



BMJ Open Clinical risk score to predict in-hospital mortality in COVID-19 patients: a retrospective cohort study

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ABSTRACT

Objectives Several physiological abnormalities that develop during COVID-19 are associated with increased mortality. In the present study, we aimed to develop a clinical risk score to predict the in-hospital mortality in COVID-19 patients, based on a set of variables available soon after the hospitalisation triage.

Setting Retrospective cohort study of 516 patients consecutively admitted for COVID-19 to two Italian tertiary hospitals located in Northern and Central Italy were collected from 22 February 2020 (date of first admission) to 10 April 2020.

Participants Consecutive patients ≥18 years admitted for COVID-19.

Main outcome measures Simple clinical and laboratory findings readily available after triage were compared by patients' survival status ('dead' vs 'alive'), with the objective of identifying baseline variables associated with mortality. These were used to build a COVID-19 in-hospital mortality risk score (COVID-19MRS).

Results Mean age was 67±13 years (mean±SD), and 66.9% were male. Using Cox regression analysis, tertiles of increasing age (≥75, upper vs <62 years, lower: HR 7.92; p<0.001) and number of chronic diseases (≥4 vs 0–1: HR 2.09; p=0.007), respiratory rate (HR 1.04 per unit increase; p=0.001), PaO₂/FiO₂ (HR 0.995 per unit increase; p<0.001), serum creatinine (HR 1.34 per unit increase; p<0.001) and platelet count (HR 0.995 per unit increase; p=0.001) were predictors of mortality. All six predictors were used to build the COVID-19MRS (Area Under the Curve 0.90, 95% CI 0.87 to 0.93), which proved to be highly accurate in stratifying patients at low, intermediate and high risk of in-hospital death (p<0.001).

Conclusions The COVID-19MRS is a rapid, operator-independent and inexpensive clinical tool that objectively predicts mortality in patients with COVID-19. The score could be helpful from triage to guide earlier assignment of COVID-19 patients to the most appropriate level of care.

INTRODUCTION

The first human cases of SARS-CoV-2 were reported in Wuhan, Hubei Province, China in January 2020^{1 2}; subsequently, it spread worldwide, officially being defined as a

Strengths and limitations of the study

- Risk assessment tools readily available since the triage phase of COVID-19 are lacking.
- Age, previous chronic diseases, respiratory rate, PaO₂/FiO₂, creatinine and platelet count were predictors of risk of in-hospital death.
- All six predictors were used to build a novel COVID-19 clinical risk score that proved to be highly accurate in stratifying patients at low, intermediate and high risk of death.
- Retrospective design; novel score to be validated in other, external, COVID-19 case series.

pandemic by WHO on 11 March 2020.^{3–5} Italy was the first country outside Asia to be heavily affected by the virus, with a total of 189 973 confirmed cases as of 23 April 2020. The Lombardy Region had the highest burden of mortality and strain on its healthcare system.⁶ However, a substantial reorganisation of healthcare facilities was necessary in all Italian regions to cope with the widespread and rapid increase in COVID-19 patient flow to emergency departments.

Prompt referral to the appropriate care setting (ie, low vs intermediate or high intensity) is of crucial importance to improve outcomes and healthcare resource utilisation.^{7–9} Given the high number of patients to be triaged during this emergency and the relative shortage of hospital beds, the availability of a disease-specific mortality risk score since initial triage might have been useful in identifying the appropriate level of care and reducing delay. However, there is a lack of reliable prognostic prediction models and, at present, no tool for the early stratification of mortality risk has been fully identified.¹⁰ A recent systematic review of prediction models concluded that the performance

of prognostic estimates for COVID-19 may be overoptimistic and misleading, because of the high risk of bias in patient selection, unclear outcome definition and length of follow-up.¹⁰ Recently, clinical scores to predict the occurrence of critical illness and/or fatal outcome during COVID-19 were developed in a cohort of Chinese patients belonging to more than 500 centres throughout the country.^{11 12} However, these were developed in a specific region which could potentially limit the generalisability of the risk score to other areas of the world.

Therefore, the aim of the present study was to develop a novel COVID-19 in-hospital mortality risk score (hereafter referred to as COVID-19MRS), based on data rapidly obtainable soon after hospital admission. To this end, we analysed a consecutive series of COVID-19 patients admitted to two tertiary care hospitals located in Northern and Central Italy.

METHODS

Study design

In this cohort study, we retrospectively reviewed the clinical history, laboratory and instrumental variables of all patients aged ≥ 18 years diagnosed with COVID-19,¹³ admitted to two Italian tertiary hospitals located in Northern and Central Italy (Poliambulanza Hospital, Brescia; and Careggi University Hospital, Florence) from 22 February 2020 (date of first admission in Brescia) to 10 April 2020, in order to identify a set of early predictors of mortality and build a mortality risk stratification score. The overall capacity of the two hospitals is about 1800 beds. The number of beds dedicated to COVID-19 patients progressively increased with the diffusion of the epidemic to a peak capacity of 655 (228/1200 in Careggi University Hospital and 427/600 in Poliambulanza Hospital; overall, 110 high-intensity care beds at peak).

Study population data source

A wide range of variables assessed on hospital admission were collected for each patient from electronic charts: these included demographics, number of drugs prescribed prior to admission, cardiovascular (CV) risk factors (eg, history of cigarette smoking, hypertension and diabetes), as well as data on previous chronic comorbidities (eg, CV and pulmonary diseases, cancer, depression and dementia). Functional status 2 weeks prior to hospitalisation was also assessed using the Barthel Index, in which lower values correspond to poorer function.¹⁴ Arterial blood gases, white blood cell count, lymphocyte and platelet (PLT) counts, alanine aminotransferase and aspartate aminotransferase, creatinine, creatine phosphokinase, lactate dehydrogenase (LDH) and high-sensitivity C reactive protein (CRP) were collected in all patients. Chest X-ray was also collected whenever deemed clinically indicated. Reading and interpretation of the main chest X-ray features was performed according to radiology guidelines.¹⁵ Information on respiratory support and drugs prescribed during hospital stay were

recorded. Six medical doctors (CF, MV, MC, FC, GC and FM) selectively extracted all variables from electronic charts and transferred them into a unique database and independently reviewed them for their consistency. Data were last updated on 10 April 2020.

In keeping with statements by the Italian Regulatory Authorities (<https://www.garanteprivacy.it/web/guest/home/docweb/-/docweb-display/docweb/5805552>), Ethical Committees of both hospitals (Comitato Etico Area Vasta Centro, Careggi University Hospital, Florence and Comitato Etico Fondazione Poliambulanza Hospital, Brescia, Italy) approved data collection and granted a waiver of informed consent from study participants. Patients' identity was anonymised, and information protected by password.

Study outcome

Definition of an in-hospital all-cause mortality risk score based on simple, readily available clinical and laboratory findings.

Patient and public involvement

Patients or the public were not involved in the design or conduct of our research, partially due to its retrospective nature. Public Health Authorities will be involved in the upcoming, large-scale validation of the newly presented score.

Statistical analysis and mortality risk score derivation

Continuous variables were reported as mean \pm SD or as median with IQR, respectively, for normal and non-normal distributions, whereas categorical variables were presented as counts and percentages. All variables were compared by survival status ('dead' vs 'alive') and patients still hospitalised at study closure were considered alive together with those who had been discharged during the study period. For continuous variables, comparisons were performed using t-test, analysis of variance or non-parametric tests, as appropriate. Categorical variables were compared with χ^2 test, or Fisher's exact test when any expected cell count was less than 5.

In accordance with the aim of the study, only data obtained shortly after initial triage were taken into account to build the mortality risk score. Cox multivariate regression analyses (with backward stepwise elimination) were calculated to identify baseline characteristics independently associated with the outcome, with inclusion of variables ($p < 0.10$ by univariate analysis) which were available for all patients. A two-sided $p < 0.05$ was considered statistically significant.

All continuous variables which were significantly associated with mortality by multivariate analysis were divided into tertiles and each of them was then scored from 1 to 3 to quantify the increasing mortality risk. Values obtained were then summed up to produce the mortality risk score whose predictive accuracy was tested using receiver operating characteristic (ROC) analysis. The mortality risk score was further divided into tertiles in order to identify

Table 1 Demographic and clinical characteristics on hospital admission

	Overall (N=516)	Dead (N=120)	Alive (N=396)	P value
Demographic characteristics				
Age, years, mean±SD	67±13	79±8	64±12	<0.001
Age (tertiles)				
<62, N (%)	177 (34.3)	7 (5.8)	170 (42.9)	
62–74, N (%)	171 (33.1)	27 (22.5)	144 (36.4)	
≥75, N (%)	168 (32.6)	86 (71.7)	82 (20.7)	
Hospital stay, median (IQR)	9 (5–14)	6 (3–10)	10 (6–15)	<0.001
Gender (male), N (%)	345 (66.9)	85 (70.8)	260 (65.7)	0.321
Smoking history, N (%)	112 (21.7)	26 (21.7)	86 (21.7)	0.999
Hypertension, N (%)	182 (35.3)	65 (55.6)	117 (29.6)	<0.001
Diabetes mellitus, N (%)	161 (31.4)	51 (43.6)	110 (27.8)	<0.001
CV disease, N (%)	146 (28.5)	57 (47.9)	89 (22.6)	<0.001
Previous stroke/TIA, N (%)	25 (4.9)	11 (9.1)	14 (3.5)	0.011
COPD, N (%)	36 (7.0)	12 (10)	24 (6.1)	0.120
Cancer, N (%)	50 (9.7)	23 (19.2)	27 (6.8)	<0.001
Depression, N (%)	52 (20.1)	20 (17.1)	32 (8.1)	0.005
Dementia, N (%)	18 (3.4)	12 (10.0)	6 (1.5)	<0.001
Comorbidities* mean±SD	2.1±1.7	3.2±1.9	1.8±1.6	<0.001
≥3, N (%)	179 (34.7)	68 (58.1)	111 (28.2)	<0.001
Barthel Index, mean±SD	85±28	77±27	94±13	<0.001
ACE-i/ARBs, N (%)	144 (27.9)	35 (29.2)	109 (27.5)	0.725
Drugs, N (%)	3.4±3.3	5.6±3.5	2.7±2.7	<0.001
Signs and symptoms				
Fever, N (%)	456 (89.1)	102 (87.2)	354 (89.5)	0.457
Cough, N (%)	293 (57.3)	57 (48.5)	236 (59.8)	0.032
Dyspnoea, N (%)	250 (48.9)	59 (50.4)	191 (48.5)	0.711
Respiratory rate, mean±SD	23±7	26±7	21±6	<0.001
Insomnia, N (%)	68 (13.2)	18 (15)	50 (12.6)	0.004
Diarrhoea, N (%)	47 (9.2)	10 (8.3)	37 (9.4)	0.782
Syncope, N (%)	27 (5.2)	11 (9.2)	16 (4.1)	0.023
Altered mental status, N (%)	24 (4.7)	12 (10.0)	12 (3.0)	<0.001

*Comorbidities are a composite variable including from hypertension to dementia. Percentages in brackets are calculated for numbers in columns for all dichotomous variables.

ACE-i, ACE inhibitors; ARBs, angiotensin receptor blockers; COPD, chronic obstructive pulmonary disease; CV, cardiovascular; TIA, transient ischaemic attack.

low-risk, intermediate-risk and high-risk categories, and assessed using Cox multivariate analysis. The Kaplan-Meier estimation method was computed to assess the probability of survival in patients in the different risk groups (low, intermediate and high) and compared using the log-rank test.

Analyses were performed using the SPSS V.26.0 statistical software package for Macintosh.

RESULTS

Regional trend and clinical characteristics on hospital admission

During the study period, 516 consecutive patients (301 in Brescia and 215 in Florence) diagnosed with COVID-19

were included in the study (table 1). According to date of admission, Brescia hospital anticipated both the first case (February 22 vs 25) and the peak of admissions by an average of 3 days, with a remarkably higher total and peak burden of admissions.

As of April 10, 314 (61%) patients had been discharged from hospital (273 (87%) at home and 41 (13%) to postacute facilities), 82 (16%) were still hospitalised, while 120 (23.2%) had died. Notably, no death occurred on the day of admission.

The mean age was 67±13 years (range 21–95) and 345 (66.9%) patients were male. Demographic and clinical characteristics of non-survivors and survivors are reported in table 1. Non-survivors were significantly

older (79 ± 8 vs 64 ± 12 years, $p<0.001$). Indeed, in-hospital fatality rate sharply increased with age and was more than five times higher in individuals aged ≥ 75 years (51.2% vs <75 years 9.8%; $p<0.001$). Conversely, prognosis was similar for both genders. The median hospital stay was 9 (IQR 5–14) days, significantly longer in survivors. Non-survivors also presented with a higher prevalence of CV risk factors, a greater burden of chronic comorbidities, and were more functionally impaired as indicated by a lower Barthel Index Score (table 1). Previous use of ACE inhibitors or angiotensin receptor blockers was similar in both groups while, in accordance with their higher burden of comorbidities, non-survivors reported a greater number of drugs chronically assumed prior to hospitalisation. The majority of patients presented with fever (89.1%) and/or cough (57.3%). Of note, non-survivors reported cough less frequently (48.5% vs 59.8%; $p=0.032$), but had a significantly higher prevalence of insomnia, syncope or altered mental status. While the prevalence of dyspnoea was similar in both groups (overall, 48.9%), respiratory rate on admission was higher in non-survivors than in survivors (26 ± 7 vs 21 ± 6 breaths/min; $p<0.001$).

Laboratory and imaging findings

Nasopharyngeal swab was positive in 499 (97%) patients. In the remaining patients, initially suspected diagnosis was confirmed by subsequent swabs, sputum or bronchoalveolar lavage. Laboratory findings are presented in table 2. In the entire population, median $\text{PaO}_2/\text{FiO}_2$ ratio was 269 (IQR 217–319), and values <200 were significantly associated with the probability of death. Lymphocytopenia was present in 61% of the population, more frequently among non-survivors than survivors (71% vs 58%; $p=0.011$), who also had lower PLT count and higher serum creatinine. CRP and LDH were increased in both groups and higher in non-survivors. Chest X-ray was abnormal in $>95\%$ of cases, with a trend towards a higher prevalence of interstitial or mixed (both interstitial and consolidation) patterns in deceased patients.

Medical management and clinical outcomes

Non-survivors required non-invasive (continuous positive airway pressure and biphasic positive airway pressure modes) or invasive ventilation more frequently than survivors (table 3). While antibiotics were prescribed more frequently to non-survivors, heparin, hydroxychloroquine,

Table 2 Laboratory and imaging findings on admission

	Overall (N=516)	Dead (N=120)	Alive (N=396)	P value
Laboratory findings				
$\text{PaO}_2/\text{FiO}_2$, median (IQR)	269 (217–319)	226 (169–271)	281 (232–335)	<0.001
<200 , N (%)	101 (19.6)	42 (35.0)	59 (15.0)	<0.001
≥ 200 , N (%)	415 (80.4)	78 (65.0)	337 (85.1)	
Haematocrit, % median (IQR)	41 (38–44)	39 (35–43)	42 (39–44.75)	<0.001
Haemoglobin, g/L median (IQR)	130 (117–143)	129 (117–141)	133 (122–143)	0.203
WBC, ($\times 10^9/\text{L}$) median (IQR)	6.31 (5–9)	7.11 (5–10.23)	6 (4.98–8.47)	0.009
Lymphocytes, ($\times 10^9/\text{L}$) median (IQR)	0.90 (0.70–1.24)	0.77 (0.70–1.07)	0.90 (0.70–1.24)	<0.001
Lymphocytopenia, N (%)	316 (61)	85 (71)	231 (58)	0.011
Platelets, ($\times 10^9/\text{L}$) median (IQR)	182 (142–234)	156 (117–218)	187 (152–238)	0.001
ALT, U/L median (IQR)	31 (19–51)	26 (16–42)	32 (19–58)	0.004
AST, U/L median (IQR)	46 (30–69)	50 (35–71)	45 (28–69)	0.181
Serum creatinine, mg/dL median (IQR)	0.94 (0.79–1.22)	1.23 (0.92–1.91)	0.90 (0.79–1.13)	<0.001
CPK, U/L median (IQR)	110 (64–228)	130 (60–208)	108 (64–208)	0.085
LDH, U/L median (IQR)	351 (268–480)	473 (338–610)	335 (266–437)	<0.001
CRP, mg/L median (IQR)	94 (44.3–161.8)	138 (85–188)	77 (37–152)	<0.001
Imaging	N=486	N=114	N=372	
Chest X-ray				
Negative, N (%)	20 (4.1)	2 (1.8)	18 (4.8)	0.053
Consolidation, N (%)	67 (13.8)	12 (10.5)	55 (14.8)	
Interstitial, N (%)	346 (71.2)	81 (71.1)	265 (71.2)	
Mixed, N (%)	53 (10.9)	19 (16.7)	34 (9.1)	

Percentages in round brackets are calculated for numbers in columns for all dichotomous variables

ALT, alanine aminotransferase; AST, aspartate aminotransferase; CPK, creatine phosphokinase; CRP, C reactive protein; LDH, lactate dehydrogenase; WBC, white blood cell count.

Table 3 Treatment strategies

	Overall (N=516)	Dead (N=120)	Alive (N=396)	P value
Respiratory support				
None, N (%)	57 (11.0)	2 (1.7)	55 (13.9)	<0.001
Oxygen, N (%)	334 (64.7)	78 (65.0)	256 (65.1)	
Non-invasive ventilation, N (%)	65 (12.6)	23 (19.2)	42 (10.6)	
Invasive ventilation, N (%)	60 (11.6)	17 (14.2)	43 (10.9)	
Drugs				
Antibiotics, N (%)	407 (78.9)	106 (88.3)	301 (76.0)	0.003
Heparin, N (%)	299 (57.9)	57 (47.5)	242 (61.1)	0.008
Hydroxychloroquine, N (%)	268 (51.9)	43 (35.8)	225 (56.8)	<0.001
Lopinavir/ritonavir, N (%)	247 (50.7)	39 (32.5)	208 (52.5)	<0.001
Corticosteroids, N (%)	176 (34.1)	45 (37.5)	131 (33.1)	0.371
Monoclonal antibodies, N (%)	57 (11.3)	3 (2.5)	54 (13.6)	<0.001

antiviral agents (combination of lopinavir/ritonavir) and monoclonal antibodies (mAbs, tocilizumab) were prescribed more frequently to survivors. In contrast, corticosteroid therapy was adopted in similar proportions in the two groups. Patients receiving mAbs were younger (65 ± 9 vs 68 ± 14 years, $p<0.01$) and had lower serum creatinine (0.9 ± 0.3 vs 1.2 ± 0.9 mg/dL, $p=0.024$).

Predictors of mortality and development of the mortality risk score

At Cox multivariate regression analysis (table 4), age, number of chronic comorbidities, respiratory rate and serum creatinine emerged as positive predictors, while $\text{PaO}_2/\text{FiO}_2$ ratio and PLT count were negative predictors of death. Online supplemental table 1 summarises all

candidate variables that were excluded by stepwise backward deletion. Interestingly, preadmission functional status as assessed by Barthel Index and the number of drugs previously assumed were excluded from the model.

Variables included in the model (table 4) were used to calculate the mortality risk score intended for rapid patient's risk assessment on hospital admission. In this regard, age, number of comorbidities, respiratory rate, $\text{PaO}_2/\text{FiO}_2$, serum creatinine and PLT count reclassified into tertiles were used to build the mortality risk score with identification of three risk strata as reported in table 5. ROC analysis performed on the clinical risk score yielded an Area Under the Curve (AUC) of 0.90 (online supplemental figure 1, 95% CI 0.87 to 0.93). Kaplan-Meier survival analysis developed using the tertiles of the clinical score showed an excellent stratification of risk (figure 1; intermediate-risk vs low-risk HR: 4.134, 95% CI 1.725 to 9.905; high-risk vs low-risk HR: 22.173, 95% CI 9.681 to 50.783, $p<0.001$). A cut-off score of ≤ 8 identified a subset of 63 (12.2%) patients without fatalities during the study period, which therefore may be defined as 'at very low risk'.

Table 4 Cox multivariable regression analyses of determinants of in-hospital mortality

Variables	HR	95.0% CI	P value
Age (tertiles)			
62–74 versus <62 years	2.86	1.23 to 6.64	0.014
≥ 75 versus <62 years	7.92	3.60 to 17.43	<0.001
Number of comorbidities (tertiles)			
2–3 versus 0–1	1.85	1.11 to 3.08	0.018
≥ 4 versus 0–1	2.09	1.23 to 3.55	0.007
Respiratory rate (breaths/min), for unit increase	1.04	1.02 to 1.07	0.001
$\text{PaO}_2/\text{FiO}_2$, for unit increase	0.995	0.992 to 0.997	<0.001
Creatinine (mg/dL), for unit increase	1.34	1.18 to 1.51	<0.001
Platelets ($10^9/\text{L}$), for unit increase	0.995	0.992 to 0.998	0.001

History of cardiovascular disease, hypertension, diabetes, depression, dementia and cancer were included into 'comorbidities'. Variables excluded ($p>0.10$) from both models: No. of drugs, Barthel Index, C reactive protein.

DISCUSSION

In this study, we developed the COVID-19MRS that was shown to be able to stratify the risk of in-hospital death in COVID-19 patients since their admission. This score includes a composite of six objective, operator-independent variables (age, number of chronic comorbidities, respiratory rate, $\text{PaO}_2/\text{FiO}_2$, serum creatinine and platelet count) usually available within a couple of hours after hospitalisation. The score identified three categories at increasing risk of death with a high level of accuracy. The scoring process suggests that, while low-risk patients may be assigned safely to low-intensity care, higher intensity wards should be alerted during triage for the intermediate-risk and high-risk patients. Moreover, the score seems to allow for the identification of about

Table 5 Variables and relative scores to calculate the COVID-19 clinical risk score

Age (years)	Score	Comorbidities (N)	Score	RR (breaths/min)	Score	PaO ₂ /FiO ₂	Score	Creatinine (mg/dL)	Score	Platelet count (10 ⁹ /L)	Score	Risk categories (sum of individual variable scores)
<62	1	≤1	1	≤20	1	>300	1	<0.83	1	>212	1	Low ≤10
62–74	2	2–3	2	21–24	2	236–299	2	0.83–1.12	2	156–211	2	Intermediate=11–13
≥75	3	≥4	3	≥25	3	<236	3	≥1.13	3	<156	3	High risk ≥14

Categories represent the tertile distribution of each variable.

10%–15% of ‘very low-risk’ patients (score ≤8) with no events who, though symptomatic for proven COVID-19, might be immediately discharged home, with the sole indication to health status monitoring.

Performance of prognostic estimates for COVID-19 is under scrutiny as thought to be over optimistic and misleading, because of the high risk of bias in patient selection.¹⁰ As a case in point, a score based on a large cohort of COVID-19 patients in China found that age was associated with greater risk of death.¹¹ However, the mean age of this cohort was 49±16 years, which is 15–20 years less than observed in most European and US studies published to date. Although apparently similar in terms of objectives, we stratified the risk of death in a consecutive cohort of patients who shared demographic and clinical characteristics similar to other European and US studies.^{3–5} We, therefore, believe that our COVID-19MRS may hold potential generalisability for other countries. The early identification of patients at risk of clinical deterioration and death is of primary importance, considering that median interval from hospital admission to the Intensive Care Unit (ICU) is around 3 days¹⁶. Given that our proposed score is predictive of mortality based on six inexpensive, operator-independent and rapidly obtainable parameters, it could help clinicians to identify high-risk patients with poor prognosis since the triage phase.

One in four patients in our cohort of Italian COVID-19 cases died and age was the strongest driver of an adverse outcome. In fact, compared with patients younger than 62 years of age, the risk of death was almost three and eight times higher in individuals 62–74 and 75+ years of age, respectively. Such an exponential risk growth persisted after adjusting for burden of comorbidities and a series of clinical characteristics. Such a strong association between older age and prognosis has been observed in previous studies on COVID-19 both in China and in other countries, although with a less rapid increase in age-specific risk.¹⁷ This difference could be attributed to the lower median age reported in those studies and to the fact that we explored a wider age range (21–95 years), with one-third of our population above the age of 75.^{17 18} In COVID-19, age has been associated with variable degrees of increasing risk of admissions to ICU, onset of acute respiratory distress syndrome, myocardial damage and fatal outcome.^{16 19–22} This observation also holds true for previous epidemic or pandemic outbreaks, such as SARS and Middle East respiratory syndrome where, as in COVID-19, the respiratory system is both the entry route and the main target of viral infection.^{23 24} We could argue that lung senescence, resulting in decreased elasticity, increased end-expiratory lung volume and disrupted alveolar integrity,²⁵ together with kidney senescence,²⁶

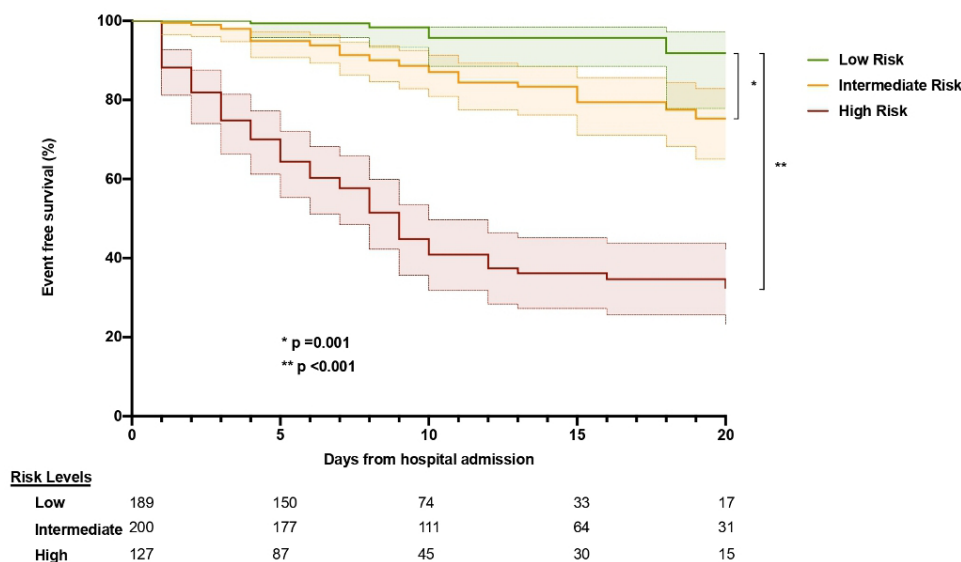


Figure 1 Kaplan-Meier analysis of overall survival of patients diagnosed with COVID-19 according to three risk categories. Shaded areas represent lower and upper 95% CIs.

may predispose per se to SARS-CoV-2-related acute respiratory and renal failure even in otherwise relatively robust elderly individuals. This hypothesis is consistent with the observation that age and three functional indicators of target organs (respiratory rate, PaO₂/FiO₂ and serum creatinine) emerged as independent predictors of in-hospital mortality, after adjusting for comorbidities.

The observation of the highly negative impact of age suggests that in the absence of specifically effective drug therapy and vaccination,²⁷ social isolation and the prevention of infecting contacts are key issues particularly relevant in individuals aged 70–75 years and over. These data may represent a call to action for health authorities, in order to update management policies in the community in general and in nursing homes in particular, where in fact the highest mortality rates occurred in Italy and in other countries.²⁸

While a low PLT count was frequently observed in non-survivor COVID-19 patients,^{16 19 22 29} in our cohort, lower values were directly associated with adverse outcome, suggesting a possible role of COVID-19-related coagulopathy in determining a poor outcome.^{30 31}

Present therapeutic recommendations on COVID-19 have a limited level of evidence,³² and have evolved during progression of the pandemic wave. Most of our patients received oxygen or mechanical ventilation support and antibiotics; conversely one in two patients were treated by antiviral and/or anti-inflammatory drugs. Given the nature of our study, we are unable to draw any firm conclusions regarding treatment efficacy, as specific analyses would be required, which were beyond the scope of the present work.

Some limitations of our study have to be acknowledged. First, the retrospective and observational nature of our analysis does not allow us to draw any firm conclusions about therapeutic strategies. Second, some laboratory parameters, which proved to be of prognostic relevance in other studies,^{19 22} were not collected for all individuals in our sample, possibly due to the different degrees of severity of patients (ie, very mildly affected vs critically ill patients at presentation). Therefore, we cannot rule out that variables excluded from the scoring system would have had a significant impact on mortality prediction. However, consistent with our purpose, we considered variables only available soon after admission. Third, since nasopharyngeal swabs were our key criterion for SARS-CoV-2 detection, we did not assess viremia, while the correlation of viral load with disease severity is still a matter of debate. Moreover, case ascertainment methodological bias, which may impact on patient selection and outcome, cannot be excluded as partial explanation for the findings observed. Indeed, the vast majority of patients included in the present analysis had a positive RT-PCR on first testing and only in a minority of cases was sputum or bronchoalveolar lavage needed to confirm the infection. Fourth, 82 (15.9%) out of 516 patients were still in-hospital at the time of closure of follow-up. Nevertheless, after excluding these patients from our analysis,

results were fully confirmed, with a 0.90 AUC of the predictive score (data not shown). Finally, we do not have information regarding the time span between symptom onset and admission, which might have had an impact on either clinical or laboratory parameters that we sampled on hospital admission.

In conclusion, we developed a scoring system (COVID-19MRS) that objectively and accurately predicts in-hospital mortality COVID-19 patients. This score, simply based on age, number of chronic comorbidities, respiratory rate, PaO₂/FiO₂, serum creatinine and platelet count, is a rapid and inexpensive clinical tool, which could be helpful for earlier identification of in-hospital mortality risk and, hence, assignment to the appropriate level of care and treatment of COVID-19 patients. Studies in clinical series different from ours are needed to validate the present scoring system.

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