Inflammatory biomarkers as independent prognosticators of 28-day mortality for COVID-19 patients admitted to general medicine or ICU wards: a retrospective cohort study

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Abstract

Inflammatory biomarkers may be associated with disease severity and increased mortality in COVID-19 patients but have not been studied in North American populations. We sought to determine whether a set of commonly ordered inflammatory biomarkers can predict 28-day mortality. We analyzed a multi-centered (four) COVID-19 registry cohort from March 4th to December 7th, 2020. This cohort included COVID-19-positive patients admitted to medical wards or intensive care units. Patients presenting to the emergency department for COVID-19 symptoms and then subsequently discharged were also included. We performed Cox-regression analysis to measure whether commonly used biomarkers were associated with an increased 28-day mortality. Of 336 COVID-19-positive patients, 267 required hospital admission, and 69 were seen in the emergency room and discharged. The median age was 63 years (IQR 80–50) and the female-to-male ratio was 49:51. Derivation of internally validated cut-offs suggested that C-reactive protein ≥ 78.4 mg/L, neutrophil-to-lymphocyte ratio ≥ 6.1, lymphocyte-to-white blood cell ratio < 0.127, and a modified Glasgow prognostic score equal to 2 vs. 1 or 0 were associated with the highest increased risk of 28-day mortality. We provide early estimates of cut-off values for inflammatory biomarkers and indices measured at the time of admission that may be useful to clinicians for predicting 28-day mortality in North American COVID-19 patients.

Keywords Biomarkers · COVID-19 · NLR · CRP · mGPS

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Introduction

COVID-19 is a respiratory disease caused by the SARS-CoV-2 infection. COVID-19 pathogenesis is associated with upregulation of inflammatory pathways and associated inflammatory biomarkers, which underlie the biological plausibility for their use clinically. For example, severe cases of COVID-19 are characterized by excessive expression of pro-inflammatory mediators with a relative paucity of anti-inflammatory mediators [1]. Infection causes the respiratory epithelial cells to release interleukins (IL) such as IL-6 [2]. IL-6 upregulates hepatic C-reactive protein (CRP) production and other acute phase reactants such as ferritin. COVID-19-infected individuals may have higher serum IL-6 levels due to an abnormal population of CD14+CD16+ inflammatory monocytes. In addition to these pro-inflammatory mediators, severe COVID-19 cases present with lymphopenia, particularly of CD4+ T-cells, which includes regulatory T-cells important in the modulation of pro-inflammatory immune responses [3]. Therefore, both CRP and lymphopenia have been consistently referenced biomarkers for disease severity in COVID-19 patients. Other inflammatory biomarkers share similar biological plausibility. Low serum albumin for example has been postulated to be associated with increased mortality risk in COVID-19 patients due to its antioxidant and anticoagulant properties [4]. For this reason, the modified Glasgow Prognostic score (mGPS) is an interesting choice for investigation, as it is a composite score of CRP and albumin. The hypothesis being that if each of these markers were associated with increased mortality in COVID-19, the composite correction for each may provide a better estimate of its association with COVID-19 infection. Furthermore, ratios and composite scores of the complete blood count (CBC) such as neutrophil-to-lymphocyte (NLR) count have been hypothesized and studied in COVID-19. There is significant potential for these ratios, including previously understudied ones such as lymphocyte-to-white blood cell ratio, platelet-to-white blood cell ratio, and platelet-to-lymphocyte ratio.

Inflammatory biomarkers are used for diagnosis, prognosis, and theragnosis, have been proposed as a useful metric for prognosticating the disease course of patients with COVID-19 [4, 5]. Studies have examined the prognostic use of inflammatory biomarkers in prognosticating disease severity, 28-day mortality, and progression to Acute Respiratory Distress Syndrome (ARDS) [6–8]. A recent meta-analysis of 16 studies conducted in China, found consistent evidence of a positive association between inflammatory biomarkers and COVID-19 severity, however, noted the need for further research to clarify whether these results would be consistent across different countries and populations [5]. In addition, rigorously validated biomarker cutoffs, that could potentially be used by clinicians to risk stratify patients with COVID-19, have not been established.

Our objective was to examine the association between commonly ordered inflammatory biomarkers on admission, and 28-day mortality in a North American sample of COVID-19 patients, and to determine the optimal cut-off values on the relevant biomarkers for predicting mortality. We hypothesized that commonly ordered biomarkers and indices such as CRP and NLR would be useful to clinicians for estimating the risk of 28-day mortality.

Methods

Study design and population

We conducted an analysis of a registry cohort study of patients in three community hospital and one academic centre from the McMaster Multi-Regional Hospital Coronavirus Registry (COREG) between March 4th and December 7th 1st 2020. COREG is a multicentered data registry collecting information on positive SARS-CoV-2 PCR cases in Ontario, Canada. The region encompasses a catchment of approximately 1,000,000 persons, 3 community and 1 academic hospital. Our patient cohort includes emergency department (ED) visits and admissions into hospital including the intensive care units (ICU). Our study received ethics approval from the Hamilton Integrated Research Ethics Board (11,169-C) and Tri-Hospital Research Ethics Board (GRH RC 2020-166).
Study population

We selected all patients with a positive PCR for COVID-19 at three community centres and one academic centre, with records between March 4th, 2020 and December 7th, 2020, inclusive of all known hospitalized cases up to the time of analysis ($n = 350$). We excluded those with no known outcome at time of analysis ($n = 4$) and those without any of the target blood work ($n = 10$), for a final sample size of 336 (Fig. 1).

Biomarkers

Our study examined several commonly ordered admission biomarkers: white blood cells (WBC), platelets, neutrophils, and lymphocytes, and ratios of these counts, and indices including NLR, platelet-to-lymphocyte ratio (PLR), lymphocyte–WBC ratio, and platelet–WBC ratio. These ratios were chosen to mirror previous research, as well as add to the potential list of inflammatory indices available for clinicians. We also included albumin, C-reactive protein (CRP), lactate dehydrogenase (LDH), ferritin, D-dimer, and creatine kinase (CK). We also calculated the modified Glasgow prognostic score, which is an ordinal measure combining albumin and CRP that has been validated in prognosticating various cancers [9]. It is calculated by attributing one point for each of CRP ≥ 10 mg/L and Albumin < 35 g/L. The range of the scale for mGPS is from 0 to 2, with 2 portending the worst prognosis.

Statistical methods

Means (standard deviation), median (interquartile range) and frequency (percentages) statistics were used to describe the study population characteristics, and biomarkers stratified by visit type. We classified patients based on their visit type: patients admitted to the emergency department with COVID-19 symptoms, but were later stable enough for discharge, admission to a medicine unit, and admission to an intensive care unit (ICU). A patient who was admitted to an ICU at any time during their hospital stay was classified in the ICU cohort. As many of the biomarkers have skewed distributions, we performed a logarithm transformation of all biomarkers except for the mGPS. We also standardized each biomarker (except for mGPS) to have a mean of zero and standard deviation of one to allow for comparison of hazard ratios across biomarkers. The primary outcome of this study was time to 28-day mortality.

We examined the association between each biomarker with mortality using Cox regression. Each biomarker was examined as a continuous measure both univariately and adjusted for age and sex. In addition to HRs and 95% confidence intervals, we reported the c-index for the Cox models as a measure of predictive discriminability [10]. We assessed the proportional hazards assumptions with a Schoenfeld residual test and by inspecting survival curves. A sensitivity analysis on the associations between biomarkers and hospital and ICU admission (disease severity) was conducted with binary logistic regression and can be found in supplementary Table 1.

Biomarkers that were significantly associated with morality in the adjusted analysis were further examined using cutoffs on the original scale of the biomarker. A priori cutoffs were selected based on previous literature [8, 11–19], primarily COVID-19 biomarker studies. For biomarkers with no previous COVID-19 studies, cutoffs were used from other areas of research, such as cancer and heart disease [11, 12]. We also derived a novel cutoff for each biomarker, which aimed to maximize the predictive discriminability of the biomarker (c-index) for mortality. This cutoff was selected by splitting each biomarker into deciles and using tenfold cross-validation to select the decile cutoff with the highest c-index. We generated Kaplan–Meier curves on the original data for each of the derived cutoffs.

Results

Patient characteristics

Of the 336 patients included in the study, 69 (20.5%) visited the emergency department and were subsequently discharged, 150 (60.7%) were admitted to the medicine wards, and 63 (18.7%) were admitted to ICU (Table 1). The median age was 61 years old (IQR 50–80) and the female (166, 49%) to male (170, 51%) ratio was balanced. However, there were disproportionately more males admitted to the ICU than females [40 (63%) vs. 23 (37%)]. Symptoms of fever ($n = 199, 59%$) and cough ($n = 214, 64%$) were the most common presenting complaint for all admission types. Most patients ($n = 235, 70.16%$) had at least one comorbidity, and hypertension was the most prevalent comorbidity ($n = 159, 47%$). The median time to presentation from onset of symptoms was approximately 5 days (IQR = 10–2).

Biomarkers

There were significant differences between visit types in the median levels for all the biomarkers examined (Table 1). For example, the median admission NLR for patients discharged from the emergency department was 2.6 (IQR 1.9–4.2) compared to ICU 7.2 (IQR 5.0–11.2). Similarly, median CRP was 4.0 mg/L (IQR 2.0–16.0) versus 126.0 mg/L (IQR 53.0–195.0), respectively.
Missing biomarker data were handled with multivariate imputation by chained equations using 20 imputations [20]. Our analysis of the missing data found that missingness was most strongly related to disease severity, which can be largely accounted for by analyzing the outcomes and other biomarker data collected (e.g., patients admitted to ICU were more likely to have the target biomarkers ordered than those admitted to a medicine unit or emergency department only.

### Table 1 Stratifying characteristics by visit types

| Characteristic | Overall, N = 336<sup>a</sup> | ED only, N = 69<sup>a</sup> | ICU, N = 63<sup>a</sup> | Ward, N = 204<sup>a</sup> | p value<sup>b</sup> |
|---------------|-------------------------------|---------------------------|-------------------------|---------------------------|-------------------|
| Sex at birth  |                               |                           |                         |                           | 0.074             |
| Female        | 166 (49%)                     | 37 (54%)                  | 23 (37%)                | 106 (52%)                 |                   |
| Male          | 170 (51%)                     | 32 (46%)                  | 40 (63%)                | 98 (48%)                  |                   |
| Age           | 63 (50, 80)                   | 47 (33, 54)               | 57 (51, 70)             | 72 (58, 85)               | < 0.001           |
| Vitals and symptoms |               |                           |                         |                           |                   |
| Temperature   | 21 37.10 (36.70, 38.10)       | 37.10 (36.88, 37.50)      | 37.75 (36.90, 38.90)    | 37.10 (36.65, 38.10)      | 0.030             |
| Oxygen saturation | 18 95.0 (93.0, 97.0)        | 97.0 (96.0, 99.0)         | 93.0 (90.0, 95.0)       | 95.0 (93.0, 97.0)         | < 0.001           |
| Fever         | 0 199 (59%)                   | 44 (64%)                  | 49 (78%)                | 106 (52%)                 | < 0.001           |
| Shortness of breath | 0 191 (57%)                   | 34 (49%)                  | 57 (90%)                | 100 (49%)                 | < 0.001           |
| Cough         | 0 214 (64%)                   | 48 (70%)                  | 49 (78%)                | 117 (57%)                 | 0.007             |
| Fatigue/Malaise | 0 143 (43%)                 | 30 (43%)                  | 30 (48%)                | 83 (41%)                  | 0.6               |
| Wheezing      | 0 14 (4.2%)                   | 2 (2.9%)                  | 4 (6.3%)                | 8 (3.9%)                  | 0.6               |
| Muscle aches (Myalgia) | 0 57 (17%)               | 15 (22%)                  | 19 (30%)                | 23 (11%)                  | 0.001             |
| Diarrhea      | 0 55 (16%)                    | 6 (8.7%)                  | 13 (21%)                | 36 (18%)                  | 0.13              |
| Comorbidities |                               |                           |                         |                           |                   |
| Hypertension  | 0 159 (47%)                   | 12 (17%)                  | 31 (49%)                | 116 (57%)                 | < 0.001           |
| Coronary artery disease | 0 45 (13%)                 | 5 (7.2%)                  | 7 (11%)                 | 33 (16%)                  | 0.14              |
| Dementia      | 0 48 (14%)                    | 1 (1.4%)                  | 0 (0%)                  | 47 (23%)                  | < 0.001           |
| COPD          | 0 66 (20%)                    | 9 (13%)                   | 11 (17%)                | 46 (23%)                  | 0.2               |
| Chronic kidney disease | 0 35 (10%)               | 0 (0%)                    | 6 (9.5%)                | 29 (14%)                  | 0.004             |
| Diabetes mellitus | 0 90 (27%)               | 7 (10%)                   | 17 (27%)                | 66 (32%)                  | 0.002             |
| Biomarkers    |                               |                           |                         |                           |                   |
| D-dimer       | 214 0.8 (0.3, 1.6)            | 0.2 (0.2, 0.2)            | 1.0 (0.7, 3.0)          | 0.8 (0.4, 1.6)            | < 0.001           |
| White blood cells | 26 6.6 (5.1, 9.5)            | 6.1 (4.8, 7.5)            | 7.7 (5.8, 11.3)         | 6.6 (5.1, 9.6)            | 0.003             |
| Neutrophils   | 26 4.8 (3.5, 7.2)             | 3.9 (3.0, 5.2)            | 6.1 (4.3, 9.2)          | 4.8 (3.4, 7.6)            | < 0.001           |
| CRP           | 129 66 (16, 137)              | 3 (2, 16)                 | 126 (53, 195)           | 66 (18, 124)              | < 0.001           |
| Platelets     | 28 212 (167, 279)             | 198 (164, 252)            | 191 (158, 252)          | 228 (174, 287)            | 0.047             |
| Ferritin      | 219 456 (150, 978)            | 95 (39, 546)              | 808 (405, 1,355)        | 333 (90, 605)             | < 0.001           |
| Creatinine kinase | 171 90 (46, 220)            | 52 (43, 71)               | 132 (59, 351)           | 94 (43, 209)              | 0.002             |
| Lymphocyte    | 26 1.00 (0.70, 1.50)          | 1.30 (1.05, 1.70)         | 0.80 (0.60, 1.20)       | 1.00 (0.60, 1.50)         | < 0.001           |
| LDH           | 155 346 (214, 535)            | 370 (186, 490)            | 373 (281, 661)          | 308 (206, 518)            | 0.032             |
| NLR           | 26 5.0 (2.8, 8.7)             | 2.6 (1.9, 4.2)            | 7.2 (5.0, 11.2)         | 5.0 (3.0, 8.8)            | < 0.001           |
| PLR           | 28 219 (147, 310)             | 151 (118, 210)            | 228 (175, 301)          | 231 (159, 334)            | < 0.001           |
| P/WR          | 28 31 (23, 42)                | 33 (27, 44)               | 27 (20, 34)             | 33 (24, 46)               | 0.002             |
| L/WR          | 26 0.15 (0.09, 0.23)          | 0.25 (0.18, 0.31)         | 0.12 (0.08, 0.15)       | 0.15 (0.09, 0.22)         | < 0.001           |
| Albumin       | 157 35.0 (32.0, 39.0)         | 40.0 (37.5, 40.0)         | 33.0 (30.0, 37.0)       | 35.0 (32.0, 39.0)         | < 0.001           |
| mGPS          | 194                            |                           |                         |                           | < 0.001           |
| 0             | 16 (11%)                      | 9 (64%)                   | 3 (6.2%)                | 4 (5.0%)                  |                   |
| 1             | 60 (42%)                      | 5 (36%)                   | 17 (35%)                | 38 (48%)                  |                   |
| 2             | 66 (46%)                      | 0 (0%)                    | 28 (58%)                | 38 (48%)                  |                   |
| Time since symptom onset | 10 5 (2, 10)                | 6 (3, 11)                 | 7 (3, 10)               | 5 (1, 9)                  | 0.062             |
| Died          | 0 51 (15%)                    | 0 (0%)                    | 12 (19%)                | 39 (19%)                  | < 0.001           |

<sup>a</sup>Statistics presented: n (%); Median (IQR)

<sup>b</sup>Statistical tests performed: Chi-square test of independence; Kruskal–Wallis test; Fisher’s exact test
visits). We performed an independent audit of the COREG study database to ensure no missing data were due to collection error. All analyses were done using R 4.0 [21].

### 28-day mortality

In this cohort, there were 51 (15.0%) deaths, 39 (19%) of which were patients on the wards, 12 (19%) in ICU and no deaths were recorded in patients who were discharged from the emergency department. The relative high mortality rate in our ward patients is due to the high numbers of long-term care patients and elderly patients with whom ICU care was not within their goals of care. Seven of the 14 biomarkers examined demonstrated a significant association with higher risk of 28-day mortality adjusted for age and sex. The largest effect size for a one standard deviation increase was observed for log-CRP with an adjusted HR of 2.43 (95% CI 1.44, 4.07) (Table 2). Other significant but smaller effect sizes were found for log-WBC aHR 1.77 (95% CI 1.32, 2.37); log-NLR 1.76 (95% CI 1.37, 2.26); log-lymphocyte count 0.77 (95% CI 0.59, 0.99); log-platelet–WBC ratio 0.56 (95% CI 0.45, 0.69); and log-lymphocyte–WBC ratio 0.60 (95% CI 0.48, 0.76). A one-unit increase in mGPS was also significantly associated with 28 day with an aHR of 2.94 (95% CI 1.41, 6.13).

The Schoenfeld-residual test indicated a potential violation of the proportional hazard’s assumption for CRP. Inspection of survival curves showed overlapping survival curves for the first 5 days, which we did not consider a significant enough violation to warrant changing our analytic approach (Fig. 2).

### Inflammatory biomarker cutoffs for 28-day mortality

The newly derived cutoffs which maximized the predictive discriminability of each biomarker for in hospital mortality were as follows: WBC ≥ 3.75 × 10^9/L, NLR ≥ 6.1, CRP ≥ 78.4 mg/L, Lymphocytes ≤ 0.8 × 10^9/L, P–WBC ratio ≥ 31.0, L–WBC ratio ≤ 0.127. All a priori and derived cutoffs were significantly associated with increased 28-day mortality. In the unadjusted analysis, the cutoff with the largest effect size was mGPS ≥ 2 (HR, 4.88 (95% CI 2.17, 10.99) followed by L–WBC ratio ≥ 0.228 (0.25 (95% CI 0.13, 0.5) (Table 3). In adjusted analysis, the cutoff with the largest effect size was L–WBC ratio ≥ 0.228 (aHR 0.12 (0.03, 0.53), followed by NLR ratio ≥ 3 (aHR, 5.69 (1.7, 19.03). Kaplan–Meier curves were generated for the derived cutoffs and for the mGPS (Fig. 2).

### Sensitivity analysis

Analysis of the outcomes of hospital and ICU admission produced generally similar results to the main analysis (supplemental table). A notable difference was that CK and d-dimer (positively) and albumin (negatively) were

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**Table 2** Cox regression analysis for both unadjusted and adjusted 28-day mortality by biomarker

| Biomarker | HR^a (95% CI) | p value | c-index | aHR^b (95% CI) | p value | c-index |
|-----------|---------------|---------|---------|----------------|---------|---------|
| CBC       |               |         |         |                |         |         |
| WBC       | 1.79 (1.37, 2.35) | <0.001 | 0.66   | 1.77 (1.32, 2.37) | <0.001 | 0.82   |
| Platelets | 0.90 (0.68, 1.20) | 0.480  | 0.51   | 0.83 (0.63, 1.11) | 0.204  | 0.80   |
| Lymphocytes | 0.68 (0.52, 0.90) | 0.007  | 0.61   | 0.77 (0.59, 0.99) | 0.044  | 0.81   |
| CBC ratios |            |         |         |                |         |         |
| NLR       | 2.07 (1.58, 2.71) | <0.001 | 0.71   | 1.76 (1.37, 2.26) | <0.001 | 0.85   |
| PLR       | 1.34 (1.03, 1.76) | 0.032  | 0.59   | 1.16 (0.90, 1.50) | 0.255  | 0.80   |
| P/WBC     | 0.57 (0.46, 0.71) | <0.001 | 0.66   | 0.56 (0.45, 0.69) | <0.001 | 0.84   |
| L/WBC     | 0.51 (0.40, 0.66) | <0.001 | 0.72   | 0.60 (0.48, 0.76) | <0.001 | 0.85   |
| Acute phase reactants | |         |         |                |         |         |
| CRP       | 2.42 (1.46, 4.01) | 0.001  | 0.68   | 2.43 (1.44, 4.07) | 0.001  | 0.84   |
| CK        | 1.54 (1.01, 2.35) | 0.045  | 0.62   | 1.37 (0.90, 2.10) | 0.134  | 0.81   |
| Ferritin  | 1.41 (0.82, 2.40) | 0.192  | 0.59   | 1.48 (0.92, 2.37) | 0.103  | 0.81   |
| d-dimer   | 1.42 (1.02, 1.96) | 0.040  | 0.63   | 1.20 (0.82, 1.76) | 0.336  | 0.80   |
| LDH       | 1.06 (0.70, 1.59) | 0.787  | 0.51   | 1.16 (0.78, 1.73) | 0.453  | 0.80   |
| Albumin   | 0.49 (0.37, 0.66) | <0.001 | 0.71   | 0.63 (0.46, 0.84) | 0.003  | 0.82   |
| Composite scores | |         |         |                |         |         |
| mGPSb     | 4.02 (1.98, 8.16) | <0.001 | 0.70   | 2.94 (1.41, 6.13) | 0.005  | 0.83   |

^aFor each increase in one standard deviation of the natural logarithm of the biomarker

^b mGPS is an ordinal measure and was, therefore, not transformed. Hazard ratios represent change in risk for each increase of one unit of the modified Glasgow Prognostic Score
associated with ICU admission but were not associated with 28-day mortality.

**Interpretation**

We found that commonly available clinical inflammatory biomarkers were significantly associated with 28-day mortality. Specifically, C-reactive protein, neutrophil-to-lymphocyte ratio, and lymphocyte-to-white blood count ratio showed the most consistent and strong associations with mortality in our study. The modified Glasgow prognostic score also demonstrated a consistently strong association with elevated risk for mortality. Our findings align with previous published biomarker studies, albeit in populations outside of North America. Our work further extends this prior work by providing newly derived optimal cutoff values for predicting mortality for the associated biomarkers. Furthermore, we examined new biomarkers (L/WBC ratio and mGPS) which had not been previously studied in COVID-19 and found they were significantly associated with mortality.

**NLR, CRP, and L/W ratio as predictors of 28-day mortality**

To our knowledge, there have been no prior studies examining the association between blood and clinical biomarkers with clinical outcomes in COVID-19 in populations outside of China. A recent meta-analysis which included 16 Chinese studies found an association between several important blood biomarkers and their association with disease severity in COVID-19 hospitalized patients. [5] These included NLR, which was associated with increased odds of developing COVID-19 pneumonitis and worse disease severity [14, 15, 22, 23]. Similarly, CRP above values ≥ 41.8 mg/L, were shown to significantly correlate with worse disease severity, particularly on radiological CT chest findings. [8, 9] There

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**Fig. 2** Kaplan–Meier curves for NLR, CRP, mGPS and L/WBC ratio with non-imputed data
is also an overall paucity of studies examining the predictors of in hospital mortality. Only CRP has been investigated for its association with COVID-19-related mortality and so far, the results have been mixed and inconclusive [18, 19]. To date, L/WBC ratio has not been examined in COVID-19, even though it has predictive implications in distinguishing between bacterial tonsillitis and glandular fever. [24] The biologic plausibility for CRP has been previously discussed. NLR may be associated with fundamental pathogenesis as well. Elevated neutrophils are associated with elevated secretion of IL-6 and TNF-alpha, causing a general inflammatory responses syndrome that may lead to underlying acute respiratory distress syndrome when present systematically. The ratio of NLR encompasses both this general inflammatory state and the underlying disease severity represented by lymphopenia. Similarly, L/WBC may represent the inverse of NLR, whereas the lymphopenia represents underlying disease severity from depleted lymphocytes to the general WBC count, which is mostly made up of neutrophils.

Largely, our results are consistent with biomarker studies in COVID-19 from China, but notably our data are from a North American cohort and focuses on 28-day mortality as the primary outcome. Our findings indicate that NLR is a useful biomarker for predicting elevated risk of mortality among patients with COVID-19 within the ward and ICU settings. In particular, at a cut off of greater than or equal to 6.1 and, there was an absolute increase in 28 day in hospital mortality of 19.9% (28.5% vs. 8.6%). This is an easily accessible biomarker available as a part of routine laboratory investigations ordered by clinicians, which can be used to risk stratify COVID-19 patients admitted to the ward or ICU. We also found CRP was a strong predictor, and admission to hospital and ICU. At our derived cutoff of 78.4 mg/L, CRP was associated with an absolute increase in mortality of 15.6% (24.5% vs 8.9%). However, unlike NLR, CRP may not be routinely performed in all admitting patients and, therefore, is likely not as easily accessible. Lastly, we are the first to investigate the role of L/WBC ratio as a biomarker in COVID-19 patients. L/WBC ratio had the largest effect with an absolute decrease in mortality of 17.2.4% (7.0% vs. 26.3%) with a cut off at 0.127. Like NLR, L/WBC is readily available as a part of the complete blood count. Of note, our reported optimal cut-off values are consistent with studies

| Biomarker | HR (95% CI) | p value | c-index | aHR (95% CI) | p value | c-index |
|-----------|------------|---------|---------|-------------|---------|---------|
| WBC       |            |         |         |             |         |         |
| ≥ 11.1×10^9/L | 3.14 (1.71, 5.75) | <0.001 | 0.60    | 2.54 (1.39, 4.65) | 0.003 | 0.82    |
| ≥ 7.5×10^9/L | 3.39 (1.86, 6.18) | <0.001 | 0.65    | 2.90 (1.59, 5.29) | 0.008 | 0.83    |
| NLR       |            |         |         |             |         |         |
| ≥ 3       | 5.65 (1.71, 18.73) | 0.005 | 0.60    | 5.69 (1.7, 19.03) | 0.006 | 0.82    |
| ≥ 5       | 3.25 (1.65, 6.37) | <0.001 | 0.63    | 3.52 (1.8, 6.87) | <0.001 | 0.84    |
| ≥ 6.1a    | 3.66 (1.97, 6.78) | <0.001 | 0.66    | 3.15 (1.71, 5.82) | <0.001 | 0.83    |
| CRP       |            |         |         |             |         |         |
| ≥ 41.8 mg/L | 3.16 (1.37, 7.29) | 0.009 | 0.62    | 3.15 (1.36, 7.32) | 0.009 | 0.82    |
| ≥ 78.4 mg/La | 3.08 (1.47, 6.45) | 0.004 | 0.63    | 2.75 (1.34, 5.66) | 0.008 | 0.82    |
| Lymphocytes|            |         |         |             |         |         |
| ≥ 0.8×10^9/L | 0.43 (0.24, 0.77) | 0.005 | 0.60    | 0.43 (0.24, 0.78) | 0.006 | 0.82    |
| ≥ 1.0×10^9/L | 0.46 (0.25, 0.83) | 0.011 | 0.59    | 0.45 (0.24, 0.82) | 0.011 | 0.82    |
| ≥ 0.8×10^9/Lb | 0.43 (0.24, 0.77) | 0.005 | 0.60    | 0.43 (0.24, 0.78) | 0.006 | 0.82    |
| mGPS      |            |         |         |             |         |         |
| ≥ 2       | 4.88 (2.17, 10.99) | <0.001 | 0.69    | 3.33 (1.45, 7.66) | 0.006 | 0.83    |
| P/WBC     |            |         |         |             |         |         |
| ≥ 20      | 0.31 (0.17, 0.57) | <0.001 | 0.60    | 0.29 (0.16, 0.53) | <0.001 | 0.82    |
| ≥ 30      | 0.47 (0.26, 0.85) | 0.014 | 0.59    | 0.40 (0.22, 0.72) | 0.003 | 0.82    |
| ≥ 31c     | 0.32 (0.17, 0.62) | 0.001 | 0.63    | 0.30 (0.16, 0.58) | <0.001 | 0.83    |
| L/WBC     |            |         |         |             |         |         |
| ≥ 0.228   | 0.12 (0.03, 0.5) | 0.005 | 0.61    | 0.12 (0.03, 0.53) | 0.006 | 0.83    |
| ≥ 0.127m  | 0.25 (0.13, 0.5) | <0.001 | 0.65    | 0.26 (0.13, 0.51) | <0.001 | 0.84    |

*a*Indicates cut off of 60th percentile

*b*Indicates cut off of 40th percentile

*c*Indicates cut off of 50th percentile

*ROC cross-validated cut offs are designated by percentile and all else are literature-driven cut offs*
in other disease populations, increasing the external validity of our results.

**Modified Glasgow prognostic score**

The mGPS is a composite of two common inflammatory biomarkers: CRP and albumin, both of which have individually been investigated in this analysis. The mGPS is validated for prognostication of relapse and increased mortality in different types of cancer [25–27], and can also predict clinical outcomes in patients with heart failure with preserved ejection fraction [28]. While no studies to date have investigated its usefulness in COVID-19, we have demonstrated that mGPS is a potential tool for predicting risk of 28-day mortality. For a score of 2 vs. 1 or 0, we found an absolute increase of 26.3% (26.3% vs. 0.0%) mortality risk. This was the largest increase amongst our biomarkers but needs to be interpreted more cautiously given the high proportion of missing mGPS data. Also, while an mGPS score of 2 versus 1 or 0 conferred the greatest relative risk of death compared to cutoffs in other biomarkers in the unadjusted analysis, this did not hold in the adjusted analysis. The relationship between mGPS and COVID-19 needs to be studied further to validate its role in prognostication.

**Other inflammatory biomarkers**

The remainder of the analysis showed mixed results for many other biomarkers that are mostly in keeping with the available literature on the topic, particularly for LDH, CK and Albumin. [8, 12, 18, 19, 29–31]. Notably, our analysis did not support D-dimer’s use for prognosticating mortality risk, however, it must note that we are limited in our recommendation due to the high degree of missing D-dimer data. However, our analysis did find that it was significant in predicting admission to hospital. Previous studies have been inconsistent in reporting the role of D-dimer, varying from reports to support its role in predicting disease severity, to its association with modestly increased mortality risk and even extreme high risk of mortality [8, 11, 17–19]. Other biomarkers studied include platelets and lymphocytes. These biomarkers have had very little literature written regarding associations with disease severity in COVID-19. However, our findings are mostly in keeping with these earlier findings. [8, 32].

Our analysis found varying associations between biomarkers and disease severity (admission to ICU or hospital) and 28-day mortality. The reason for why specific biomarkers may predict disease severity (admission to hospital or ICU) but not overall mortality or vice-versa may be explained by correlations between these biomarkers and clinician decision making.

**Strengths and limitations**

Our study has many strengths, including the relatively large size of our patient cohort and the fact that this cohort is specifically North American. Furthermore, we tested previously established COVID-19 biomarkers, as well as novel inflammatory indices including L/WBC ratio and mGPS. Our analysis of the previously studied biomarkers bolsters evidence for generalizability in North American populations and provides specific cut-off values for clinicians to utilize for predicting mortality. We believe that these preliminary North American studies provide clinicians an important piece of information that helps determine objective risk stratification. Interestingly, there is often a discrepancy of underlying pathophysiology that portends a worst outcome and the initial presentation of COVID-19 patients. Inflammatory biomarkers and useful cut offs can be used to help make objective risk stratification of patients. This can be useful in patients who come in initially well but have elevated inflammatory markers, which may prompt for more initial aggressive monitoring and treatment. Furthermore, this study is the first North American study that may serve as the basis for integration clinical prediction models validated for specific questions, including treatment decisions.

Missing data are a notable limitation. The missingness of our data varied by biomarker and was associated with disease severity, namely biomarkers such as CRP or Ferritin were more likely to be ordered in patients in the ICU. Although we used multiple imputation to account for the missing data, there are known limitations to the method [33]. Furthermore, it is difficult to deal with potential superimposed infection that may confound whether the biomarkers truly represent COVID-19 infection or the unknown superimposed infection. We think that using admission blood at least reduces the risk of bias but avoids potential infection that often occurs in COVID-19 patients. However, it would be difficult to rule out this confounder if potential infections occurred prior to admission. Lastly, given that all the biomarkers tested exhibited significant correlation, multivariate models were not produced, as the models would experience high degrees of multicollinearity.

**Conclusion and relevance**

We have demonstrated that several commonly used inflammatory biomarkers can be used to predict 28-day mortality, including C-reactive protein, neutrophil-to-lymphocyte ratio, and lymphocyte-to-white blood count ratio. We also demonstrated the predictive value of less commonly used composite measures such as the modified Glasgow prognostic score. The validation of these specific biomarkers may
be useful in monitoring response to therapies, especially as more and more drug therapies are developed and tested. We derived novel cutoff values for these biomarkers which can be used at the time of admission to help predict mortality and inform clinicians’ decision-making in North American populations. Because components of the complete blood count are readily available, we would recommend the routine use of these biomarkers in prognosticating disease courses. Further study needs to be done to determine whether clinicians can use these biomarkers to guide therapies in patients with COVID-19.

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Compliance with ethical standards

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