Cumulative oxygen deficit is a novel predictor for the timing of invasive mechanical ventilation in COVID-19 patients with respiratory distress

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Background and objectives: The timing of invasive mechanical ventilation (IMV) is controversial in COVID-19 patients with acute respiratory hypoxemia. The study aimed to develop a novel predictor called cumulative oxygen deficit (COD) for the risk stratification.

Methods: The study was conducted in four designated hospitals for treating COVID-19 patients in Jingmen, Wuhan, from January to March 2020. COD was defined to account for both the magnitude and duration of hypoxemia. A higher value of COD indicated more oxygen deficit. The predictive performance of COD was calculated in multivariable Cox regression models. Results: A number of 111 patients including 80 in the non-IMV group and 31 in the IMV group were included. Patients with IMV had substantially lower PaO 2 (62 (49, 89) vs. 90.5 (68, 125.25) mmHg; p < 0.001), and higher COD (-6.87 (-29.36, 52.38) vs. -231.68 (-1040.78, 119.83) mmHg.day) than patients without IMV. As compared to patients with COD < 0, patients with COD > 30 mmHg.day had higher risk of fatality (HR: 3.79, 95% CI: 2.57 to 16.93; p = 0.037), and those with COD > 50 mmHg.day were 10 times more likely to die (HR: 10.45, 95% CI: 1.28 to 85.37; p = 0.029). Conclusions: The study developed a novel predictor COD which considered both magnitude and duration of hypoxemia, to assist risk stratification of COVID-19 patients with acute respiratory distress.
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Abstract

Background and objectives: The timing of invasive mechanical ventilation (IMV) is controversial in COVID-19 patients with acute respiratory hypoxemia. The study aimed to develop a novel predictor called cumulative oxygen deficit (COD) for the risk stratification.

Methods: The study was conducted in four designated hospitals for treating COVID-19 patients in Jingmen, Wuhan, from January to March 2020. COD was defined to account for both the magnitude and duration of hypoxemia. A higher value of COD indicated more oxygen deficit. The predictive performance of COD was calculated in multivariable Cox regression models.

Results: A number of 111 patients including 80 in the non-IMV group and 31 in the IMV group were included. Patients with IMV had substantially lower PaO₂ (62 (49, 89) vs. 90.5 (68, 125.25) mmHg; p < 0.001), and higher COD (-6.87 (-29.36, 52.38) vs. -231.68 (-1040.78, 119.83) mmHg · day) than patients without IMV. As compared to patients with COD < 0, patients with COD > 30 mmHg · day had higher risk of fatality (HR: 3.79, 95% CI: 2.57 to 16.93; p = 0.037), and those with COD > 50 mmHg · day were 10 times more likely to die (HR: 10.45, 95% CI: 1.28 to 85.37; p = 0.029).

Conclusions: The study developed a novel predictor COD which considered both magnitude and duration of hypoxemia, to assist risk stratification of COVID-19 patients with acute respiratory distress.
Introduction

Coronavirus disease 2019 (COVID-19) has spread all over the world since its first outbreak in Wuhan, China in December 2019 (Wang et al., 2020; Novel Coronavirus Pneumonia Emergency Response Epidemiology Team, 2020). The fatality rate was reported to be around 5% all over the world (Phua et al., 2020). A substantial number of patients (19%) infected with the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) will develop respiratory distress and acute lung injury (Wu & McGoogan, 2020; Ruan et al., 2020). Respiratory support becomes important for this type of severe patients (Yang et al., 2020). The surviving sepsis guideline of critically ill COVID-19 patients recommended use of oxygen supplementation to maintain pulse oximetry $>90\%$, followed by non-invasive mechanical ventilation (NIV), high-flow nasal canula (HFNC), invasive mechanical ventilation (IMV) and extracorporeal membrane oxygenation (ECMO).

However, there is no specific recommendations for the timing of transition from non-invasive support to IMV, and the recommendations are largely based on expert opinions. For example, the guideline recommends “close monitoring for worsening of respiratory status, and early intubation in a controlled setting if worsening occurs” (Alhazzani et al., 2020). This recommendation is based on best practice statement and there is no data on when IMV should be initiated. In clinical practice, the judgement of “worsening” is subjective and varied substantially between different institutions and physicians. The timing of initiation of IMV is not standardized and is mainly determined by subjective judgement. On the one hand, IMV is able to reverse catastrophic hypoxemia and maintain tissue oxygenation, which is life-saving for COVID-19 patients with severe hypoxemia. On the other hand, IMV can cause ventilator-induced lung injury (Herasevich et al., 2011; Cressoni et al., 2016), and patients on IMV usually require large dose of sedatives, analgesics and even neuromuscular blockades (Jakob et al., 2012; Bellani et
al., 2016; Chang et al., 2020). These drugs have significant adverse effects (Barr et al., 2013; Murray et al., 2016). Thus, it is difficult to determine the appropriate timing of IMV. In our experience, we proposed that the timing of transition from non-invasive oxygenation to IMV should consider both the magnitude of hypoxemia and the duration of the hypoxemia. Thus, we developed a novel marker called Cumulative Oxygen Deficit (COD) to reflect both dimensions. By using a single predictor, we reduce a two-dimension feature to a one-dimension parameter that is comparable among different patients. In our study, we hypothesized that the COD before IMV could be a better biomarker than PaO$_2$ to predict survival outcome.
Methods

Study design and setting

The study was conducted in four designated hospitals for treating COVID-19 patients in Jingmen, Wuhan, from January to March 2020. Medical records were retrospectively reviewed to identify eligible patients and variables. Laboratory tests and type of ventilation support were recorded as longitudinal data. The study was designed as a longitudinal study that all patients were followed until hospital discharge or death. One subject contributed several observation units. Patients were divided into groups with IMV and without IMV during hospitalization. The study was approved by the ethics committee of the First People’s hospital of Jingmen (Approval number: 202002007) and the ethics committee of Sir Run Run Shaw hospital (20200407-32). Individual patient data were de-identified before analysis. Informed consent was waived as determined by the IRB due to retrospective nature of the study design in accordance to the local regulations.

Study population

COVID-19 was confirmed by either 1) genetic sequencing showed highly homogenous sequence with the known novel coronavirus; or 2) novel coronavirus nucleic acid was positive as confirmed by real time (RT)-PCT in respiratory or blood specimen (Jin et al., 2020; Alhazzani et al., 2020). All patients with respiratory distress with one of the following criteria were eligible: respiratory rate > 30/min, or oxygen saturation ≤ 93%, or PaO2/FiO2 ratio ≤ 300 mmHg. We screened medical records on admission and identified patients with pulse oximetry ≤ 93% on room air and requires oxygen therapy (OT). Exclusion criteria included: 1) patients with chronic obstructive pulmonary disease with baseline pulse oximetry < 92%; 2) pregnant women; 3)
subjects younger than 18 years old; 4) patients with do-not-resuscitate order; and 5) patients with comorbidities such as severe burn, recent major stroke with paralysis, terminally ill malignancy, immunodeficiency and dialysis-dependent renal failure.

Clinical variables

Demographics such as age and sex were recorded. Comorbidities were recorded in broad categories such as those involving respiratory system and cardiovascular system. The smoking history were extracted from the medical records. All laboratory variables were recorded in a longitudinal manner. These included serum lactate, arterial partial oxygen pressure (PaO$_2$), arterial partial pressure of carbon dioxide (PaCO$_2$), base excess (BE), pH, C-reactive protein (CRP), Lymphocyte count, and fraction of inspired oxygenation (FiO$_2$) were extracted.

Respiratory support included OT, NIV, HFNC, IMV and ECMO. The transition time from one type to another was recorded to create a number of time intervals at which a subject was on a specific type of respiratory support. Laboratory variables were then matched to each time interval by their respective measurement time. This created a dataset of counting process that included the start time and end time for an interval.

Clinical outcomes included vital status at hospital discharge, length of stay in the hospital were recorded.

Calculation of cumulative oxygen deficit (COD)

For patients with IMV, COD was calculated before the use of IMV. Figure 1 is a sample patient used to illustrate the calculation of COD: COD (mmHg · day) = 80 × (t$_5$ − t$_1$) −

\[ \sum_{i=1}^{4} (x_{i+1} + x_{i}) \cdot \frac{(t_{i+1} - t_{i})}{2} \]

where $x_i$ is the value of PaO$_2$ measured in mmHg, and $t_i$ is the time at which $x_i$ is measured. The low end value of PaO$_2$ was 80 mmHg in our hospital and this value is also physiologically reasonable that the oxygen saturation will not continue to rise above
this reference value (Collins et al., 2015). Thus, the COD accounted for both magnitude and
duration of hypoxemia before IMV. We hypothesized that the longer a patient was on hypoxemia
before IMV, the worse of the survival outcome. On the other hand, the outcome would be not so
bad if hypoxemia was immediately corrected with IMV even if the magnitude of hypoxemia is
large.

Statistical analysis

Demographic and laboratory data were compared between patients with and without IMV.
Quantitative data were first tested for normality by using the Kolmogorov-Smirnov (K-S)
normality test. Normal data were expressed as mean and standard deviation and were compared
between groups with t test. Non-normally distributed data were expressed as median and
interquartile range (IQR) and were compared with Wilcoxon Rank Sum test. Categorial variables
were expressed as the number and percentage and were compared using chi-square or Fisher’s
exact test if appropriate (Zhang et al., 2017).

Alluvial plot was employed to visualize how patients transitioned from different types of
respiratory support over time. In patients with IMV, we created multivariable Cox regression
model to explore the independent predictors of survival outcome. The COD was categorized into
four categories at cutoff values of 0, 30 and 50 mmHg · day. A COD value of 30 mmHg · day is
equivalent to 60 mmHg for 1.5 days, and a negative value indicates no oxygen deficit. Other
variables such as time from admission to IMV, PaO$_2$, PaCO$_2$, Lactate, lymphocyte count, CRP
and BE were adjusted for in the model. These variables were included in multivariable
regression model because they were considered to be confounders by domain expertise and/or
univariate analysis with p < 0.2. The predictive performance of COD was compared with PaO$_2$
before intubation and the time from admission to intubation. We reported time-dependent AUC
for the discriminations from day 7 to 28 after hospital admission (Kamarudin, Cox & Kolamunnage-Dona, 2017).

All statistical analyses were performed using RStudio (Version 1.1.463; R version: 4.0.0).

## Results

### Study population

A total of 111 patients met the inclusion criteria and were included for analysis. No patients were excluded due to COPD, pregnancy CVA and paralysis. There was no patient being excluded from the participating hospitals. There were 80 patients who did not need IMV, and 31 patients required IMV during hospitalization. Patients with IMV had substantially lower PaO\(_2\) (62 (49, 89) vs. 90.5 (68, 125.25) mmHg; \(p < 0.001\)), higher pH (7.44 (7.38, 7.47) vs. 7.40 (7.35, 7.43); \(p = 0.006\)), higher serum lactate (2.5 (1.7, 3.1) vs. 1.7 (1.1, 2.85) mmol/L, \(p < 0.036\)) and higher COD (-6.87 (-29.36, 52.38) vs. -231.68 (-1040.78, 119.83) mmHg · day\(^{-1}\)) than patients without IMV during hospitalization (Table 1). These variables were reported as the first value during hospitalization. The time courses of the transition from different types of respiratory support are shown in Figure 2. It is noted that larger proportion of patients required IMV in the non-survivors.

### Independent association of COD and survival outcome in IMV patients

COD was independently associated with survival outcome in multivariable Cox regression model. As compared to patients with COD < 0, patients with COD from 0 to 30 mmHg · day\(^{-1}\) were not more likely to die, whereas those with COD > 30 mmHg · day\(^{-1}\) had higher risk of fatality (HR: 3.79, 95% CI: 2.57 to 16.93; \(p = 0.037\)), and those with COD > 50 mmHg · day\(^{-1}\) were 10 times more likely to die (HR: 10.45, 95% CI: 1.28 to 85.37; \(p = 0.029\)). The time from
admission to intubation, PaO2 and lymphocyte count were not associated with survival outcome (Table 2). The time-dependent AUCs of COD, PaO2 and the time from admission to intubation are shown in Figure 3. It showed that COD had consistently higher AUCs from day 14 to 21. In other words, COD was the best predictor after day 14. Table 3 shows factors associated with IMV.

Discussion

The study developed a novel biomarker COD which considered both magnitude and duration of hypoxemia, to assist the timing of IMV in patients with COVID-19. In patients with IMV during hospitalization, COD before intubation was a strong predictor of survival outcome. Patients with COD > 30 mmHg · day, which is equivalent to a persistent hypoxemia with PaO2 of 60 mmHg for 1.5 days, are more likely to die during hospitalization. Patients in crowded hospital during COVID-19 pandemic were more likely to experience this situation. The time dependent AUCs of COD were substantially higher than that of the PaO2 or the time from admission to intubation alone. Clinical implication of this finding is that we need to consider both the magnitude and duration of hypoxemia before IMV is considered. Long duration of mild hypoxemia, which is prevalent in clinical practice under NIV, may be dangerous for COVID-19 patients.

Many studies have been conducted to address the question on whether NIV should be used for patients with pulmonary/direct ARDS, but the results are conflicting (Chawla et al., 2020). NIV was not associated with improved mortality or length of stay, compared with patients who were intubated without trying NIV in a cohort of Middle East Respiratory Syndrome (MERS) patients. Furthermore, most patients (92.4%) who had tried NIV were eventually managed with IMV (Alraddadi et al., 2019). However, this was a retrospective study and the initiation of IMV
was not standardized prospectively. Our study indicated that large COD can be harmful and the
correction of COD with IMV might be beneficial. This could be explained by potential adverse
effects of NIV including large tidal volumes and injurious transpulmonary pressures (Brochard et
al., 2014). These adverse effects of NIV could be avoided by using IMV. For example, protective
ventilation strategy can be performed with IMV (Zhang et al., 2015; Fan, Brodie & Slutsky,
2018), but it is impossible under NIV. Furthermore, the use of NIV or HFNC can delay IMV,
leading to emergency or more unstable intubations (Brochard, 2003). Thus, IMV should be
considered as early as possible if the COD reaches 30 mmHg · day, without trying NIV or
HFNC to delay intubation.

PaO$_2$ and its derivatives such as PF ratio are well established risk factor for mortality outcome in
patients with ARDS. Thus, PF ratio is used to classify ARDS patients into mild, moderate and
severe cases (ARDS Definition Task Force et al., 2012). However, this risk classification system
considers only the magnitude of hypoxemia (Cartotto et al., 2016; Dai et al., 2019). Our results
suggest that the duration of hypoxemia can be equally important. A strength of our study was
that all measurements of PaO$_2$ were collected longitudinally, allowing for the calculation of the
area under the PaO$_2$-day curve to derive a novel predictor. Our analysis focused on patients with
IMV and found that the predictive performance for survival outcome of COD before intubation
was substantially better than PaO$_2$ or the time from admission to intubation. The latter two
indices are the two components of COD. The combination of the two indices substantially
improves the predictive discrimination for mechanically ventilated patients. Although direct
causal inference that the use of IMV to reduce COD can improve survival outcome cannot be
established with current analysis, our result identified a modifiable risk factor for survival
outcome. It is reasonable to deduce that reducing COD as early as possible with IMV can be beneficial.

Several limitations should be acknowledged in the study. First, the study was retrospective in design, and many unmeasured confounders may exist to influence the choice of respiratory supports (Uddin et al., 2016). The presence of such unmeasured confounders will compromise the effectiveness of the propensity score matching procedure. Second, the use of NIV or HFNC was completely at the discretion of the treating physician. There was no standard protocol in participating hospitals. Thus, it is difficult to determine whether the use of NIV or HFNC could benefit COVID-19 induced ARDS. Third, for patients without IMV, we calculated the COD across all days of hospitalization. This could be biased because the time-dimension was longer than the IMV group. However, since non-IMV group generally did not have oxygen deficit across hospital stay, the COD was substantially lower than the IMV group. Finally, we only included broad categories of comorbidity burden in our analysis (i.e., respiratory system, cardiovascular system), because the retrospective design of the study did not allow detailed information for the calculation of the Elixhauser's comorbidity index. It is well known that Elixhauser's comorbidity index is a good quantity for risk stratification of hospitalized patients (Elixhauser et al., 1998). However, this index is designed to work with ICD-9-CM codes in administrative database, which is not applicable to data collected in retrospective studies.

Conclusions

In conclusion, the study developed a novel predictor COD, which considered both magnitude and duration of hypoxemia, to assist the timing of IMV in patients with COVID-19. We suggest
IMV should be the preferred ventilatory support once the COD reaches 30 mmHg · day, as mortality increases beyond this value.
Figure legends

Figure 1. Schematic illustration of the Calculation of cumulative oxygen deficit (COD). The COD was calculated as the difference of the areas under the reference curve and the PaO$_2$-day curve (the light green area in the figure).

Figure 2. Alluvium plot showing the transitions of respiratory supports over time.

Figure 3. Time-dependent AUCs for cumulative oxygen deficit, PaO$_2$ and the time from admission to intubation. The AUC of cumulative oxygen deficit was substantially higher than the other two indices from day 14 to 24.
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Table 1 (on next page)

Table 1 Comparison between patients with and without invasive mechanical ventilation

Comparison between patients with and without invasive mechanical ventilation
| Variables                          | Total (n = 111) | No IMV (n = 80) | IMV (n = 31) | p  |
|-----------------------------------|----------------|----------------|--------------|----|
| Age, Median (IQR)                 | 57 (45.5, 68.5)| 56 (40, 67.25)| 64 (54.5, 70.5)| 0.076|
| Sex, male (%)                     | 60 (55)        | 39 (50)        | 21 (68)      | 0.143|
| Comorbidities                     |                |                |              |    |
| Respiratory system, n (%)         | 18 (16)        | 13 (16)        | 5 (16)       | 1   |
| Cardiovascular system, n (%)      | 41 (37)        | 29 (36)        | 12 (39)      | 0.983|
| Smoking history, n (%)            | 2 (2)          | 1 (1)          | 1 (3)        | 0.482|
| PaCO2 (mmHg), Median (IQR)        | 43 (39, 47.5)  | 44 (41, 48)    | 39 (36, 45)  | 0.051|
| PaO2 (mmHg), Median (IQR)         | 82 (60, 120)   | 90.5 (68, 125.25) | 62 (49, 89) | < 0.001|
| pH, Median (IQR)                  | 7.42 (7.35, 7.44) | 7.40 (7.35, 7.43) | 7.44 (7.38, 7.47) | 0.006|
| BE (mmol/l), Median (IQR)         | 2.4 (-2.05, 5.1) | 2.35 (-2.57, 5.1) | 2.4 (0.2, 5.2) | 0.342|
| Lactate (mmol/l), Median (IQR)    | 1.9 (1.3, 3)   | 1.7 (1.1, 2.85) | 2.5 (1.7, 3.1) | 0.036|
| PF (mmHg), Median (IQR)           | 224.32 (106.14, 348.02) | 250.05 (194.14, 363.64) | 111.25 (71.98, 200.27) | < 0.001|
| CRP (mg/dl), Median (IQR)         | 8.9 (2.75, 53.2) | 5.5 (1.78, 35.83) | 19.5 (9.85, 78.05) | < 0.001|
| Lymphocyte count (× 10^9/L), Median (IQR) | 0.78 (0.28, 0.86) | 0.78 (0.43, 0.89) | 0.43 (0.28, 0.66) | 0.005|
| COD (mmHg · day), Median (IQR)    | -50.38 (-827.91, 72.67) | -231.68 (-1040.78, 119.83) | -6.87 (-29.36, 52.38) | 0.018|
| COD, n (%)                        |                |                |              | < 0.001|
| < 0 (mmHg · day)                  | 56 (50)        | 49 (61)        | 7 (23)       |     |
| 0-30 (mmHg · day)                 | 14 (13)        | 3 (4)          | 11 (35)      |     |
| 30 – 50 (mmHg · day)              | 9 (8)          | 5 (6)          | 4 (13)       |     |
| > 50 (mmHg · day)                 | 32 (29)        | 23 (29)        | 9 (29)       |     |
| Hospital LOS (days), Mean ± SD    | 26.55 ± 11.78  | 26.1 ± 11.35   | 27.7 ± 12.97 | 0.548|
| Mortality, n (%)                  | 22 (20)        | 13 (16)        | 9 (29)       | 0.211|

Abbreviations: LOS: length of stay; SD: standard deviation; COD: cumulative oxygen deficit; CRP: C-reactive protein; IQR: interquartile range; BE: base excess
Table 2

Multi-variant Cox regression model to explore the independent association of cumulative oxygen deficit and hazard of death in patients with invasive mechanical ventilation.

| Abbreviations: | COD: cumulative oxygen deficit; CRP: C-reactive protein; IQR: interquartile range; BE: base excess; HR: hazard ratio; CI: confidence interval |
Table 2 Multivariable Cox regression model to explore the independent association of cumulative oxygen deficit and hazard of death in patients with invasive mechanical ventilation

| COD (< 0 as reference) | HR [95% CI]   | p    |
|------------------------|---------------|------|
| 0-30                   | 1.45 [0.98, 3.47] | 0.052|
| 30-50                  | 3.79 [2.57, 16.93] | 0.037|
| >50                    | 10.45 [1.28, 85.37] | 0.029|
| Time from admission to intubation | 0.78 [0.59, 1.03] | 0.078|
| Age                    | 1.20 [1.00, 1.44] | 0.056|
| Sex (female as reference) | 9.84 [0.43, 225.59] | 0.152|
| PaCO2                  | 0.85 [0.60, 1.19] | 0.344|
| PaO2                   | 1.00 [0.97, 1.03] | 0.939|
| pH                     | 0.70 [0.44, 1.13] | 0.144|
| BE                     | 1.89 [0.70, 5.08] | 0.206|
| CRP                    | 0.98 [0.94, 1.01] | 0.233|
| Lymphocyte count       | 2.21 [0.04, 115.48] | 0.695|

Abbreviations: COD: cumulative oxygen deficit; CRP: C-reactive protein; IQR: interquartile range; BE: base excess; HR: hazard ratio; CI: confidence interval
**Table 3** (on next page)

Table 3 Factors associated with use of IMV

Abbreviations: CRP: C-reactive protein; HR: hazard ratio; CI: confidence interval
Table 3 Factors associated with use of IMV

| Variables                        | HR [95% CI]       | p    |
|----------------------------------|-------------------|------|
| PaO2 (for every 10-mmHg increase)| 0.91 [0.84, 0.99] | 0.022|
| PaCO2                            | 1.02 [0.98, 1.06] | 0.337|
| Age                              | 1.02 [0.99, 1.04] | 0.235|
| pH (for every 0.01 increase)     | 1.02 [0.96, 1.09] | 0.565|
| CRP                              | 1.00 [0.99, 1.01] | 0.824|
| Lymphocyte count                 | 0.27 [0.09, 0.81] | 0.020|
| Lactate                          | 1.56 [1.17, 2.08] | 0.003|

Abbreviations: CRP: C-reactive protein; HR: hazard ratio; CI: confidence interval
Figure 1

Figure 1. Schematic illustration of the Calculation of cumulative oxygen deficit (COD).

The COD was calculated as the difference of the areas under the reference curve and the PaO₂-day curve (the light green area in the figure).
Sample Patient

PaCO₂ (mmHg)

Respiratory Support Type
- IMV
- OT

Days after Admission

Manuscript to be reviewed
Figure 2

Figure 2. Alluvial plot showing the transitions of respiratory supports over time.
Figure 3

Figure 3. Time-dependent AUCs for cumulative oxygen deficit, PaO$_2$ and the time from admission to intubation.

The AUC of cumulative oxygen deficit was significantly higher than the other two indices from day 14 to 24.