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Prognostic accuracy of triage tools for adults with suspected COVID-19 in a prehospital setting: an observational cohort study

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ABSTRACT

Background Tools proposed to triage patient acuity in COVID-19 infection have only been validated in hospital populations. We estimated the accuracy of five risk-stratification tools recommended to predict severe illness and compared accuracy to existing clinical decision making in a prehospital setting.

Methods An observational cohort study using linked ambulance service data for patients attended by Emergency Medical Service (EMS) crews in the Yorkshire and Humber region of England between 26 March 2020 and 25 June 2020 was conducted to assess performance of the Pandemic Respiratory Infection Emergency System Triage (PRIEST) tool, National Early Warning Score (NEWS2), WHO algorithm, CRB-65 and Pandemic Medical Early Warning Score (PMEWS) in patients with suspected COVID-19 infection. The primary outcome was death or need for organ support.

Results Of the 7549 patients in our cohort, 17.6% (95% CI 16.8% to 18.5%) experienced the primary outcome. The NEWS2 (National Early Warning Score, version 2), PMEWS, PRIEST tool and WHO algorithm identified patients at risk of adverse outcomes with a high sensitivity (>0.95) and specificity ranging from 0.3 (NEWS2) to 0.41 (PRIEST tool). The high sensitivity of NEWS2 and PMEWS was achieved by using lower thresholds than previously recommended. On index assessment, 65% of patients were transported to hospital and EMS decision to transfer patients achieved a sensitivity of 0.84 (95% CI 0.83 to 0.85) and specificity of 0.39 (95% CI 0.39 to 0.40).

Conclusion Use of NEWS2, PMEWS, PRIEST tool and WHO algorithm could improve sensitivity of EMS triage of patients with suspected COVID-19 infection. Use of the PRIEST tool would improve sensitivity of triage without increasing the number of patients conveyed to hospital.

BACKGROUND

Emergency Medical Service (EMS) and other urgent and emergency care practitioners assessing patients with suspected COVID-19 infection in the community must rapidly determine whether patients need treatment in hospital or can safely remain at home. The overall risk of mortality in patients with confirmed infection is around 1%, and if conveyance is too liberal, hospitals could be

Key messages

What is already known on this subject

- Current methods recommended to risk assess and determine if patients with suspected COVID-19 in the community and prehospital setting require treatment in hospital are consensus based.
- Triage tools such as the PRIEST score have been shown to accurately predict adverse outcomes in patients with suspected COVID-19 infection in the ED. Validation in a prehospital setting could aid clinical decision making.

What this study adds

- We retrospectively assessed the ability of five triage tools in patients evaluated in the prehospital setting by a single ambulance service. Any of the tools could potentially improve identification of patients with suspected COVID-19 infection who were at risk of adverse outcomes, compared with existing practice.
- NEWS2 (National Early Warning Score, version 2), PMEWS, PRIEST tool and WHO algorithm identified patients at risk of adverse outcome with high sensitivity (>0.95).
- Use of the PRIEST tool could lead to significant gains in sensitivity without increasing the number of patients conveyed to hospital.

overwhelmed by patients who require no specific treatment.¹ However, failing to identify a patient at risk of serious deterioration could lead to avoidable harm.²

Prognostic research has almost exclusively been conducted in hospital settings, and current national and international guidelines for risk stratification of patients with suspected COVID-19 in the community are consensus based.^{1 3-5} Clinical acuity scores, such as the UK Royal College of Physicians National Early Warning Score, version 2 (NEWS2), have been suggested in some guidelines as a way to risk stratify patients with suspected COVID-19 infection in the community.⁶ The WHO decision-making algorithm for respiratory infection and CRB-65 are used to risk stratify patients with bacterial pneumonia and



1

PMEWS for use in patients with influenza.^{7–9} However, the accuracy of these risk-stratification tools has only been validated in hospitalised or non-COVID populations.

NEWS2 has shown good prediction of adverse outcome in patients attending the ED with suspected COVID-19.⁷ The PRIEST tool was derived by adding age, sex and performance status to NEWS2, and validation showed improved prediction compared with NEWS2 alone.^{10 11} Validation of the PRIEST tool, NEWS2 and other clinical risk-stratification tools recommended for use in hospital in a community setting^{7 10 12-14} could identify the most accurate means to triage need for hospitalisation, thereby reducing unnecessary hospital attendances and improve the identification of those most at risk of serious adverse outcomes.

Our study aimed to:

- 1. Estimate the accuracy of risk-stratification tools recommended to predict severe illness in adults with suspected COVID-19 infection in a prehospital setting.
- 2. Compare the accuracy of risk-stratification tools to existing clinical decision making around transport to hospital.

METHODS

Study design

This observational cohort study used linked routinely collected EMS data to assess the accuracy in a community setting of five clinical risk-stratification tools (PRIEST tool, NEWS2, WHO algorithm, CRB-65 and PMEWS) recommended for use in hospitalised patients with COVID-19 or similar respiratory infections (triage tools shown in online supplemental material 1).^{7 10}12-14

Setting

Patients with suspected COVID-19 infection attended by EMS provided by Yorkshire Ambulance Service NHS Trust (YAS). EMS provided by YAS covers a region in the north of England of approximately 6000 square miles and with a population of *5.5* million.

Data sources and linkage

EMS providers complete an electronic patient report form (ePRF) each time they attend an emergency call, which records presenting patient characteristics and clinical care in a standardised manner. YAS provided a dataset of ePRF data for all EMS responses between the 26 March 2020 and 25 June 2020 where the attending ambulance staff recorded a clinical impression of suspected or confirmed COVID-19 infection. The dataset consisted of patient identifiers, demographic data, measured physiological parameters, other available clinical information and the outcome of the assessment (including whether the patient was conveyed to hospital). Ambulance attendances were linked to routinely collected limited COVID-related general practice (GP) records, ED attendances, hospital inpatient admissions, critical care periods and death registrations from the UK Office of National Statistics (online supplemental material 2).

Inclusion criteria

Our final cohort consisted of all adult (aged 16 years and over) patients at the time of first (index) EMS attendance between 26 March and 25 June 2020, in which the attending ambulance staff recorded a clinical impression of suspected or confirmed COVID-19 infection and who were successfully traced by NHS Digital.

Outcome

The primary outcome was death, renal, respiratory or cardiovascular organ support (identified from death registration and critical care data) at 30 days from index attendance. This includes basic and advanced cardiovascular and respiratory organ support such as high-flow oxygen administration, non-invasive ventilation and close monitoring due to the risk of respiratory failure in a higher dependency setting.

The secondary outcome was death up to 30 days from index contact.

Patient characteristics

Physiological parameters were extracted from the first (primary) set of clinical observations recorded by the ambulance crew. Consistent with methods used to estimate the Charlson Comorbidity Index from the available routine data, comorbidities were included if recorded 12 months before the index EMS attendance.¹⁵ ¹⁶ In a similar way, only immunosuppressant drug prescriptions documented in GP records within 30 days before the index attendance contributed to the immunosuppression comorbidity variable. Pregnancy status was based on GP records recorded in the previous 9 months. Frailty in patients older than 65 years was derived from the latest recorded Clinical Frailty Scale (CFS) score (if recorded) in the electronic GP records prior to index attendance.¹⁷ Patients under the age of 65 years were not given a CFS score since it is not validated in this age group. However, if a CFS score was required to calculate a triage tool and the patient was under the age of 65 years, it was assumed to be 1. Performance status was estimated from the CFS.

Analysis

We retrospectively applied the five triage tools to our cohort to assess their accuracy for the primary and secondary outcomes.^{7 10 12-14} Online supplemental material 1 provides details of scoring and handling missing data for the triage tools. For each tool we plotted the receiver operating characteristic (ROC) curve and calculated the area under the ROC curve (c-statistic) for discriminating between patients with and without adverse outcomes. We calculated sensitivity, specificity, positive predictive value (PPV) and negative predictive value (NPV) at the following prespecified decision-making thresholds based on recommended or usual use: 0 vs 1+ CRB-65; 0-1 vs 2+ NEWS2; 0-2 vs 3+ PMEWS; 0-4 vs 5+ PRIEST; 0 vs 1 WHO score. A score of ≥ 1 for CRB-65 and the WHO score are recommended thresholds for indicating consideration of hospital admission in bacterial pneumonia.^{18 19} The NEWS2 and PMEWS thresholds used are lower than previously proposed (0-3 vs 4+ NEWS and 0-3 vs 4+ PMEWS) for triaging patient acuity and are based on the assessment of their performance in a UK ED population of patients with suspected COVID-19 infection, where higher thresholds gave suboptimal sensitivity.¹¹ The threshold for the PRIEST score is also based on performance in this ED cohort.¹¹ These tools were compared with the sensitivity, specificity, PPV and NPV of EMS clinicians' decision to transfer patients to hospital. The International Severe Acute Respiratory Infection Consortium (Coronavirus Clinical Characterisation Consortium) prediction model was not validated as it requires investigations that are available in a hospital setting, including blood tests, and is intended for prediction of inpatient mortality.²⁰ All analyses were based on assessment during the index EMS attendance and completed with SAS V.9.4.

Sample size

We a priori assessed the required sample size on the estimated precision of the area under the ROC curve based on a likely



Triage tool performance

Sensitivity, specificity, PPV and NPV for predicting the primary composite outcome using predefined score thresholds are provided in table 2, and the secondary outcome of death is given in table 3. Sensitivity and specificity statistics are provided for every score threshold in online supplemental material 4, and calibration plots for the primary outcome in online supplemental material 5 (WHO algorithm excluded as the binary nature of the decision rule prevented calculation). The ROC curves for these analyses are shown in figures 2–3).

EMS decision to transfer patients to hospital had a sensitivity of 0.84 (95% CI 0.83 to 0.85) and specificity 0.39 (95% CI 0.39 to 0.40) for the primary outcome. PPV was 0.23 (95% CI 0.22 to 0.23) and NPV was 0.92 (95% CI 0.92 to 0.92). Hypothetical use of any of the five triage tools would have achieved a higher sensitivity than the decision to transfer to hospital by the EMS crews within the cohort, but in the case of NEWS2, WHO algorithm and PMEWS, this was at a cost of a lower specificity (table 2). Of the tools assessed at the predetermined thresholds, CRB-65 achieved the highest specificity but at the



Figure 1 STROBE flow diagram of study population selection.

EMS, Emergency Medical Service; STROBE, Strengthening the Reporting of Observational Studies in Epidemiology; YAS, Yorkshire Ambulance Service NHS Trust; APC, Admitted Patient Care; CC, Critical Care; ECDS, Emergency Care Data Set; DEMO, Demographics; DR, Death Registrations; GDPPR, General Practice Data for Pandemic Planning and Research.

5% event rate in a cohort of 6000 patients (online supplemental material 3).²¹

Patient and public involvement

The Sheffield Emergency Care Forum (SECF) is a public representative group interested in emergency care research.²² Members of SECF advised on the development of the PRIEST study and two members joined the Study Steering Committee. A PRIEST study patient-public involvement (PPI) group was created during the study which included patients who had been admitted to hospital with COVID-19 or their family members. Although not involved in data linkage or conducting the analyses, both PPI groups were consulted regarding study design, particularly the ethical implications of using routine health data for research. All study findings were presented and discussed with the PPI groups. Members helped with interpretation of findings particularly regarding acceptable risk of misclassification.

RESULTS

All totals presented from NHS Digital-derived datasets (sex, number of current medications, comorbidities, CFS and outcomes) are rounded to the nearest 5, with small numbers suppressed to comply with NHS Digital data disclosure guidance.

Study population

Figure 1 and table 1 summarise study cohort derivation and the characteristics of 7549 included individual adult patients. In total, 1330 patients (17.6%, 95% CI 16.8% to 18.5%) experienced the primary outcome (death or organ support) and 1065 (14.1%, 95% CI 13.4% to 14.9%) the secondary outcome

Marincowitz C, et al. Emerg Med J 2022;0:1-8. doi:10.1136/emermed-2021-211934

Table 1 Patient characteristics	by outcome			
Characteristic	Statistic/level	Adverse outcome	No adverse outcome	Total
	Ν	1330 (17.6%)*	6220 (82.4%)*	7549
Age (years)*	Mean (SD)	74.5 (15.4)	56.9 (19.4)	60 (20)
	Median (IQR)	78 (65–86)	56 (42–73)	59 (45–77)
	Range	19–103	16–105	16–105
Sex*	Male	760 (57.3%)	2825 (45.4%)	3590 (47.5%)
	Female	570 (42.7%)	3390 (54.6%)	3960 (52.5%)
No of current medications*	Ν	1330	6220	7549
	Mean (SD)	4.5 (3.3)	3.2 (3.3)	3.4 (3.3)
	Median (IQR)	4 (2–7)	2 (0–5)	3 (0–6)
	Range	0–19	0–19	0–19
Comorbidities*	Cardiovascular disease	95 (7%)	290 (4.6%)	380 (5.1%)
	Chronic respiratory disease	375 (28%)	1855 (29.8%)	2230 (29.5%)
	Diabetes	390 (29.2%)	995 (16%)	1380 (18.3%)
	Hypertension	610 (45.8%)	1765 (28.4%)	2375 (31.4%)
	Immunosuppression	280 (21.1%)	930 (15%)	1215 (16.1%)
	Active malignancy	60 (4.6%)	115 (1.9%)	180 (2.3%)
	Renal impairment	55 (4.1%)	125 (2%)	180 (2.4%)
	Stroke	30 (2.3%)	85 (1.4%)	115 (1.5%)
Clinical frailty*	Not Applicable (age <65 years)	330 (47.5%)	3985 (86.4%)	4310 (81.3%)
	Missing	645	1605	2250
	1–3	20 (4.7%)	40 (6.4%)	60 (5.8%)
	4–6	75 (20.5%)	240 (37.7%)	310 (31.4%)
	7–9	270 (74.8%)	350 (55.9%)	620 (62.8%)
AVPU**	Missing†	13	58	71
	Alert	1002 (76%)	5860 (95.1%)	6862 (91.8%)
	Confusion	125 (9.5%)	188 (3.1%)	313 (4.2%)
	Voice	100 (7.6%)	84 (1.4%)	184 (2.5%)
	Pain	64 (4.9%)	21 (0.3%)	85 (1.1%)
	Unresponsive	27 (2%)	7 (0.1%)	34 (0.5%)
GCS	Ν	1297	6085	7382
	Mean (SD)	13.7 (2.4)	14.8 (0.8)	14.6 (1.3)
	Median (IQR)	15 (14–15)	15 (15–15)	15 (15–15)
	Range	3–15	3–15	3–15
Diastolic BP (mm Hg)	N	1278	6029	7307
	Mean (SD)	76.7 (17.7)	84.5 (15.9)	83.1 (16.5)
	Median (IQR)	76 (65–87)	84 (74–94)	83 (72–93)
	Range	0–193	22-16/	0–193
Systolic BP (mm Hg)	N Marin (CD)	1277	6032	/309
	Mean (SD)	133.2 (25.8)	140.2 (23.2)	139 (23.9)
	Median (IQR)	132 (116-148)	139 (124–153)	138 (123-152)
Dulas voto (hasta (min)	Range	05-238	33-237	33-238
ruise rate (beats/min)	Moon (SD)	100 2 (22 5)		7455
	Median (JOP)	00.(84, 115)	90 (19.5)	90.7 (20.1)
	Banga	99 (04-115) 29 104	94 (82–109) 7, 100	95 (62-110) 7, 104
Respiratory rate (breaths/min)	N	1315	6145	7-194
Respiratory rate (breathsmin)	Mean (SD)	30.1 (10)	23 1 (6 0)	24.4.(8)
	Median (IOR)	28 (22-36)	20.1 (0.5)	24.4 (0)
	Range	0-76	0-84	0_84
Oxygen saturation	Missing	36	109	145
oxygen saturation	>95% on air	142 (11%)	3532 (57.8%)	3674 (49.6%)
	94%–95% on air	134 (10.3%)	854 (14%)	988 (13.3%)
	92%–93% on air	109 (8 4%)	449 (7 3%)	558 (7.5%)
	<92% on air or 0, given	910 (70 3%)	1274 (20.9%)	2184 (29 5%)
Blood alucose (mmol/L)	N	982	4021	5003
Store gracose (mmol/L)	Mean (SD)	8.1 (4)	69(32)	7,2 (3,4)
	Median (IOR)	6.8 (5.6–9)	6 (5 2–7 3)	6.2 (5.2–7.7)
	Denne	0.0 25	11.220	0.0 25
	пануе	0.3–30	1.1-33.8	Continued

Table 1 Continued				
Characteristic	Statistic/level	Adverse outcome	No adverse outcome	Total
Temperature (°C)	N	1301	6115	7416
	Mean (SD)	38.1 (1.2)	37.8 (1.1)	37.8 (1.1)
	Median (IQR)	38.2 (37.4–38.9)	37.7 (37–38.5)	37.8 (37–38.6)
	Range	32–42	34–41.7	32–42

*To comply with NHS Digital disclosure guidance, totals for these variables are rounded to the nearest 5 which may result in apparent disparities in the overall totals.

cost of sensitivity, and the PRIEST tool achieved a balance between sensitivity, specificity and C-statistic of 0.83 (95% CI 0.82 to 0.84). The triage tools generally demonstrated better discrimination (except NEWS2) and a higher sensitivity for the secondary outcome, but a lower specificity (table 3). However, when predicting inpatient admission, the tools showed lower sensitivity, higher specificity and similar overall discrimination (online supplemental material 6). A lower threshold could be used for the PRIEST score, NEWS2 and PMEWS to increase the sensitivity, at the cost of reduced specificity, for this less critical outcome.

DISCUSSION

Summary

The NEWS2, PMEWS, PRIEST tool and WHO algorithm identified patients at risk of adverse outcome with high sensitivity (>0.95) and specificity ranging between 0.3 (NEWS2) and 0.41 (PRIEST tool). They are, therefore, potentially suitable for use as triage tools to select patients for transfer to hospital. The high sensitivity of NEWS2 and PMEWS was achieved by using lower thresholds (NEWS2 0–1 vs 2+ and PMEWS 0–2 vs 3+) than previously recommended, based on performance in an ED population of patients with suspected COVID-19 infection.¹¹

At index attendance, 65% of patients were transported to hospital. Although a useful comparator for triage tool performance, the observed accuracy of EMS decision making to transfer patients to hospital does not account for clinical best interest decisions not to covey patients to hospital who subsequently deteriorated, or patient wishes not to be conveyed. The sensitivity (0.84, 95% CI 0.83 to 0.85) and specificity (0.39, 95% CI 0.39 to 0.40) achieved by EMS decision making is nonetheless similar to that of tools used to triage undifferentiated patient acuity in the ED.²³

To be clinically useful to EMS crews, the use of a triage tool would need to improve on the existing sensitivity of clinical decision making, thereby reducing the risk of not transporting a patient to hospital who subsequently deteriorates, without leading to a disproportionately large increase in hospital conveyance. Use of any of the five triage tools at the prespecified thresholds would potentially improve on the sensitivity of existing EMS decision making. However, the use of PMEWS, WHO algorithm or NEWS2 would lead to up to a 10% increase in ED conveyances (table 2). Use of both CRB-65 and the PRIEST score would lead to improvements in sensitivity without sacrificing specificity. CRB-65 achieved the highest specificity of any of the tools (0.54, 95% CI 0.53 to 0.54), and its use would reduce the number of patients conveyed to hospital by around 10%. However, patients not conveyed to hospital would have around a 4% risk of subsequently deteriorating. The PRIEST tool achieved a sensitivity of 0.97 (95% CI 0.97 to 0.97) without increasing the number of patients transported to hospital. Using the PRIEST tool, patients who were not conveyed to hospital would have a 2% risk of subsequent deterioration (compared with an estimated 8% on EMS decision making in this cohort).

Strengths and limitations

Previous evaluations of triage tool accuracy and prognostic COVID-19 prognostic research in the prehospital setting are limited by only including patients who were subsequently admitted to hospital.²⁴⁻²⁷ This is the first evaluation to use a large cohort of patients identified from routinely collected EMS records and linked to nationally collected, patient-level, healthcare data to provide robust outcome data for all patients including those not conveyed to hospital. We had low rates of missing data in the variables used in the triage tools assessed (table 1). The PRIEST tool was robust to the removal of the performance status parameter; when doing so the C-statistic remained at 0.83 (95% CI 0.82 to 0.84). We also assessed the performance of triage tools in a cohort of patients with suspected infection which, in the absence of accurate universally available rapid COVID-19 diagnostic tests, reflects the population which EMS staff must clinically triage. Most existing research either

Table 2	Triage tool diagnostic accuracy statistics (95% CI) for predicting any adverse outcome								
Tool	N*	n (%) adverse outcome†	C-statistic	Threshold	n (%) above threshold	Sensitivity	Specificity	PPV	NPV
CRB-65	7469	1315 (18)	0.79 (0.78 to 0.80)	>0	4010 (54)	0.89 (0.88 to 0.89)	0.54 (0.53 to 0.54)	0.29 (0.29 to 0.30)	0.96 (0.95 to 0.96)
NEWS2	7433	1315 (18)	0.80 (0.78 to 0.81)	>1	5574 (75)	0.96 (0.96 to 0.96)	0.30 (0.29 to 0.30)	0.23 (0.22 to 0.23)	0.97 (0.97 to 0.97)
PMEWS	7460	1315 (18)	0.81 (0.80 to 0.83)	>2	5352 (72)	0.98 (0.97 to 0.98)	0.34 (0.33 to 0.34)	0.24 (0.24 to 0.24)	0.99 (0.98 to 0.99)
PRIEST	7471	1315 (18)	0.83 (0.82 to 0.84)	>4	4932 (66)	0.97 (0.97 to 0.97)	0.41 (0.40 to 0.41)	0.26 (0.25 to 0.26)	0.98 (0.98 to 0.99)
WHO	7471	1315 (18)	0.64 (0.64 to 0.65)	>0	5539 (74)	0.98 (0.97 to 0.98)	0.31 (0.30 to 0.31)	0.23 (0.23 to 0.24)	0.98 (0.98 to 0.99)

*Patients with less than three parameters were excluded from analysis when estimating performance.

†Totals rounded to nearest 5.

NEWS2, National Early Warning Score, version 2; NPV, negative predictive value; PPV, positive predictive value.

Marincowitz C, et al. Emerg Med J 2022;0:1-8. doi:10.1136/emermed-2021-211934

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Tool	N*	n (%) 30 days of death†	C-statistic	Threshold	n (%) above threshold	Sensitivity	Specificity	PPV	NPV
CRB-65	7469	1055 (14)	0.84 (0.83 to 0.85)	>0	4010 (54)	0.95 (0.95 to 0.96)	0.53 (0.53 to 0.54)	0.25 (0.25 to 0.26)	0.99 (0.98 to 0.99)
NEWS2	7433	1055 (14)	0.78 (0.77 to 0.80)	>1	5574 (75)	0.95 (0.95 to 0.96)	0.28 (0.28 to 0.29)	0.18 (0.18 to 0.18)	0.97 (0.97 to 0.98)
PMEWS	7460	1050 (14)	0.81 (0.80 to 0.83)	>2	5352 (72)	0.98 (0.98 to 0.98)	0.32 (0.32 to 0.33)	0.19 (0.19 to 0.20)	0.99 (0.99 to 0.99)
PRIEST	7471	1055 (14)	0.85 (0.84 to 0.86)	>4	4932 (66)	0.98 (0.97 to 0.98)	0.39 (0.39 to 0.40)	0.21 (0.20 to 0.21)	0.99 (0.99 to 0.99)
WHO	7471	1055 (14)	0.65 (0.64 to 0.65)	>0	5539 (74)	0.99 (0.99 to 1.00)	0.30 (0.30 to 0.30)	0.19 (0.19 to 0.19)	1.00 (1.00 to 1.00)

 Table 3
 Triage tool diagnostic accuracy statistics (95% CI) for predicting death within 30 days

*Patients with less than three parameters were excluded from analysis when estimating performance. †Totals rounded to nearest 5.

NEWS2, National Early Warning Score, version 2; NPV, negative predictive value; PPV, positive predictive value

aimed to determine if patients with suspected infection have COVID-19 or to risk stratify patients with confirmed infection in a hospital setting.²⁸

Our evaluation of triage tool accuracy is limited to a single ambulance service, although one covering a large population across the North of England, so the results may not be generalisable to other healthcare settings. Other ambulance services may serve populations with a different risk profile, provide different types of EMS response or have different thresholds and guidelines regarding when to convey patients to hospital. The population used is likely to have similar baseline characteristics to that used to derive and validate the PRIEST score in an ED population, as it was conducted at the same time at hospitals in the region (and elsewhere in the UK).¹⁰ A sensitivity analysis in which patients recruited to the ED-based PRIEST study were removed from analysis, did not affect estimates of triage tool performance. The PRIEST tool may perform less well if applied to a different, especially non-UK, healthcare setting.

We assumed that if comorbidities were not recorded in routine data within the previous 12 months of the index event, then they were not present. Our cohort is based on the clinical impression of likely COVID-19 infection as determined by EMS crews. This is partly determined by the prevalence of COVID-19 infection which varied during the study period; however, YAS guidance stated possible COVID-19 infection should be considered in all patients with shortness of breath, cough or fever and in patients with a history of close contact with someone with these symptoms.

Our analysis is conducted on a cohort of patients that predates COVID-19 vaccination and improved community testing in the UK. This may affect the population characteristics of those with suspected COVID-19 being assessed by EMS crews. Older adults with COVID-19 may now have a lower risk of serious adverse outcomes due to vaccination than when our data were collected (over 90% of adults aged 70 years or above are now vaccinated in the UK), while infections are now more common in younger unvaccinated adults.²⁹ This could reduce the prevalence of the primary study outcome. This would not necessarily affect the estimated sensitivity and specificity of the assessed triage tools but could affect the reported NPV and PPV. Increased community testing could reduce the proportion of patients with symptoms from respiratory diseases other than COVID-19 suspected of having COVID-19 when assessed by EMS crews. In the ED cohort of the PRIEST study, patients with COVID-19 had approximately double the mortality rate and rate of organ support compared with patients with other respiratory illnesses.³⁰ We retrospectively applied risk-stratification tools to our cohort and they may perform differently if used by EMS crews. Further prospective validation may be required to assess the impact of implementation and determine the optimal threshold for clinical use of triage tools to account for changes in the population of patients with suspected COVID-19 infection.



Figure 2 Receiver operating characteristic curves showing triage tool performance for predicting any adverse outcome.



Figure 3 Receiver operating characteristic curves showing triage tool performance for predicting death within 30 days.

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Implications

Clinical tools should be used in conjunction with clinical decision making when determining whether a patient needs to be conveyed to hospital by EMS crews. As previously highlighted, there may be good clinical reasons why patients who subsequently deteriorated were not conveyed to hospital in this cohort. It may also be appropriate based on initial EMS assessment, to leave a patient at home on first attendance, who then deteriorates. Within these limitations which may give the tested triage tools an apparent advantage over observed EMS practice, our study provides evidence that the use of existing clinical triage tools may improve clinical decision making in a prehospital setting where the prevalence of serious adverse outcomes is similar to the ED.

In healthcare contexts where minimising risk of adverse outcomes in those not conveyed to hospital is the priority, the use of PMEWS or the WHO criteria may be recommended, as they appear to optimise sensitivity. Use of the COVID-specific PRIEST tool would achieve almost the same gains in sensitivity (0.97 vs 0.98) without leading to a corresponding increase in patients being unnecessarily conveyed to hospital. The use of CRB-65 would maximise specificity over gains in sensitivity, with a 4% risk of adverse outcomes in patients left to self-care in the community. This may be appropriate in resource-constrained healthcare contexts, and as oxygen saturations do not form part of the assessment tool, it can be practically applied to a large range of healthcare settings.

Further research assessing triage tool performance alongside clinical judgement in the prehospital setting would be helpful to determine whether triage tools would improve accuracy of decisions to transfer patients to hospital in practice. In November 2020, the PRIEST tool was used within the YAS senior clinical support cell to assist with clinical decision making when providing advice to EMS crews. Support cell advice was based on the risk of subsequent deterioration (as estimated by the PRIEST score) alongside other patient factors including patient preference and best interest decision making at the end of life. Given the high prevalence of adverse outcomes in this cohort, the findings may not be applicable to other lower risk community settings (eg, patients being assessed by general practitioners) and therefore similar research is needed for these populations.

CONCLUSION

The NEWS2, PMEWS, PRIEST tool and WHO algorithm achieved high estimated sensitivities with respect to death or organ support. Although there may be good clinical reasons why patients who deteriorated were not conveyed to hospital, the use of any these triage tools would potentially reduce the risk of false-negative triage and non-conveyance of a patient who subsequently deteriorates. Use of NEWS2, PMEWS and WHO algorithm would increase the proportion of conveyed patients, while the PRIEST tool could lead to significant gains in sensitivity without increasing the number of patients conveyed to hospital.

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Patient consent for publication Not required.

Ethics approval This study involves human participants and was approved by North West—Haydock Research Ethics Committee (REC reference: 12/NW/0303). The study was granted permission to conduct its research on COVID as an Urgent Public Health study in the UK under the Control of Patient Information notice.

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Data availability statement Data may be obtained from a third party and are not publicly available. The data used for this study are subject to data sharing agreements with NHS Digital and YAS which prohibits further sharing of individual level data. The datasets used are obtainable from these organisations subject to necessary authorisations and approvals.

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REFERENCES

- Bentivegna M, Hulme C, Ebell MH. Primary care relevant risk factors for adverse outcomes in patients with COVID-19 infection: a systematic review. J Am Board Fam Med 2021;34:S113–26.
- 2 Gandhi RT, Lynch JB, del Rio C. Mild or moderate Covid-19. *N Engl J Med Overseas Ed* 2020;383:1757–66.
- 3 Clinical care of severe acute respiratory infections tool kit, Accessed 27 Aug 2020. World health organisation. Available: https://www.who.int/publications-detail/clinicalcare-of-severe-acute-respiratory-infections-tool-kit
- 4 COVID-19 rapid guideline: managing COVID-19. Available: https://www.nice.org.uk/ guidance/ng191
- 5 Interim clinical guidance for management of patients with confirmed coronavirus disease (COVID-19). Available: https://www.cdc.gov/coronavirus/2019-ncov/hcp/ clinical-guidance-management-patients.html
- 6 Greenhalgh T, Thompson P, Weiringa S, et al. What items should be included in an early warning score for remote assessment of suspected COVID-19? qualitative and Delphi study. BMJ Open 2020;10:e042626.
- 7 Clinical care of severe acute respiratory infections tool kit. Available: https://www. who.int/publications-detail/clinical-care-of-severe-acute-respiratory-infections-tool-kit
- 8 Community-acquired pneumonia (non COVID-19). Available: https://bestpractice.bmj. com/topics/en-gb/3000108/management-recommendations
- 9 Goodacre S, Challen K, Wilson R, et al. Evaluation of triage methods used to select patients with suspected pandemic influenza for hospital admission: cohort study. *Health Technol Assess* 2010;14:173–236.
- 10 Goodacre S, Thomas B, Sutton L, et al. Derivation and validation of a clinical severity score for acutely ill adults with suspected COVID-19: the Priest observational cohort study. PLoS One 2021;16:e0245840.
- 11 Thomas B, Goodacre S, Lee E, et al. Prognostic accuracy of emergency department triage tools for adults with suspected COVID-19: the Priest observational cohort study. Emerg Med J 2021;38:emermed-2020-210783:587–93.
- 12 Physicians RCo. National early warning score (news) 2: standardising the assessment of acute-illness severity in the NHS. updated report of a working Party. London: RCP, 2017.
- 13 Lim WS, van der Eerden MM, Laing R, *et al.* Defining community acquired pneumonia severity on presentation to hospital: an international derivation and validation study. *Thorax* 2003;58:377–82.

Original research

- 14 Challen K, Bright J, Bentley A, et al. Physiological-social score (PMEWS) vs. CURB-65 to triage pandemic influenza: a comparative validation study using communityacquired pneumonia as a proxy. BMC Health Serv Res 2007;7:33.
- 15 Campbell MJ, Jacques RM, Fotheringham J, et al. Developing a summary hospital mortality index: retrospective analysis in English hospitals over five years. BMJ 2012;344:e1001.
- 16 Street A, Maynou L, Gilbert T, et al. The use of linked routine data to optimise calculation of the hospital frailty risk score on the basis of previous hospital admissions: a retrospective observational cohort study. Lancet Healthy Longev 2021;2:e154–62.
- 17 Rockwood K, Song X, MacKnight C, *et al*. A global clinical measure of fitness and frailty in elderly people. *CMAJ* 2005;173:489–95.
- 18 McNally M, Curtain J, O'Brien KK, et al. Validity of British Thoracic Society guidance (the CRB-65 rule) for predicting the severity of pneumonia in general practice: systematic review and meta-analysis. Br J Gen Pract 2010;60:e423–33.
- 19 World Health Organisation. Clinical care of severe acute respiratory infections tool kit. Available: https://www.who.int/publications-detail/clinical-care-of-severe-acuterespiratory-infections-tool-kit [Accessed 27 Aug 2020].
- 20 Knight SR, Ho A, Pius R, et al. Risk stratification of patients admitted to hospital with covid-19 using the ISARIC who clinical characterisation protocol: development and validation of the 4C mortality score. BMJ 2020;370:m3339.
- 21 Steyerberg EW, Harrell FE, Borsboom GJ, et al. Internal validation of predictive models: efficiency of some procedures for logistic regression analysis. J Clin Epidemiol 2001;54:774-81.

- 22 Hirst E, Irving A, Goodacre S. Patient and public involvement in emergency care research. *Emerg Med J* 2016;33:665–70.
- 23 Zachariasse JM, van der Hagen V, Seiger N, et al. Performance of triage systems in emergency care: a systematic review and meta-analysis. BMJ Open 2019;9:e026471.
- 24 Lancet EA, Gonzalez D, Alexandrou NA, et al. Prehospital hypoxemia, measured by pulse oximetry, predicts hospital outcomes during the new York City COVID-19 pandemic. J Am Coll Emerg Physicians Open 2021;2:e12407.
- 25 Dillon K, Hook C, Coupland Z, et al. Pre-Hospital lowest recorded oxygen saturation independently predicts death in patients with COVID-19. Br Paramed J 2020;5:59–65.
- 26 Saberian P, Tavakoli N, Hasani-Sharamin P, et al. Accuracy of the pre-hospital triage tools (qSOFA, news, and PRESEP) in predicting probable COVID-19 patients' outcomes transferred by emergency medical services. Caspian J Intern Med 2020;11:536–43.
- 27 Inada-Kim M, Chmiel FP, Boniface MJ. Validation of home oxygen saturations as a marker of clinical deterioration in patients with suspected COVID-19. *medRxiv* 2020.
- 28 Wynants L, Van Calster B, Collins GS, et al. Prediction models for diagnosis and prognosis of covid-19: systematic review and critical appraisal. BMJ 2020;369:m1328.
- 29 Antonelli M, Penfold RS, Merino J, *et al.* Risk factors and disease profile of postvaccination SARS-CoV-2 infection in UK users of the COVID symptom study APP: a prospective, community-based, nested, case-control study. *Lancet Infect Dis* 2022;22:43–55.
- 30 Goodacre S, Thomas B, Lee E, et al. Characterisation of 22445 patients attending UK emergency departments with suspected COVID-19 infection: observational cohort study. PLoS One 2020;15:e0240206.

Supplementary Material 1: Triage tool scoring details

CRB-65:

The CRB-65 score uses four parameters, each scoring 1 point when positive and zero if negative, to give a total score between zero and five.

Four parameters:

- 1. Confusion: GCS-V is less than 4 or GCS total is less than 15 or AVPU is recorded as V, P or U
- 2. Respiratory: rate of 30 breaths per minute or more
- 3. Blood pressure: diastolic BP is 60mmHg or less or systolic BP is 90 mmHg or less
- 4. Age: 65 years or more

Missing data: The rules above effectively classify missing data as normal. The CRB-65 score is recommended for community settings where access to blood testing is more limited. It is a 4-point scale that does not include urea with the threshold as <2 for low risk, 2+ for high risk. If a patient has fewer than three of the five parameters complete, the score was not be calculated.

PMEWS:

PMEWS uses six physiological parameters and patient parameters to calculate a score from zero to 19. The score is calculated by taking the score in the table below dependent on each of the six physiological parameters then adding points for two patient parameters after if they are positive.

Physiological:

Score	3	2	1	0	1	2	3
Respiratory Rate	≤8			9-18	19-25	26-29	≥30
SaO₂	<89	90-93	94-96	>96			
Pulse Rate	≤40	41-50		51-100	101-110	111-129	≥130
Systolic BP	≤70	71-90	90-100	>100			
Temperature		≤35.0	35.1-36.0	36.1-37.9	38-38.9	≥39	
Neuro				Alert	Confused Agitated*	Voice	Pain Uncon

* confused/agitated will be defined based on GCS-V<4 or GCS total<15

Patient:

1. Add 1 point if age>65

- 2. Add 1 point if either:
 - a. Patient lives alone / no fixed abode or
 - b. has a co-morbidity (respiratory, cardiac, renal, immunosuppressed, diabetes)
 - c. performance status is more than two suggesting limited activity can self-care, limited activity limited self-care, or bed/chair bound no self-care.

Missing data: If data is missing one or two variables then the normal score (zero) was assumed. If more than three variables are missing the patient was excluded. These rules effectively classify missing data as normal. If AVPU was missing and GCS was recorded, impute the following AVPU scores using GCS: 0 if GCS=15, 1 if GCS=12-14; 2 if GCS=9-11 and 3 if GCS<9.

NEWS2:

The NEWS2 has seven parameters which are scores from zero to three providing an overall score between zero and 20. The scores for each parameter can be found in the table below.

Score	3	2	1	0	1	2	3
Respiratory Rate	≤8		9-11	12-20		21-24	≥25
SaO ₂	≤91	92-93	94-95	≥96			
Pulse Rate	≤40		41-50	51-90	91-110	111-130	≥131
Systolic BP	≤90	91-100	101-110	111-219			≥220
Temperature	≤35.0		35.1-36.0	36.1-38.0	38.1-39.0	≥39.1	
Neuro				Alert			Confusion, Voice, Pain,
							Unresponsive
Air or		Oxygen					
Oxygen		(based on		Air			
		FiO2>21%, or					
		FiO ₂ >0 L/min)					

Missing data: Any missing data will be imputed with the value zero, therefore classifying missing as normal. The score was calculated if fewer than three of the parameters were available.

WHO decision making algorithm for hospitalisation with pneumonia:

The WHO decision making algorithm for hospitalisation with pneumonia suggests an adult patient is admitted (score 1) if any of the following are present:

- respiratory rate >30/minute,
- oxygen saturation <90%,
- respiratory distress (not included in this evaluation),
- age >60,
- any of the following comorbidities; hypertension, diabetes, cardiovascular disease, chronic respiratory disease, renal impairment or immunosuppression

As a subjective clinical assessment of respiratory distress is not routinely consistently recorded in our data this was not be included. Any missing data was assumed as normal. A score was not calculated if fewer than three of the above were complete.

The PRIEST clinical severity Score:

The core PRIEST clinical severity score consists of the seven parameters of NEWS2 and age, sex and performance status. We will assessed the full score and a version which omits performances status as this may not be reliably available from GP records. Scores for each parameter are as follows.

Variable	Range	Score
Respiratory rate (per minute)	12-20	0
	9-11	1
	21-24	2
	<9 or >24	3
Oxygen saturation (%)	>95	0
	94-95	1
	92-93	2
	<92	3
Heart rate (per minute)	51-90	0
	41-50 or 91-110	1
	111-130	2
	<41 or >130	3
Systolic BP (mmHg)	111-219	0
	101-110	1
	91-100	2
	<91 or >219	3

Temperature (2 C)	36.1-38.0	0
	35.1-36.0 or 38.1-39.0	1
	>39.0	2
	<35.1	3
Alertness	Alert	0
	Confused or not alert	3
Inspired oxygen	Air	0
	Supplemental oxygen	2
Sex	Female	0
	Male	1
Age (years)	16-49	0
	50-65	2
	66-80	3
	>80	4
Performance status	Unrestricted normal activity	0
	Limited strenuous activity, can do light activity	1
	Limited activity, can self-care	2
	Limited self-care	3
	Bed/chair bound, no self-care	4

Missing data in NEWS2 parameters was handled in the same way as for NEWS2. Performance status was be assumed normal if a clinical frailty scale has not been completed in linked GP records

Supplementary Material 2: Data Sources and linkage

Health and social care data relating to the population in England within the UK National Health Service (NHS) is managed by NHS Digital. We provided patient identifiers to NHS Digital to trace patients in our cohort and supply additional individual level demographic, co-morbidity and outcome data. NHS Digital identified records in their collections belonging to patients in our cohort, and provided data on patient demographics, limited COVID-related general practice (GP) records, emergency department attendances, hospital inpatient admissions, critical care periods, and death registrations from the UK Office of National Statistics.

YAS and NHS Digital removed records where patients indicated that they did not wish their data to be used for research purposes, via the NHS data opt-out service.¹⁵ The study team also excluded patients who had opted out of any part of the PRIEST study and those with inconsistent records (e.g. multiple deaths recorded or death before latest activity). Patient identifiers across all datasets were replaced with a consistent pseudo-identifier to enable the identification and linkage of records belonging to the same patient across all datasets but without revealing any patient's identity.

Supplementary Material 3: Sample Size calculation (Precision of AUROC in cohort 6000 patients with a 5% event rate).

C- statistic	Standard error
0.85	0.014
0.80	0.016
0.75	0.017
0.70	0.017
0.65	0.018

Supplementary Material 4: Performance of triage tools across the whole range of available scores

		Primary: any ac	lverse outcome	Seconda	ry: death
Tool	Threshold	Sensitivity	Specificity	Sensitivity	Specificity
CRB-65	>0	0.89 (0.88,0.89)	0.54 (0.53,0.54)	0.95 (0.95,0.96)	0.53 (0.53,0.54)
	>1	0.6 (0.59,0.61)	0.83 (0.83,0.84)	0.71 (0.7,0.72)	0.83 (0.83,0.84)
	>2	0.25 (0.24,0.26)	0.96 (0.96,0.96)	0.3 (0.29,0.32)	0.96 (0.96,0.96)
	>3	0.05 (0.04,0.05)	1 (1,1)	0.06 (0.05,0.06)	1 (1,1)
NEWS2	>0	0.99 (0.99,0.99)	0.16 (0.16,0.16)	0.99 (0.98,0.99)	0.15 (0.15,0.16)
	>1	0.96 (0.96,0.96)	0.3 (0.29,0.3)	0.95 (0.95,0.96)	0.28 (0.28,0.29)
	>2	0.93 (0.93,0.94)	0.41 (0.41,0.42)	0.93 (0.93,0.94)	0.4 (0.39,0.4)
	>3	0.89 (0.88,0.9)	0.53 (0.52,0.53)	0.89 (0.88,0.9)	0.51 (0.5,0.51)
	>4	0.84 (0.83,0.84)	0.62 (0.62,0.63)	0.83 (0.82,0.84)	0.6 (0.6,0.61)
	>5	0.77 (0.76,0.77)	0.7 (0.7,0.71)	0.77 (0.76,0.78)	0.69 (0.68,0.69)
	>6	0.65 (0.64,0.66)	0.78 (0.78,0.79)	0.64 (0.63,0.65)	0.76 (0.76,0.77)
	>7	0.51 (0.5,0.52)	0.85 (0.85,0.85)	0.51 (0.5,0.52)	0.83 (0.83,0.84)
	>8	0.39 (0.38,0.4)	0.9 (0.9,0.91)	0.4 (0.39,0.41)	0.89 (0.89 <i>,</i> 0.9)
	>9	0.27 (0.26,0.28)	0.95 (0.94,0.95)	0.28 (0.27,0.29)	0.94 (0.94,0.94)
	>10	0.18 (0.18,0.19)	0.97 (0.97,0.97)	0.2 (0.19,0.21)	0.97 (0.96,0.97)
	>11	0.12 (0.11,0.13)	0.98 (0.98,0.98)	0.13 (0.12,0.14)	0.98 (0.98,0.98)
	>12	0.06 (0.06,0.07)	0.99 (0.99,0.99)	0.07 (0.06,0.07)	0.99 (0.99,0.99)
	>13	0.04 (0.03,0.04)	1 (1,1)	0.04 (0.04,0.05)	1 (1,1)
	>14	0.02 (0.01,0.02)	1 (1,1)	0.02 (0.01,0.02)	1 (1,1)
	>15	0.01 (0.01,0.01)	1 (1,1)	0.01 (0.01,0.01)	1 (1,1)
	>16	0 (0,0)	1 (1,1)	0 (0,0.01)	1 (1,1)
	>17	0 (0,0)	1 (1,1)	0 (0,0)	1 (1,1)
PMEWS	>0	1 (1,1)	0.08 (0.07,0.08)	1 (1,1)	0.07 (0.07,0.08)
	>1	0.99 (0.99,0.99)	0.2 (0.2,0.21)	0.99 (0.99,1)	0.2 (0.19,0.2)
	>2	0.98 (0.97,0.98)	0.34 (0.33,0.34)	0.98 (0.98,0.98)	0.33 (0.32,0.33)
	>3	0.94 (0.94,0.95)	0.47 (0.46,0.47)	0.95 (0.94,0.95)	0.45 (0.45,0.46)
	>4	0.89 (0.88,0.89)	0.58 (0.58,0.59)	0.89 (0.88,0.9)	0.56 (0.56,0.57)
	>5	0.8 (0.8,0.81)	0.68 (0.67,0.68)	0.82 (0.81,0.83)	0.66 (0.66,0.66)
	>6	0.69 (0.69,0.7)	0.77 (0.77,0.78)	0.71 (0.7,0.72)	0.76 (0.75,0.76)
	>7	0.55 (0.54,0.56)	0.85 (0.84,0.85)	0.57 (0.56,0.59)	0.83 (0.83,0.84)
	>8	0.41 (0.4,0.42)	0.9 (0.9,0.9)	0.43 (0.42,0.44)	0.89 (0.89,0.89)
	>9	0.29 (0.29,0.3)	0.94 (0.94,0.94)	0.31 (0.3,0.32)	0.94 (0.93,0.94)
	>10	0.19 (0.18,0.19)	0.97 (0.97,0.97)	0.21 (0.2,0.22)	0.97 (0.97,0.97)
	>11	0.11 (0.11,0.12)	0.99 (0.99,0.99)	0.13 (0.12,0.14)	0.99 (0.99,0.99)
	>12	0.07 (0.07,0.08)	1 (1,1)	0.08 (0.08,0.09)	1 (0.99,1)
	>13	0.04 (0.03,0.04)	1 (1,1)	0.04 (0.04,0.05)	1 (1,1)
	>14	0.01 (0.01,0.02)	1 (1,1)	0.02 (0.01,0.02)	1 (1,1)
	>15	0.01 (0.01,0.01)	1 (1,1)	0.01 (0.01,0.01)	1 (1,1)

Γ

		Primary: any ad	lverse outcome	Seconda	ry: death
Tool	Threshold	Sensitivity	Specificity	Sensitivity	Specificity
	>16	0 (0,0)	1 (1,1)	0 (0,0)	1 (1,1)
PRIEST	>0	1 (1,1)	0.05 (0.05,0.05)	1 (1,1)	0.05 (0.05,0.05)
	>1	1 (1,1)	0.12 (0.12,0.13)	1 (1,1)	0.12 (0.12,0.12)
	>2	1 (0.99,1)	0.21 (0.21,0.22)	1 (1,1)	0.21 (0.2,0.21)
	>3	0.99 (0.99,0.99)	0.31 (0.3,0.31)	0.99 (0.99,1)	0.29 (0.29,0.3)
	>4	0.97 (0.97,0.97)	0.41 (0.4,0.41)	0.98 (0.97,0.98)	0.39 (0.39,0.4)
	>5	0.95 (0.95,0.95)	0.49 (0.49,0.5)	0.96 (0.95,0.96)	0.48 (0.47,0.48)
	>6	0.92 (0.92,0.93)	0.57 (0.56,0.57)	0.94 (0.93,0.95)	0.55 (0.55,0.55)
	>7	0.89 (0.88,0.89)	0.64 (0.63,0.64)	0.91 (0.9,0.92)	0.62 (0.62,0.62)
	>8	0.83 (0.82,0.83)	0.7 (0.7,0.71)	0.86 (0.85,0.87)	0.69 (0.68,0.69)
	>9	0.75 (0.74,0.76)	0.76 (0.76,0.76)	0.8 (0.79,0.81)	0.75 (0.74,0.75)
	>10	0.65 (0.64,0.66)	0.81 (0.81,0.82)	0.7 (0.69,0.71)	0.8 (0.8,0.81)
	>11	0.53 (0.52,0.54)	0.86 (0.86,0.87)	0.59 (0.58,0.6)	0.86 (0.85,0.86)
	>12	0.44 (0.43,0.44)	0.9 (0.9,0.9)	0.5 (0.49,0.51)	0.9 (0.9,0.9)
	>13	0.35 (0.34,0.36)	0.93 (0.93,0.94)	0.41 (0.4,0.42)	0.93 (0.93,0.93)
	>14	0.28 (0.27,0.28)	0.95 (0.95,0.96)	0.33 (0.32,0.34)	0.95 (0.95,0.95)
	>15	0.21 (0.2,0.22)	0.97 (0.97,0.97)	0.25 (0.24,0.26)	0.97 (0.97,0.97)
	>16	0.14 (0.14,0.15)	0.98 (0.98,0.98)	0.17 (0.17,0.18)	0.98 (0.98,0.98)
	>17	0.11 (0.1,0.11)	0.99 (0.99,0.99)	0.13 (0.12,0.14)	0.99 (0.99,0.99)
	>18	0.07 (0.06,0.07)	0.99 (0.99,0.99)	0.08 (0.08,0.09)	0.99 (0.99,0.99)
	>19	0.04 (0.04,0.05)	1 (1,1)	0.05 (0.05,0.06)	1 (1,1)
	>20	0.03 (0.02,0.03)	1 (1,1)	0.03 (0.03,0.04)	1 (1,1)
	>21	0.02 (0.01,0.02)	1 (1,1)	0.02 (0.02,0.02)	1 (1,1)
	>22	0.01 (0.01,0.01)	1 (1,1)	0.01 (0.01,0.01)	1 (1,1)
	>23	0 (0,0.01)	1 (1,1)	0 (0,0.01)	1 (1,1)
	>24	0 (0,0)	1 (1,1)	0 (0,0)	1 (1,1)
	>25	0 (0,0)	1 (1,1)	0 (0,0)	1 (1,1)
WHO	>0	0.98 (0.97,0.98)	0.31 (0.3,0.31)	0.99 (0.99,1)	0.3 (0.3,0.3)



Supplementary 5: Flexible calibration curves using locally weighted scatterplot smoothing (LOESS) with 95% CIs

Supplementary Material 6: Triage tool diagnostic accuracy statistics (95% CI) for predicting inpatient admission

Tool	N*	N (%) admitted**	C-statistic	Threshold	N (%) with score	Sensitivity	Specificity	PPV	NPV
CRB-65	7460	3890 (52%)	0.73	>0	4010 (5.4%)	0.75	0.69	0.72	0.71
	7409		(0.72, 0.74)	>0	4010 (54%)	(0.74, 0.75)	(0.68, 0.69)	(0.71, 0.72)	(0.70, 0.72)
NEWS2	7422	3875 (52%)	0.81			0.92	0.43	0.64	0.83
	7455		(0.80, 0.82)	>1	5574 (75%)	(0.91, 0.92)	(0.43, 0.44)	(0.63, 0.64)	(0.82, 0.83)
PMEWS	7460	3890 (52%)	0.83	>2	E2E2 (72%)	0.92	0.50	0.67	0.85
	7460		(0.82, 0.84)	>2	5552 (72%)	(0.92, 0.92)	(0.50, 0.51)	(0.66, 0.67)	(0.85, 0.86)
PRIEST	7471	3890 (52%)	0.83	.1	4022 (66%)	0.90	0.60	0.71	0.84
	/4/1		(0.82, 0.84)	24	4952 (00%)	(0.89 <i>,</i> 0.9)	(0.59, 0.60)	(0.70, 0.71)	(0.84, 0.85)
WHO	7471	3890 (52%)	0.67	>0	EE20 (749/)	0.90	0.43	0.63	0.80
	/4/1		(0.66, 0.68)	>0	5559 (74%)	(0.90, 0.90)	(0.43, 0.44)	(0.63, 0.64)	(0.80, 0.81)

*Patients with 3 or more missing triage tool parameters were excluded from analysis when estimating performance

**Numbers rounded to nearest 5