



Predicting Risk Score for Mechanical Ventilation in Hospitalized Adult Patients Suffering from COVID-19

Samira Kafan¹, Kiana Tadbir Vajargah^{2,*}, Mehrdad Sheikhvatan^{3,4}, Gholamreza Tabrizi⁴, Ahmad Salimzadeh^{5,6}, Mahnaz Montazeri⁷, Fazeleh Majidi⁴, Negin Maghuli⁶ and Marzieh Pazoki^{1,**}

¹Department of Pulmonary Medicine, Sina Hospital, Tehran University of Medical Sciences, Tehran, Iran

²Student's Scientific Research Center (SSRC), Tehran University of Medical Sciences, Tehran, Iran

³Heidelberg Medical Hospital, Heidelberg, Germany

⁴Research Development Center, Sina Hospital, Tehran University of Medical Sciences, Tehran, Iran

⁵Rheumatology Research Center, Sina Hospital, Tehran University of Medical Science, Tehran, Iran

⁶Department of Internal Medicine, Sina Hospital, Tehran University of Medical Sciences, Tehran, Iran

⁷Department of Infectious Diseases, Sina Hospital, Tehran University of Medical Sciences, Tehran, Iran

*Corresponding author: Student's Scientific Research Center (SSRC), Tehran University of Medical Sciences, Tehran, Iran. Email: k-tadbir@students.tums.ac.ir

**Corresponding author: Department of Pulmonary Medicine, Sina Hospital, Tehran University of Medical Sciences, Tehran, Iran. Email: pazokimr@tums.ac.ir

Received 2021 January 15; Revised 2021 March 08; Accepted 2021 March 12.

Abstract

Background: COVID-19 has become a pandemic since December 2019, causing millions of deaths worldwide. It has a wide spectrum of severity, ranging from mild infection to severe illness requiring mechanical ventilation. In the middle of a pandemic, when medical resources (including mechanical ventilators) are scarce, there should be a scoring system to provide the clinicians with the information needed for clinical decision-making and resource allocation.

Objectives: This study aimed to develop a scoring system based on the data obtained on admission, to predict the need for mechanical ventilation in COVID-19 patients.

Methods: This study included COVID-19 patients admitted to Sina Hospital, Tehran University of Medical Sciences from February 20 to May 29, 2020. Patients' data on admission were retrospectively recruited from Sina Hospital COVID-19 Registry (SHCo-19R). Multivariable logistic regression and receiver operating characteristic (ROC) curve analysis were performed to identify the predictive factors for mechanical ventilation.

Results: A total of 681 patients were included in the study; 74 patients (10.9%) needed mechanical ventilation during hospitalization, while 607 (89.1%) did not. Multivariate logistic analysis revealed that age (OR,1.049; 95% CI:1.008-1.091), history of diabetes mellitus (OR,3.216; 95% CI:1.134-9.120), respiratory rate (OR,1.051; 95% CI:1.005-1.100), oxygen saturation (OR,0.928; 95% CI:0.872-0.989), CRP (OR,1.013; 95% CI:1.001-1.024) and bicarbonate level (OR,0.886; 95% CI:0.790-0.995) were risk factors for mechanical ventilation during hospitalization.

Conclusions: A risk score has been developed based on the available data within the first hours of hospital admission to predict the need for mechanical ventilation. This risk score should be further validated to determine its applicability in other populations.

Keywords: Coronavirus Disease, COVID-19, Mechanical Ventilation, Risk Score

1. Background

Coronavirus disease 2019 (COVID-19), the disease caused by a strain of the coronavirus family, was first identified in Wuhan, China, in December 2019. It rapidly spread to most countries all over the world. There have been more than 116 million laboratory-confirmed cases and more than 2.5 million deaths caused by the COVID-19 pandemic (1).

COVID-19 is a multisystem inflammatory disease and has a wide spectrum of severity, ranging from mild to severe and critical illness. Patients may experience fever,

cough, headache, diarrhea, or serious complications such as cytokine storm, acute respiratory distress syndrome, respiratory failure, sepsis, etc. (2-6). In severe cases when respiratory distress is not sufficiently treated by oxygen alone, treatment may proceed to mechanical ventilation in patients experiencing hypoxemia or increased work of breathing (7-9).

According to a review article, 14 - 32% of infected patients develop a critical illness, and 29-89% of patients with COVID-19 require mechanical ventilation support (10). Although critically ill patients with COVID-19 are not the majority of cases, they require more sophisticated care and

are at higher risk of death (11). Moreover, in the middle of a pandemic, health care systems face various challenges of which limited availability of resources is important because this limitation can impose a lot of stress on health-care workers (12).

Early detection of factors that can predict the need for mechanical ventilation can be helpful in several respects; it gives a better understanding of the situation and helps with better management and resource allocation. It also underlines the patients likely to develop unfavorable outcomes and provides opportunities for early interventions in this high-risk group of patients.

2. Objectives

We aimed to identify the factors which can be early predictors of the need for mechanical ventilation in the course of the disease and develop a multivariable model to predict the need for mechanical ventilation during hospitalization.

3. Methods

In this single-center retrospective cohort study, we enrolled 681 patients with COVID-19 admitted to Sina Hospital, Tehran University of Medical Sciences, from February 20 to May 29, 2020. The protocol of this study complied with the Declaration of Helsinki was approved by the Tehran University of medical sciences (IR.TUMS.VCR.REC.1399.005). All included patients were over 18 years of age with suspected COVID-19 symptoms. The diagnosis of COVID-19 was based on a positive result using Real-time polymerase-chain-reaction (RT-PCR) detection for a routine oropharyngeal swab or endotracheal sample specimens. We also enrolled some patients according to the WHO's interim guidance and Iranian national committee of COVID-19 into our study (13, 14), including patients with ground-glass opacity (GGO) alone or GGO accompanied by consolidation in chest computed tomography (CT), along with the history compatible with COVID-19. The algorithm of patient care for individuals presenting with respiratory symptoms to Sina Hospital emergency department has been published before (15).

Patients' demographics, clinical information (history and physical examination), laboratory values, radiologic findings, and outcomes regarding mechanical ventilation were recruited from Sina Hospital COVID-19 Registry (SHCo-19R) (15).

3.1. Study Outcomes

The outcome in this study was the need for mechanical ventilation, which was decided upon after consultation with respiratory specialists on an individualized basis for each patient.

3.2. Statistical Analysis

We expressed descriptive data as median (Inter-Quartile Range) for continuous variables and number (percentage) for categorical variables. Means of continuous variables were compared using independent group *t*-tests. Proportions for categorical variables were compared using the chi-square test. Univariable and multivariable analyses to identify factors associated with the need for mechanical ventilation from COVID-19 were performed by the multivariable logistic regression modeling. Considering the total number of mechanical ventilation ($n = 74$) in our study, variables were chosen for the multivariate model based on univariable analysis ($P < 0.05$). The odds ratio (OR) along with the 95% CI were reported. In multivariable analysis, only six variables were statistically significant that were included in the study model. Other significant variables in univariate analysis were excluded due to collinearity or lack of sufficient data in most cases. For risk stratifying and developing a risk score, the final logistic regression model was used as the method previously described by Sullivan et al. (16). In this regard, all study variables that were significant in the multivariable logistic regression model were considered categorical binary variables (even continuous variables based on the best cut-off value for each variable with the optimal sensitivity and specificity by analyzing the area under the receiver operating characteristic (ROC) curve). For all variables structured as the categorical pattern, the distance between a variable and its base (reference) category in regression coefficient units was equal to the size of the coefficient. To obtain final scores for each risk factor, we needed the constant of the scoring system that corresponded to one point in the risk score system. The point value for each risk factor was calculated by dividing the distance of each risk factor from the base category in regression coefficient units by this constant. The total risk score was obtained by adding up the points for all significant risk factors. The predicted risk associated with each risk score was recalibrated by multiplying it by the ratio of the observed mechanical ventilation rate and was finally rescored as low, moderate, and high risk according to the quartile of the percentage of predicted risk. All statistical analyses were performed using SPSS version 22.0 software (IBM, Armonk, New York), and $P < 0.05$ was considered statistically significant.

4. Results

A total of 681 confirmed patients with COVID-19 were included in this study. Baseline characteristics and laboratory findings on admission are presented in appendix 1 and 2, respectively. Of the total number of patients, 74 (10.9%) patients needed mechanical ventilation during

hospitalization, whereas 607 patients (89.1%) did not. Patients who required mechanical ventilation were significantly older than patients who did not. The median age in the mechanical ventilation and non-mechanical ventilation group were 68.58 ± 14.6 and 56.06 ± 16 years, respectively.

Mechanical ventilation and non-mechanical ventilation groups were significantly different considering respiratory rate on admission, arterial oxygen saturation on admission, white blood cell count, red cell distribution width (RDW), urea, potassium, calcium, phosphorus, magnesium, lactate dehydrogenase, erythrocyte sedimentation rate, C-reactive protein, creatine phosphokinase, aspartate aminotransferase, alanine aminotransferase, ferritin, troponin, and bicarbonate.

Also, there was a significant difference between the two groups considering the history of hypertension, diabetes mellitus, cardiovascular disease, malignancy, lung disease, and cerebrovascular accident. Moreover, patients requiring mechanical ventilation more commonly experienced myalgia and were taking metformin, beta-blockers, and aspirin in comparison to the non-mechanical ventilation group. In contrast, the difference between mechanical ventilation and non-mechanical ventilation groups was not significant in terms of gender, history of transplantation, chronic kidney disease, and CT scan findings.

Multivariate logistic analysis revealed that age (OR, 1.049; 95% CI: 1.008-1.091; $P = 0.02$), history of diabetes mellitus (OR, 3.216; 95% CI: 1.134 - 9.120; $P = 0.028$), respiratory rate on admission (OR, 1.051; 95% CI: 1.005 - 1.100; $P = 0.031$), oxygen saturation on admission (OR, 0.928; 95% CI: 0.872 - 0.989; $P = 0.021$), CRP (OR, 1.013; 95% CI: 1.001 - 1.024; $P = 0.032$) and HCO_3^- level (OR, 0.886; 95% CI: 0.790 - 0.995; $P = 0.04$) were risk factors for mechanical ventilation during hospitalization (Table 1).

To categorize five continuous variables of age, CRP, HCO_3^- level, respiratory rate, and oxygen saturation, the area under ROC curve (AUC) was analyzed, yielding the best cut-off values of 55 years for age, 18/min for respiratory rate, 22 mEq/L for HCO_3^- level, 65 mg/L for CRP, and oxygen saturation less than 85% (Figure 1). The results of ROC analyses showing $\text{AUC} \geq 0.650$ are outlined in Table 2.

4.1. Construction of a Scoring Model

The risk score included the following parameters on admission: age > 55, a history of diabetes mellitus, a respiratory rate > 18/minute, oxygen saturation < 85%, CRP > 65 mg/L, and $\text{HCO}_3^- < 22$ mEq/L. The point scores for the obtained risk parameters in the logistic regression analysis are summarized in Table 3.

The total risk score was calculated to be 21, and each risk parameter gets one point if it is not presented to the patient. There are six risk parameters. Thus, the minimum

risk score will be 6 for individuals with no pointed risk factors and 21 for individuals older than 55 years, with a positive history of diabetes, with the symptom of tachypnea and decreased HCO_3^- level and raised CRP on admission. Finally, a total risk score of less than 10, between 10 to 15, and higher than 15 was considered low, moderate, and high risk for mechanical ventilation due to COVID-19, respectively.

5. Discussion

This single-center study analyzed demographic, clinical, laboratory, and radiological characteristics of patients with COVID-19 admitted to Sina Hospital, Tehran, to retrospectively develop a risk score model for predicting the need for mechanical ventilation during hospitalization.

In this retrospective cohort study, 10.9% of patients required mechanical ventilation during hospitalization, while 89.1% did not. We found that patients requiring mechanical ventilation were of advanced age, had more comorbidities (including diabetes mellitus, hypertension, lung diseases, cardiovascular diseases, cerebrovascular accident, and malignancies), and more clinical and laboratory abnormalities.

In the midst of the COVID-19 pandemic, medical resources, including mechanical ventilators, face a significant shortage. According to the challenge of resource scarcity, it is crucial to have a model upon which decisions about resource allocation can be made. Developing a hospital-based risk score can provide clinicians with a valuable tool to stratify the risk of requiring mechanical ventilation during hospitalization.

Using six variables that are either available at hospital admission or can be obtained within the first hours of admission, we developed a clinical risk score to estimate the patient's risk of requiring mechanical ventilation during hospitalization. Age, history of diabetes mellitus, respiratory rate, oxygen saturation, hs-CRP, and bicarbonate level were included in the risk score.

Our findings were compatible with the previous studies, which have shown that advanced age and comorbidities can put the patients at higher risk for developing severe illness and poor prognosis. (2, 17-19) Among comorbidities, diabetes mellitus happened to have the most powerful correlation with the need for mechanical ventilation in our study, which was consistent with previous studies. (20-23)

According to the statistical analysis of the data on admission, we found that fasting blood glucose and HbA1c were not significantly different in mechanical ventilation and non-mechanical ventilation groups. Based on these findings, we hypothesize that the diabetes mellitus itself, rather than whether the blood glucose is controlled, can

Table 1. Multivariable Logistic Regression Model for Predicting the Need for Mechanical Ventilation

	Coefficient	P-Value (Multivariate)	Odds Ratio (OR)	95%CI for OR	
				Lower	Upper
Age	0.047	0.020 ^a	1.049	1.008	1.091
Gender	- 0.495	0.379	0.610	0.202	1.838
Hypertension	- 0.223	0.698	0.800	0.259	2.468
Diabetes mellitus	1.168	0.028 ^a	3.216	1.134	9.120
Cardiac disease	- 0.998	0.098	0.369	0.113	1.200
Lung disease	- 0.885	0.242	0.413	0.094	1.819
Respiratory Rate	0.050	0.031 ^a	1.051	1.005	1.100
SpO ₂	- 0.074	0.021 ^a	0.928	0.872	0.989
W.B.C	0.045	0.421	1.046	0.937	1.168
Potassium	0.081	0.853	1.085	0.457	2.575
ESR	- 0.014	0.147	0.986	0.968	1.005
CRP	0.013	0.032 ^a	1.013	1.001	1.024
HCO ₃ ⁻	- 0.121	0.040 ^a	0.886	0.790	0.995
Constant	3.949	0.320	51.861		

Abbreviations: SpO₂, peripheral capillary oxygen saturation; WBC, white blood count; ESR, erythrocyte sedimentation rate; CRP, C-reactive protein; HCO₃⁻, bicarbonate.

^a Statistically significant P-values.

Table 2. Results of ROC Analysis with Variables Showing AUC \geq 0.65 for Mechanical Ventilation

Parameter	AUC	Lower Limit	Upper Limit	P-Value	Cut-Off	Sensitivity	Specificity
Age	0.670	0.587	0.753	< 0.001	55	80.4%	53.3%
RR	0.660	0.565	0.755	< 0.001	18	76.1%	65.7%
SpO ₂	0.711	0.623	0.799	< 0.001	85	78.2%	60.4%
CRP	0.726	0.653	0.799	< 0.001	65	76.1%	59.8%
HCO ₃ ⁻	0.672	0.586	0.757	< 0.001	22	77.9%	50.0%

Abbreviations: RR, respiratory rate; SpO₂, peripheral capillary oxygen saturation; CRP, c-reactive protein; HCO₃⁻, bicarbonate; AUC, area under ROC curve.

affect the course of the disease. Further studies are warranted to test this hypothesis.

In line with previous studies, we found higher CRP and respiratory rate, and lower oxygen saturation are associated with poor clinical outcomes regarding the need for mechanical ventilation. In fact, CRP is an acute-phase reactant and a marker of inflammation that has been associated with disease progression in MERS, H1N1 influenza, and recently COVID-19. (18, 24-30).

To the best of our knowledge, the relationship of bicarbonate and disease outcome regarding mechanical ventilation in patients with COVID-19 has not been thoroughly studied based on the review of recent literature. In our study, patients who required mechanical ventilation had significantly lower bicarbonate levels on admission. Given that the pH on admission was not significantly different in mechanical ventilation and non-mechanical ventilation

groups, we think that the precise amount of serum bicarbonate can be valuable in clinical settings.

Using the risk score, the clinicians can easily categorize patients into low-, moderate-, and high-risk groups. Considering that the risk score is not the only tool in decision making, it can be used to expedite and optimize decisions in the management of patients with COVID-19.

5.1. Conclusions

This study identified the clinical factors that predict the need for mechanical ventilation in adult patients with COVID-19. Based on the findings, as mentioned earlier, we developed a risk score to stratify the risk and predict the need for mechanical ventilation in hospitalized patients with COVID-19. This model, including age, history of diabetes mellitus, respiratory rate, oxygen saturation, hs-CRP, and bicarbonate, can provide the clinicians with an

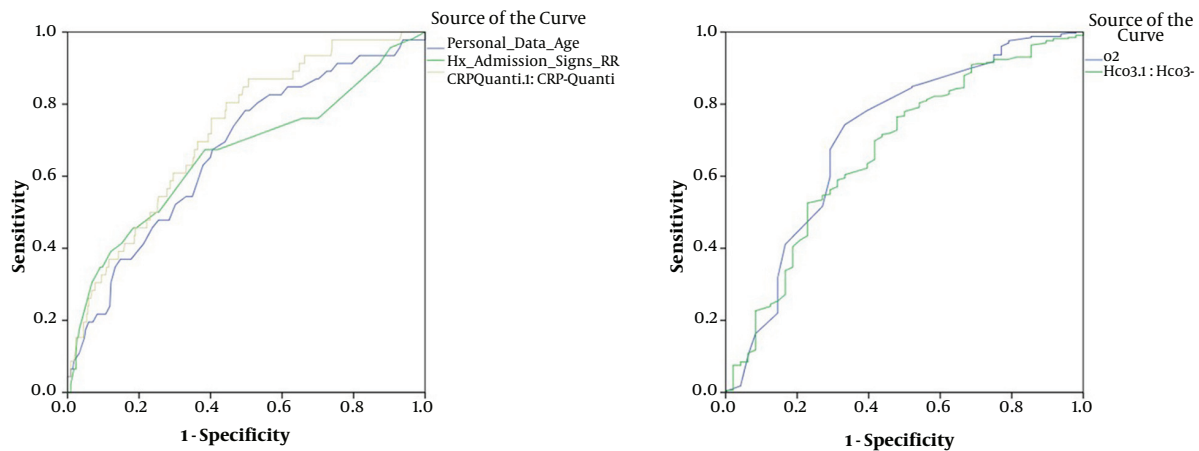


Figure 1. ROC curves of independent risk factors for mechanical ventilation in patients with COVID-19.

Table 3. Risk Scores for Mechanical Ventilation for COVID-19

Parameter	Risk Score
History of diabetes mellitus	
Positive	2
Negative	1
Decreased oxygen saturation	
< 85%	3
≥ 85%	1
Age, year	
> 55	4
≤ 55	1
Respiratory rate, per minute	
> 18	4
≤ 18	1
Raised CRP on admission, mg/L	
> 65	6
≤ 65	1
HCO₃⁻, mEq/L	
< 22	2
≥ 22	1
Total risk score	21

evidence-based tool that can facilitate and support their decision-making in managing patients with COVID-19.

5.2. Limitation

There were several limitations to this study. This single-center study had a modest sample size. Since the data used for developing the risk score were solely obtained from one

country, the results' generalizability may be potentially limited. A prospective study seems necessary to validate and confirm the reliability of the risk score.

Acknowledgments

We acknowledge all health-care workers involved in the diagnosis and treatment of patients in Sina Hospital. We are indebted to the Research Development Center of Sina Hospital for its support and members of the COVID-19 Crisis Management Committee of the Sina Hospital for their help and consult.

Footnotes

Authors' Contribution: Study concept and design: K.T., M.S., G.T., M.P., M.M., F.M., N.M., A.S., and S.K.; Acquisition of data: F.M., N.M., S.K., and K.T.; analysis and interpretation of data: M.S., S.K., and K.T.; drafting of the manuscript: K.T., M.S., G.T., M.P., M.M., F.M., N.M., A.S., and S.K.; critical revision of the manuscript for important intellectual content: K.T., M.S., G.T., M.P., M.M., F.M., N.M., A.S., and S.K.; statistical analysis: M.S.

Conflict of Interests: The authors declare that there are no conflicts of interest.

Ethical Approval: IR.TUMS.VCR.REC.1399.005

Funding/Support: This work is supported by the Tehran University of Medical Sciences [grant number: 99-1-101-47193]. This study's funding sources had no role in the study design, data collection, data analysis, data interpretation, or writing of the manuscript.

References

1. WHO. *Coronavirus disease (COVID-19) dashboard | WHO coronavirus disease (COVID-19) dashboard*. [cited 2020 December 22]. Available from: <https://covid19.who.int/>.
2. Zhou F, Yu T, Du R, Fan G, Liu Y, Liu Z, et al. Clinical course and risk factors for mortality of adult inpatients with COVID-19 in Wuhan, China: a retrospective cohort study. *Lancet*. 2020;**395**(10229):1054–62. doi: [10.1016/S0140-6736\(20\)30566-3](https://doi.org/10.1016/S0140-6736(20)30566-3).
3. WHO. *Coronavirus*. 2020, [cited 2020 August 28]. Available from: https://www.who.int/health-topics/coronavirus#tab=tab_3.
4. Roshanravan N, Seif F, Ostadrahimi A, Pouraghaei M, Ghafari S. Targeting cytokine storm to manage patients with COVID-19: A mini-review. *Arch Med Res*. 2020;**51**(7):608–12. doi: [10.1016/j.arcmed.2020.06.012](https://doi.org/10.1016/j.arcmed.2020.06.012). [PubMed: [32682575](https://pubmed.ncbi.nlm.nih.gov/32682575/)]. [PubMed Central: [PMC7303639](https://pubmed.ncbi.nlm.nih.gov/PMC7303639/)].
5. Shadvar K, Tagizadiyeh A, Gamari AA, Soleimanpour H, Mahmoodpoor A. Hemoperfusion as a potential treatment for critically ill COVID-19 patients with cytokine storm. *Blood Purif*. 2020;1–3. doi: [10.1159/000511391](https://doi.org/10.1159/000511391). [PubMed: [33171470](https://pubmed.ncbi.nlm.nih.gov/33171470/)].
6. Hassani V, Amniati S, Ahmadi A, Mohseni M, Sehat-Kashani S, Nikoubakht N, et al. Emergency tracheostomy in two airway trauma patients suspected of COVID-19: A case report. *Anesth Pain Med*. 2020;**10**(4). e104648. doi: [10.5812/aapm.104648](https://doi.org/10.5812/aapm.104648). [PubMed: [33134149](https://pubmed.ncbi.nlm.nih.gov/33134149/)]. [PubMed Central: [PMC7539045](https://pubmed.ncbi.nlm.nih.gov/PMC7539045/)].
7. World Health Organization. *Clinical management of COVID-19: interim guidance*. 2020, [cited 2020 August 28]. Available from: <https://apps.who.int/iris/handle/10665/332196>.
8. Pergolizzi JJ, Magnusson P, LeQuang JA, Breve F, Paladini A, Rekasina M, et al. The current clinically relevant findings on COVID-19 pandemic. *Anesth Pain Med*. 2020;**10**(2). e103819. doi: [10.5812/aapm.103819](https://doi.org/10.5812/aapm.103819). [PubMed: [32754437](https://pubmed.ncbi.nlm.nih.gov/32754437/)]. [PubMed Central: [PMC7352949](https://pubmed.ncbi.nlm.nih.gov/PMC7352949/)].
9. Mahmoodpoor A, Shadvar K, Ghamari AA, Mohammadzadeh Lameh M, Asghari Ardebili R, Hamidi M, et al. Management of critically ill patients with COVID-19: What we learned and what we do. *Anesth Pain Med*. 2020;**10**(3). doi: [10.5812/aapm.104900](https://doi.org/10.5812/aapm.104900).
10. Liu J, Xie W, Wang Y, Xiong Y, Chen S, Han J, et al. A comparative overview of COVID-19, MERS and SARS: Review article. *Int J Surg*. 2020;**81**:1–8. doi: [10.1016/j.ijvs.2020.07.032](https://doi.org/10.1016/j.ijvs.2020.07.032). [PubMed: [32730205](https://pubmed.ncbi.nlm.nih.gov/32730205/)]. [PubMed Central: [PMC7382925](https://pubmed.ncbi.nlm.nih.gov/PMC7382925/)].
11. Rahimzadeh P, Amniati S, Faiz SHR, Farahmandrad R, Emami SH. Clinical characteristics of critically ill patients infected with COVID-19 in rasoul akram hospital in Iran: A single center study. *Anesth Pain Med*. 2020;**10**(5):1–7.
12. Ali H, Ismail AA, Abdalwahab A. Mental stress in anesthesia and intensive care physicians during COVID-19 outbreak. *Anesth Pain Med*. 2020;**10**(5):1–6. doi: [10.5812/aapm.100673](https://doi.org/10.5812/aapm.100673).
13. WHO. *WHO Director-General's opening remarks at the media briefing on COVID-19*. 2020, [cited 2020 September 4]. Available from: <https://www.who.int/dg/speeches/detail/who-director-general-s-opening-remarks-at-the-media-briefing-on-covid-19---11-march-2020>.
14. Xu Z, Shi L, Wang Y, Zhang J, Huang L, Zhang C, et al. Pathological findings of COVID-19 associated with acute respiratory distress syndrome. *Lancet Respir Med*. 2020;**8**(4):420–2. doi: [10.1016/S2213-2600\(20\)30076-X](https://doi.org/10.1016/S2213-2600(20)30076-X).
15. Talebpour M, Hadadi A, Oraili A, Ashraf H. Iran: Sina hospital Covid-19 registry (SHCo-19R). *Ref Educ Med Cent Tehran*. 2020;**4**:1–5.
16. Sullivan LM, Massaro JM, D'Agostino RS. Presentation of multivariate data for clinical use: The Framingham Study risk score functions. *Stat Med*. 2004;**23**(10):1631–60. doi: [10.1002/sim.1742](https://doi.org/10.1002/sim.1742). [PubMed: [15122742](https://pubmed.ncbi.nlm.nih.gov/15122742/)].
17. Guan WJ, Liang WH, Zhao Y, Liang H, Chen ZS, Li YM, et al. Comorbidity and its impact on 1,590 patients with Covid-19 in China: A nationwide analysis. *Eur Respir J*. 2020;**55**(5).
18. Hou W, Zhang W, Jin R, Liang L, Xu B, Hu Z. Risk factors for disease progression in hospitalized patients with COVID-19: a retrospective cohort study. *Infect Dis (Lond)*. 2020;**52**(7):498–505. doi: [10.1080/23744235.2020.1759817](https://doi.org/10.1080/23744235.2020.1759817). [PubMed: [32370577](https://pubmed.ncbi.nlm.nih.gov/32370577/)]. [PubMed Central: [PMC7212540](https://pubmed.ncbi.nlm.nih.gov/PMC7212540/)].
19. Labenz C, Kremer WM, Schattenberg JM, Worns MA, Toenges G, Weimann A, et al. Clinical frailty scale for risk stratification in patients with SARS-CoV-2 infection. *J Investig Med*. 2020;**68**(6):1199–202. doi: [10.1136/jim-2020-001410](https://doi.org/10.1136/jim-2020-001410). [PubMed: [32641351](https://pubmed.ncbi.nlm.nih.gov/32641351/)]. [PubMed Central: [PMC7418620](https://pubmed.ncbi.nlm.nih.gov/PMC7418620/)].
20. Yan Y, Yang Y, Wang F, Ren H, Zhang S, Shi X, et al. Clinical characteristics and outcomes of patients with severe covid-19 with diabetes. *BMJ Open Diabetes Res Care*. 2020;**8**(1). doi: [10.1136/bmjdr-2020-001343](https://doi.org/10.1136/bmjdr-2020-001343). [PubMed: [32345579](https://pubmed.ncbi.nlm.nih.gov/32345579/)]. [PubMed Central: [PMC7222577](https://pubmed.ncbi.nlm.nih.gov/PMC7222577/)].
21. Hussain A, Bhowmik B, do Vale Moreira NC. COVID-19 and diabetes: Knowledge in progress. *Diabetes Res Clin Pract*. 2020;**162**:108142. doi: [10.1016/j.diabres.2020.108142](https://doi.org/10.1016/j.diabres.2020.108142). [PubMed: [32278764](https://pubmed.ncbi.nlm.nih.gov/32278764/)]. [PubMed Central: [PMC7144611](https://pubmed.ncbi.nlm.nih.gov/PMC7144611/)].
22. Guo W, Li M, Dong Y, Zhou H, Zhang Z, Tian C, et al. Diabetes is a risk factor for the progression and prognosis of COVID-19. *Diabetes Metab Res Rev*. 2020. e3319. doi: [10.1002/dmrr.3319](https://doi.org/10.1002/dmrr.3319). [PubMed: [32233013](https://pubmed.ncbi.nlm.nih.gov/32233013/)]. [PubMed Central: [PMC7228407](https://pubmed.ncbi.nlm.nih.gov/PMC7228407/)].
23. Nicholson CJ, Wooster L, Sigurslid HH, Li RF, Jiang W, Tian W, et al. Estimating risk of mechanical ventilation and mortality among adult COVID-19 patients admitted to Mass General Brigham: The VICE and DICE scores. *medRxiv*. 2020. doi: [10.1101/2020.09.14.20194670](https://doi.org/10.1101/2020.09.14.20194670). [PubMed: [32995802](https://pubmed.ncbi.nlm.nih.gov/32995802/)]. [PubMed Central: [PMC7523141](https://pubmed.ncbi.nlm.nih.gov/PMC7523141/)].
24. Zhang J, Yu M, Tong S, Liu LY, Tang LV. Predictive factors for disease progression in hospitalized patients with coronavirus disease 2019 in Wuhan, China. *J Clin Virol*. 2020;**127**:104392. doi: [10.1016/j.jcv.2020.104392](https://doi.org/10.1016/j.jcv.2020.104392). [PubMed: [32361327](https://pubmed.ncbi.nlm.nih.gov/32361327/)]. [PubMed Central: [PMC7187844](https://pubmed.ncbi.nlm.nih.gov/PMC7187844/)].
25. Herold T, Jurinovic V, Arnreich C, Lipworth BJ, Hellmuth JC, von Bergwelt-Baildon M, et al. Elevated levels of IL-6 and CRP predict the need for mechanical ventilation in COVID-19. *J Allergy Clin Immunol*. 2020;**146**(1):128–136 e4. doi: [10.1016/j.jaci.2020.05.008](https://doi.org/10.1016/j.jaci.2020.05.008). [PubMed: [32425269](https://pubmed.ncbi.nlm.nih.gov/32425269/)]. [PubMed Central: [PMC7233239](https://pubmed.ncbi.nlm.nih.gov/PMC7233239/)].
26. Ko JH, Park GE, Lee JY, Lee JY, Cho SY, Ha YE, et al. Predictive factors for pneumonia development and progression to respiratory failure in MERS-CoV infected patients. *J Infect*. 2016;**73**(5):468–75. doi: [10.1016/j.jinf.2016.08.005](https://doi.org/10.1016/j.jinf.2016.08.005). [PubMed: [27519621](https://pubmed.ncbi.nlm.nih.gov/27519621/)]. [PubMed Central: [PMC7112644](https://pubmed.ncbi.nlm.nih.gov/PMC7112644/)].
27. Vasileva D, Badawi A. C-reactive protein as a biomarker of severe H1N1 influenza. *Inflamm Res*. 2019;**68**(1):39–46. doi: [10.1007/s00011-018-1188-x](https://doi.org/10.1007/s00011-018-1188-x). [PubMed: [30288556](https://pubmed.ncbi.nlm.nih.gov/30288556/)]. [PubMed Central: [PMC6314979](https://pubmed.ncbi.nlm.nih.gov/PMC6314979/)].
28. Tan C, Huang Y, Shi F, Tan K, Ma Q, Chen Y, et al. C-reactive protein correlates with computed tomographic findings and predicts severe COVID-19 early. *J Med Virol*. 2020;**92**(7):856–62. doi: [10.1002/jmv.25871](https://doi.org/10.1002/jmv.25871). [PubMed: [32281668](https://pubmed.ncbi.nlm.nih.gov/32281668/)]. [PubMed Central: [PMC7262341](https://pubmed.ncbi.nlm.nih.gov/PMC7262341/)].
29. Rubin SJ, Falkson SR, Degner NR, Blish C. Clinical characteristics associated with COVID-19 severity in California. *Journal of Clinical and Translational Science*. 2020;**5**(1). doi: [10.1017/cts.2020.40](https://doi.org/10.1017/cts.2020.40).
30. Rahimzadeh P, Faiz HR, Farahmandrad R, Hassanlouei B, Habibi A, Emami SH, et al. Clinical features and prognosis of invasive ventilation in hospitalized patients with COVID-19: A retrospective study. *Anesthesiol Pain Med*. *Anesth Pain Med*. 2020;**10**(6):1–8. doi: [10.5812/aapm.108773](https://doi.org/10.5812/aapm.108773).