were associated with in-hospital mortality on univariable analysis; on multivariable assessment, lower pH, higher lactate, lower estimated glomerular filtration rate (eGFR), and lower platelets were the top four variables independently associated with in-hospital mortality [\(Supplementary material online,](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) Table S1 and [Figure S1](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data)). Categorical modelling of each risk indicator revealed a significant relationship

<span id="page-2-0"></span>



. .

All data are reported as column % (n), unless otherwise specified.

For rows that are limited to subgroups or have missing data, available n is specified.

BMI, body mass index; LVEF, left ventricular ejection fraction.

with mortality and guided point allocation for the CCCTN Lab-Based Risk Score ([Figure 1](#page-3-0); [Supplementary material online,](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) Table S2).

#### Lab-based risk score

In the derivation cohort, in-hospital mortality was 13.7%. The CCCTN Lab-Based Risk Score identified a strong gradient of risk for in-hospital mortality (3.2–73.5%, P< 0.001, [Figure 2A](#page-4-0)). This gradient of risk was maintained in higher-risk subgroups (shock and cardiac ar-rest; P < 0.001, [Figure 2B](#page-4-0)). The risk score demonstrated good discrimination [C-index 0.82, 95% confidence interval (CI) 0.80–0.84] in the overall cohort and higher-risk patient subgroups (cardiac arrest: Cindex 0.79, 95% CI 0.75–0.83; shock: C-index 0.72, 95% CI 0.69– 0.75). Continuous modelling of the lab indicators did not improve performance (C-index 0.82, 95% CI 0.80–0.84). Inclusion of all clinical factors from Table 1, including age and sex, also did not substantially improve discrimination (C-index 0.84, 95% CI 0.82–0.86). Performance was good across subgroups [\(Supplementary material](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) online, [Figures S2](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data)–S4).

Admission eGFR and platelets were available in >99% of patients indicating a high degree of completeness of data capture. Lactate and pH were measured selectively per the clinician's discretion and available in 59.3% and 49.9% of patients, respectively. Although discrimination was lower, performance remained acceptable when applied to only patients with data for all four labs (C-index 0.77, 95% CI 0.75– 0.80).

#### Validation and sensitivity analyses

Validation in the second campaign ( $n = 3888$ , [Supplementary material](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) online, [Table S3](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data)) demonstrated a similarly strong, graded relationship with in-hospital mortality ([Figure 2C](#page-4-0); C-index 0.82; 95% CI 0.80-0.84). Assessment of calibration revealed very good qualitative agreement between the predicted risk and the observed mortality rates in the

<span id="page-3-0"></span>

Figure I Adjusted odds ratios for in-hospital mortality in the derivation cohort by lab indicator and point allocation in the CCCTN Lab-Based Risk Score. Estimated glomerular filtration rate (eGFR) was calculated using serum creatinine in the Chronic Kidney Disease Epidemiology Collaboration (CKD-EPI) equations. Odds ratios displayed are in reference to pH >7.3, lactate <2 mmol/L, eGFR  $\geq$ 45 mL/min/1.73 m<sup>2</sup>, and platelets >100 K/µL, respectively. pH used in the analysis reflects the 'worst' pH value as was captured in the derivation cohort. ABG, arterial blood gas; CCCTN, Critical Care Cardiology Trials Network; eGFR, estimated glomerular filtration rate; VBG, venous blood gas.

validation set ( $r = 0.99$ , [Figure 2D](#page-4-0)). Discrimination was qualitatively similar to the SOFA score (C-index 0.85, 95% CI 0.83–0.87; [Supplementary material online,](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) Table S4). Subgroup analyses revealed good performance ([Supplementary material online,](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) Figure S4).

In the third campaign, initial pH was collected in 731 patients. Reassessment of the score using initial pH yielded overall good performance (C-index 0.79, 95% CI 0.77–0.81; r = 0.93; [Supplementary ma](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data)[terial online,](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) Figure S5 and [Table S4](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data)).

The external validation cohort included 9716 patients with characteristics and demographics that have been previously reported.<sup>11</sup> Stable discrimination and excellent calibration were apparent using initial pH (C-index 0.79, 95% CI 0.77-0.80;  $r = 0.99$ ; [Figure 2E](#page-4-0) and F; [Supplementary material online,](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) Table S4).

# **Discussion**

We developed a pragmatic lab-based risk score in a large wellcharacterized cohort of CICU patients. Using only four variables routinely available at the time of admission, the CCCTN Lab-Based Risk Score identified a 20-fold gradient in mortality risk with very good discrimination and calibration in two separate validation cohorts. Importantly, performance of the score was consistent across commonly encountered CICU diagnoses and in higher-risk subgroups (i.e., shock, cardiac arrest). This risk score can be easily calculated at the bedside or incorporated into electronic medical record systems. As such, the CCCTN Lab-Based Risk Score could serve to complement pre-existing, more complex risk assessment tools which are

effective but require serial measures or are applied after diagnostic testing for specific CICU subpopulations (e.g. SOFA, APACHE II, M-CARS, or IABP-SHOCK II scores).<sup>7-13</sup>

Early risk stratification using such a tool could serve to (i) guide triage and treatment decisions, (ii) optimize resource allocation, and (iii) inform initial goals of care discussions. $13$  From a research perspective, such a tool may be useful for guiding entry into clinical trials, controlling for differences in patient acuity, and facilitating riskadjusted quality assessment.<sup>14</sup> The novelty of this risk score is its effective risk stratification that might be easily applied on the 'first call' from the Emergency Department to provide an immediate gauge of mortality risk using only four lab-based variables.

#### Limitations

. .

First, our analysis cohort was predominantly from tertiary care centres; calibration may be different when applied to other environments. Second, the availability of lactate and pH integrates the initial clinical assessment of the ordering clinician and contributes to the discriminatory performance of the score. Third, worst pH was used in the first two campaigns. However, use of initial pH in the third campaign and a separate external validation cohort demonstrated consistent performance. Fourth, performance was contrasted only with the SOFA score and merits comparison with other risk models that were unable to be tested in our dataset. Fifth, cardiac biomarkers (e.g. troponin, brain natriuretic peptide) were not captured and may further complement risk prediction.

<span id="page-4-0"></span>

Figure 2 Performance of the CCCTN Lab-Based Risk Score with respect to in-hospital mortality. (A) CCCTN derivation cohort, (B) high-risk subgroups (derivation cohort), (C) CCCTN validation cohort, (D) correlation of mortality rates by CCCTN Lab-Based Risk Score in the CCCTN validation and derivation cohorts, (E) external validation cohort using initial pH, and (F) correlation of mortality rates by CCCTN Lab-Based Risk Score in the external validation cohort. CCCTN, Critical Care Cardiology Trials Network.

> .

# **Conclusion**

A simple integer risk score utilizing readily available lab indicators at time of CICU admission may accurately stratify in-hospital mortality risk.

# Supplementary material

[Supplementary material](https://academic.oup.com/ehjacc/article-lookup/doi/10.1093/ehjacc/zuac012#supplementary-data) is available at European Heart Journal: Acute Cardiovascular Care online.

## Acknowledgements

The authors take responsibility for all aspects of the reliability and freedom from bias of the data presented and their discussed interpretation.

### Funding

S.M.P. is supported by a T32 postdoctoral training grant from the National Heart, Lung and Blood Institute (T32HL007604-36). M.A.S. received research support from the National Institutes of Health Clinical

<span id="page-5-0"></span>. Center intramural research funds. The remaining authors received no financial support for the research, authorship, and/or publication of this article.

#### Conflict of interest: none declared.

#### **References**

- 1. Morrow DA, Fang JC, Fintel DJ, Granger CB, Katz JN, Kushner FG, Kuvin JT, Lopez-Sendon J, McAreavey D, Nallamothu B, Page RL, Parrillo JE, Peterson PN, Winkelman C; American Heart Association Council on Cardiopulmonary, Critical Care, Perioperative and Resuscitation, Council on Clinical Cardiology, Council on Cardiovascular Nursing, and Council on Quality of Care and Outcomes Research. Evolution of critical care cardiology: transformation of the cardiovascular intensive care unit and the emerging need for new medical staffing and training models: a scientific statement from the American Heart Association. Circulation 2012;126:1408–1428.
- 2. Sinha SS, Sioding MW, Sukul D, Prescott HC, Iwashyna TI, Gurm HS, Cooke CR, Nallamothu BK. Changes in primary noncardiac diagnoses over time among elderly cardiac intensive care unit patients in the United States. Circ Cardiovasc Qual Outcomes 2017;10:e003616.
- [3.](#page-1-0) Bohula EA, Katz JN, van Diepen S, Alviar CL, Baird-Zars VM, Park J-G, Barnett CF, Bhattal G, Barsness GW, Burke JA, Cremer PC, Cruz J, Daniels LB, DeFilippis A, Granger CB, Hollenberg S, Horowitz JM, Keller N, Kontos MC, Lawler PR, Menon V, Metkus TS, Ng J, Orgel R, Overgaard CB, Phreaner N, Roswell RO, Schulman SP, Snell RJ, Solomon MA, Ternus B, Tymchak W, Vikram F, Morrow DA; Critical Care Cardiology Trials Network. Demographics, care patterns, and outcomes of patients admitted to cardiac intensive care units: the Critical Care Cardiology Trials Network Prospective North American Multicenter Registry of Cardiac Critical Illness. JAMA Cardiol 2019;4:928–935.
- 4. Watson RA, Bohula EA, Gilliland TC, Sanchez PA, Berg DD, Morrow DA. Prospective registry of cardiac critical illness in a modern tertiary care Cardiac Intensive Care Unit. Eur Heart J Acute Cardiovasc Care 2019;8:755-761.
- 5. Katz JN, Shah BR, Volz EM, Horton JR, Shaw LK, Newby LK, Granger CB, Mark DB, Califf RM, Becker RC. Evolution of the coronary care unit: clinical

. characteristics and temporal trends in healthcare delivery and outcomes. Crit Care Med 2010;38:375-381.

- 6. Jentzer JC, van Diepen S, Murphree DH, Ismail AS, Keegan MT, Morrow DA, Barsness GW, Anavekar NS. Admission diagnosis and mortality risk prediction in a contemporary cardiac intensive care unit population. Am Heart I 2020:224: 57–64.
- 7. Argyriou G, Vrettou CS, Filippatos G, Sainis G, Nanas S, Routsi C. Comparative evaluation of Acute Physiology and Chronic Health Evaluation II and Sequential Organ Failure Assessment scoring systems in patients admitted to the cardiac intensive care unit. J Crit Care 2015;30:752–757.
- 8. Bennett CE, Wright RS, Jentzer J, Gajic O, Murphree DH, Murphy JG, Mankad SV, Wiley BM, Bell MR, Barsness GW. Severity of illness assessment with application of the APACHE IV predicted mortality and outcome trends analysis in an academic cardiac intensive care unit. *I Crit Care* 2019:50: 242–246.
- 9. Jentzer JC, Bennett C, Wiley BM, Murphree DH, Keegan MT, Gajic O, Wright RS, Barsness GW. Predictive value of the sequential organ failure assessment score for mortality in a contemporary cardiac intensive care unit population. / Am Heart Assoc 2018;7:e008169.
- [10](#page-1-0). Vincent JL, Moreno R, Takala J, Willatts S, De Mendonça A, Bruining H, Reinhart CK, Suter PM, Thijs LG. The SOFA (Sepsis-related Organ Failure Assessment) score to describe organ dysfunction/failure. Intensive Care Med 1996;22:707–710.
- [11](#page-1-0). Jentzer JC, Anavekar NS, Bennett C, Murphree DH, Keegan MT, Wiley B, Morrow DA, Murphy JG, Bell MR, Barsness GW. Derivation and validation of a novel cardiac intensive care unit admission risk score for mortality. J Am Heart Assoc 2019;8:e013675.
- 12. Pöss J, Köster J, Fuernau G, Eitel I, de Waha S, Ouarrak T, Lassus J, Harjola V-P, Zeymer U, Thiele H, Desch S. Risk stratification for patients in cardiogenic shock after acute myocardial infarction. J Am Coll Cardiol 2017;69:1913–1920.
- [13](#page-3-0). Miller PE, Jentzer J, Katz JN. Are unselected risk scores in the cardiac intensive care unit needed? J Am Heart Assoc 2021;10:e021940.
- [14](#page-3-0). Morrow DA, Antman EM, Murphy SA, Assmann SF, Giugliano RP, Cannon CP, Gibson CM, McCabe CH, Barron HV, Van de Werf F, Braunwald E. The Risk Score Profile: a novel approach to characterising the risk of populations enrolled in clinical studies. Eur Heart J 2004;25:1139-1145.