# COVID-19 Pandemic and Hospital Efficiency in Iran: Insight from an Interrupted Time Series Analysis and Pabon Lasso Model

Monireh Mahmoodpour-Azari<sup>1</sup>, Mohammad Hajizadeh<sup>2</sup>, Ali Kazemi-Karyani<sup>1</sup>, Afshar Haidari<sup>1</sup>, Satar Rezaei<sup>1,\*</sup> <sup>1</sup>Research Center for Environmental Determinants of Health, Health Institute, Kermanshah University of Medical Sciences, Kermanshah, Iran, <sup>2</sup>School of Health Administration, Faculty of Health, Dalhousie University, Halifax, Canada

**Background:** Limited evidence exists on the impact of the Coronavirus Disease 2019 (COVID-19) pandemic on hospital efficiency worldwide. This study aimed to investigate the impact of the COVID-19 pandemic on public hospitals in Iran. **Methods:** In this quasi-experimental study, monthly data on the average length of stay (ALoS), bed occupancy rate (BOR), bed turnover rate (BTR) and bed turnover interval (BTI) were collected for 58 months (36 months before and 22 months after the COVID-19 outbreak in Iran) from the health information systems of all 18 public hospitals affiliated with Kermanshah University of Medical Sciences in Kermanshah province, Iran. We used interrupted time series analysis and the Pabon Lasso model to investigate the impact of the COVID-19 pandemic on hospital efficiency.

**Results:** The monthly average ALoS, BOR, BTR and BTI before (after) the COVID-19 pandemic was 3.30 (3.48) days, 70.14% (49.37), 6.78 (4.81) patients per bed and 1.15 (2.88) days, respectively. The study indicated that a statistically significant decrease in ALoS of 0.29 and BOR of 25.09 in the first month following the COVID-19 pandemic. Compared with the before pandemic period, we observed a significant increase in the monthly trend of ALoS (coefficient = 0.021; p = 0.015), BOR (coefficient = 1.30; p = 0.002), and BTR (coefficient = 0.08, p = 0.012). We found a significant decrease in the monthly trend in BTI (coefficient = -0.11, p = 0.009) after the COVID-19 pandemic when compared with before the pandemic. Based on the Pabon Lasso model, before (after) the pandemic, 29.4% (29.4%) of the hospitals were located in zone 1 as an inefficient area, and 17.6% (35.3%) of hospitals were located in zone 3 as an efficient area.

**Conclusion:** The study demonstrated that the BOR and BTR decreased substantially after the outbreak of COVID-19. In contrast, the ALoS and BTI have significantly increased following the COVID-19 pandemic. We also found that hospitals' performance in both periods was poor, and only 30% of hospitals were located in the efficient zone (zone three) based on the Pabon Lasso model. Further studies aimed at identifying the main factors affecting lower efficiency among hospitals in Iran are recommended.

Key Words: COVID-19 pandemic, Hospital efficiency, Pabon Lasso model, Interrupted time series analysis, Iran

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\*Corresponding author: Satar Rezaei

Research Center for Environmental Determinants of Health, Health Institute, Kermanshah University of Medical Sciences, Isar Sq., across from Farabi Hospital, Kermanshah 6719851351, Iran Tel: 98–83–3828–1991, Fax: 98–83–3826–3048 E-mail: starrezaei@gmail.com

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# INTRODUCTION

The Coronavirus Disease 2019 (COVID-19) was first diagnosed in Wuhan, China, in late December 2019, and it has spread with incredible speed throughout the world [1]. On 11 March, 2020, the World Health Organization (WHO) declared it as a pandemic, and it stated that this is the first pandemic caused by a coronavirus [2]. In Iran, the first death occurred by the COVID-19 was confirmed in Qom province on 19 February, 2020 and quickly spread to other provinces of Iran [3]. The COVID-19 pandemic has resulted in health and economic crises with a significant impact on the delivery of healthcare worldwide [4]. The pandemic has disrupted healthcare utilization [5,6], delayed initial diagnosis and treatment, increased demand for telehealth [7,8] and hospital services [9,10]. A significant reduction in the number of persons seeking medical care during the COVID-19 outbreak [11-15] suggests that the pandemic is impacting the behavior of entire populations in relation to willingness to receive care among non-COVID patients [16]. Hospitals have also reported unexplained reductions in admissions for severe medical illnesses [17]. In a systematic review study on the impact of COVID-19 pandemic on health care utilization, Moynihan et al. [18] showed showed a 30% decrease in health care utilization by 30%. The reduction in health care utilization was found to be higher among people with less severe illnesses.

Hospitals in all countries, regardless of development status, have undoubtedly been affected in a number of areas during the COVID-19 pandemic. Evaluating changes in hospital performance will enable us to better prepare for future crises and design cost-effective interventions for the optimal use of limited resources in the health sector. To date, only limited studies aim to assess the effect of COVID-19 pandemic on the efficiency of hospitals in other developing and developed countries [19,20]. To the best of our knowledge, the effect of the COVID-19 pandemic on Iran's hospitals is not documented in the literature. In this study, we aimed to examine the impact of COVID-19 pandemic on the efficiency of hospitals affiliated with the Kermanshah University of Medical Sciences (KUMS) in the Kermanshah province in the west of Iran.

# MATERIALS AND METHODS

### 1. Study setting

This was a quasi-experimental and descriptive-analytical study carried out in Kermanshah province, located in the western region of Iran. The province's total population was estimated to be approximately two million in the year 2016. The province has 18 hospitals affiliated with KUMS that are responsible for the provision of most inpatient services.

#### 2. Data and study population

All 18 hospitals affiliated with KUMS were used in the study. Using a self-constructed checklist, monthly data on the number of active beds, number of inpatients, occupied bed-days and active bed-days were collected from the health information system (HIS) of hospitals. Data were collected for a period of 58 months (36 months before and 22 months after the COVID-19 outbreak. Since the first confirmed case of the COVID-19 was reported on 19 February 2020 in Iran, we used data from 19 February, 2017 to 18 February, 2020 as pre-pandemic period and 19 February, 2020 to 21 December, 2021 as post-pandemic period.

As per previous studies [21,22], we used four indicators viz., average length of stay (ALoS), bed occupancy rate (BOR), bed turnover rate (BTR) and bed turnover interval (BTI) as measures of hospital efficiency. We used the collected data and the following formulas to calculate these efficiency indicators.

$$BOR (\%) = \frac{Total inpatient days}{(Occupied beds, days)} \times 100$$

$$BTR = \frac{Number of discharges in 1 month}{Number of active beds}$$

ALoS (days) =  $\frac{\text{Total inpatient days (Occupied beds, days)}}{\text{Number of dischargs}}$ 

 $BTI (days) = \frac{30}{BTR} - ALoS$ 

#### 3. Data analysis

We used two models to examine the impact of the COVID-19 pandemic on hospital efficiency, including interrupted time series analysis (ITSA) and the Pabon Lasso model. We used the following segmented regression to estimate the effect of the COVID-19 outbreak on four hospital efficiency indicators included in the study [23,24]:

$$Y_t = \beta_0 + \beta_1 * time_t + \beta_2 * COVID19_t + \beta_3$$
$$* timeafterCOVID19_t + \varepsilon_t$$

Where  $Y_t$  presents hospital efficiency indicator (e.g., ALoS) in the month t, time<sub>t</sub> is time trend variable that takes values between 1 (first observation) and 58 (last observation). The  $COVID19_t$  is a binary variable for before (COVID19= 0) and after (COVID19 = 1) the outbreak. The *timeafterCOVID*19 $_t$  variable is assigned 0 for the period before the COVID-19 outbreak and coded between 1 and 22 for the period after the COVID-19 outbreak. In the model,  $\beta_0$  estimates the starting level of the outcome variable at time zero,  $\beta_1$  estimates the monthly change in the outcome variable before the COVID-19 outbreak (the pre-existing trend);  $\beta_2$  estimates the level change in the indicators immediately after the COVID-19 outbreak, and  $\beta_3$  estimates the monthly change in the outcome variable trend after COVID-19, compared with the pre-existing trend. The Newey-West approach was used to handle autocorrelation and possible heteroscedasticity [23,25] in the estimation. All data analysis related to ITSA model was performed using Stata statistical software, and 0.05 was considered a significant level (Version 16.0; Stata Corporation, College Station, TX, USA). Data on 18 hospitals was used in ITSA.

Pabon Lasso model was also used to determine the relative performance of the hospitals in terms of three efficiency indicators: ALoS, BOR and BTR [26]. In the Pabon Lasso model, interpretation of hospital performance is based on a chart that displays BOR on the Y-axis and BTR on the X-axis. Each hospital can be located in one zone out of four zones (Fig. 1). The definitions and interpretations of the four zones in the Pabon Lasso chart are given in Fig. 1. Since one of the hospitals was established after the outbreak of the COVID-19 pandemic, it was excluded from the Pabon Lasso analysis, and the data for 17 hospitals were included. Excel software was used to construct the Pabon Lasso graph.

## RESULTS

The summary of the variables included in the study are reported in Table 1. The overall mean (standard deviation



Fig. 1. Pabon Lasso graph illustrating the relative performance of hospitals.

| Table 1. | Summary | of | variables | included | in | the | study |
|----------|---------|----|-----------|----------|----|-----|-------|
|          |         |    |           |          |    |     |       |

|             | Total period of the study (Mean $\pm$ SD) | Before the pandemic (Mean $\pm$ SD) | After the pandemic (Mean $\pm$ SD) |
|-------------|---|-------------------------------------|------------------------------------|
| ALoS (days) | $3.37 \pm 0.25$                           | $3.30 \pm 0.15$                     | $3.48 \pm 0.33$                    |
| BOR (%)     | $62.26 \pm 13.00$                         | $70.14 \pm 5.88$                    | 49.37 ± 11.00                      |
| BTR         | $6.03 \pm 1.14$                           | $6.78 \pm 0.54$                     | $4.81 ~\pm~ 0.70$                  |
| BTI (days)  | $1.81 \pm 1.14$                           | $1.15 \pm 0.28$                     | $2.88 \pm 1.21$                    |

SD: standard deviation, ALoS: average length of stay, BOR: bed occupancy rate, BTR: bed turnover rate, BTI: bed turnover interval.

[SD]) of ALoS, BOR, BTR and BTI was 3.37 (0.25), 62.26 (13.00), 6.03 (1.14) and 1.81 (1.14), respectively, during the study period. The mean of ALOS before the COVID-19 pandemic was 3.30 (SD = 0.15), while it was 3.48 (0.33) for the period after the pandemic.

Our study indicated 3.15 days as the starting level of the ALoS (see Table 2). The ALoS statistically significantly increased by 0.0008 per month (p = 0.005) before the COVID-19 pandemic. We also found a statistically significant decrease in ALoS of 0.29 (p = 0.013) in the first month of the COVID-19 pandemic. Compared with the before the pandemic, we observed a significant increase in the monthly trend of ALoS at 0.021 (p = 0.015). The results for ALoS for the period before and after the COVID-19 pandemic have been illustrated in Fig. 2.

The results of BOR are shown in Table 3 and Fig. 3. As shown in Table 3, the starting level of the BOR was estimated at 75.95 (p = 0.001). The BOR was significantly decreasing by 0.33 (p = 0.002) per month prior to the pandemic outbreak. We observed that in the first month of the pandemic, BOR was significantly decreased by 25.09 (p = 0.001), followed by a significant increase in the monthly trend of BOR (compared to the before pandemic) of 1.30 (p = 0.002). After the COVID-19 pandemic, BOR significantly increased by 0.97 (p = 0.016) per month.

As reported in Table 4, the starting level of the BTR was estimated at 7.26. The BTR decreased by 0.03 (p = 0.001) per month before the COVID-19 pandemic. A statistically significant increase in BTR of 1.99 (p = 0.001) was observed in the first month of the pandemic. Compared with

Table 2. Estimated coefficients of segmented regression model for average length of stay (ALoS, days) before and after COVID-19 pandemic in Iran

|   | Coefficient | SE*   | p-value – | Cl (95%) |        |
|---|-------------|-------|-----------|----------|--------|
| Variables                               |             |       |           | Lower    | Upper  |
| Intercept, $\beta_0$                    | 3.15        | 0.05  | < 0.001   | 3.05     | 3.25   |
| Pre-COVID-19 slope, $\beta_1$           | 0.008       | 0.003 | 0.005     | 0.002    | -0.062 |
| Change in slope, $\beta_2$              | -0.29       | 0.11  | 0.013     | -0.51    | -83.50 |
| Change in trend, $\beta_3$              | 0.021       | 0.008 | 0.015     | 0.004    | 0.038  |
| Post-COVID-19 linear trend $^{\dagger}$ |             |       |           |          |        |
| Linear trend, $\beta_{p1}$              | 0.029       | 0.008 | 0.0004    | 0.013    | 0.045  |
|   |             |       |           |          |        |

SE: standard error, CI: confidence interval.

\*Newey-West standard errors.

<sup>†</sup>This obtained from the following time trend Equation:  $Y_{pt} = \beta_{p0} + \beta_{p1} * time_{pt} + \varepsilon_t$  where  $Y_{pt}$  is the value of ALoS at time *t* after the COVID-19 pandemic and *time<sub>pt</sub>* is the time trend variable which takes values between 1 (first observation after the pandemic) and 22 (last observation after the pandemic).

Table 3. Estimated coefficients of segmented regression model for bed occupancy rate (BOR, %) before and after COVID-19 pandemic in Iran

| Variables  | Coefficient | SE*  | p-value - | Cl (95%) |        |
|--|-------------|------|-----------|----------|--------|
| Valiables  | Coencient   |      |           | Lower    | Upper  |
| Intercept, $\beta_0$                                       | 75.95       | 1.87 | < 0.001   | 72.20    | 79.71  |
| Pre-COVID-19 slope, $\beta_1$                              | -0.33       | 0.10 | 0.002     | -0.53    | -0.12  |
| Change in slope, $\beta_2$                                 | -25.09      | 4.90 | < 0.001   | -34.93   | -15.25 |
| Change in trend, $\beta_3$                                 | 1.30        | 0.39 | 0.002     | 0.52     | 2.08   |
| Post-COVID-19 linear trend <sup><math>\dagger</math></sup> |             |      |           |          |        |
| Linear trend, $\beta_{p1}$                                 | 0.97        | 0.37 | 0.016     | 0.226    | 1.729  |

SE: standard error, CI: confidence interval.

\*Newey-West standard errors.

<sup>†</sup>This obtained from the following time trend Equation:  $Y_{pt} = \beta_{p0} + \beta_{p1} * time_{pt} + \varepsilon_t$  where  $Y_{pt}$  is the value of BOR at time *t* after the COVID-19 pandemic and *time<sub>pt</sub>* is the time trend variable which takes values between 1 (first observation after the pandemic) and 22 (last observation after the pandemic).



Fig. 2. Interrupted time series analysis with Newey-West standard errors and two lags for ALoS. Note: The date of the first confirmed case of COVID-19 in Iran (19 February, 2020) was used as the start of the pandemic period.

Table 4. Estimated coefficients of segmented regression model for bed turnover rate (BTR) before and after COVID-19 pandemic in Iran

| Coefficient | SE*  | p-value –   | Cl (95%)                                      |  |
|-------------|--|---|---|--|
|             |  |   | Lower   | Upper  |
| 7.26        | 0.21   | < 0.001   | 6.83  | 7.69   |
| -0.03       | 0.009  | < 0.001   | -2.74   | -1.25  |
| 1.99        | 0.37   | < 0.001   | -34.93  | -15.25   |
| 0.08        | 0.03   | 0.012   | 0.02  | 0.14   |
|             |  |   |   |  |
| 0.05        | 0.03   | 0.07  | -0.006  | 0.11   |
|             | Coefficient<br>7.26<br>-0.03<br>1.99<br>0.08<br>0.05 | Coefficient SE*   7.26 0.21   -0.03 0.009   1.99 0.37   0.08 0.03   0.05 0.03 | Coefficient SE* p-value -   7.26 0.21 < 0.001 | Coefficient SE* p-value Cl (9)   7.26 0.21 < 0.001 |

SE: standard error, CI: confidence interval.

\*Newey-West standard errors.

<sup>†</sup>This obtained from the following time trend Equation:  $Y_{pt} = \beta_{p0} + \beta_{p1} * time_{pt} + \varepsilon_t$  where  $Y_{pt}$  is the value of BTR at time *t* after the COVID-19 pandemic and *time<sub>pt</sub>* is the time trend variable which takes values between 1 (first observation after the pandemic) and 22 (last observation after the pandemic).

the period before the pandemic, we observed a significant increase in the monthly trend of BTR (coefficient = 0.08, p = 0.012). Segmented regression results also indicated that after the COVID-19 pandemic, the BTR increased monthly at the rate of 0.05 (p = 0.07). Fig. 4 provides a visual display of these results for the period before and after the COVID-19 pandemic.

Results also demonstrated that BTI was 1.00 at the starting month of the study period (see Table 5). There was no significant change in the BTI (p = 0.07) during the period before the COVID-19 pandemic. The estimated proportion of the BTI increased by 2.69 (p = 0.001) in the first month following the pandemic outbreak. We observed a significant decrease in the monthly trend of the BTI after the COVID-19 pandemic compared with the before the pandemic period (p = 0.009). Furthermore, the BTI decreased every month at the rate of 0.10 (p = 0.01) during the COVID-19 outbreak period. Fig. 5 provides a visual illustration of these results for the period before and after the pandemic.

The results of the Pabon Lasso model on hospital efficiency are illustrated in Fig. 6. Hospitals were located in four performance zones. Panels A and B in Fig. 5 show the location of hospitals before and after the COVID-19 pandemic, respectively. The results for the period before the COVID-19 pandemic suggest that 29.4% (five) of hospitals were located in zone 1, 17.6% (three) of hospitals were located in zone 2, 17.6% (three) of hospitals were located in zone 3, and 23.5% (four) of hospitals were located in zone 4. One hospital was located exactly on the border between





Fig. 4. Interrupted time series analysis with Newey–West standard errors and two lags for BTR. Note: The date of the first confirmed case of COVID-19 in Iran (19 February, 2020) was used as the start of the pandemic period.

Fig. 5. Interrupted time series analysis with Newey–West standard errors and two lags for BTI. Note: The date of the first confirmed case of COVID-19 in Iran (19 February, 2020) was used as the start of the pandemic period. zones 1 and 2 and another was exactly on the border between zones 2 and 3. After the pandemic, 29.4% (five) of hospitals were in zone 1, 11.8% (two) of hospitals were in zone 2, 35.3% (six) of hospitals were in zone 3 and 29.4% (four) of hospitals were in zone 4. We also observed that one hospital was located exactly on the border between



**Fig. 6.** A Pabon Lasso graph for 17 hospitals affiliated with the Kermanshah University of Medical Sciences before and after the COVID-19 pandemic.

zones 3 and 4.

## DISCUSSION

The COVID-19 pandemic has affected the performance and efficiency of hospitals and the extent to which people use health services [27]. In this study, we investigated the effect of COVID-19 outbreak on the efficiency of public hospitals in western part of Iran. As per previous studies [21,22,28], four indicators, including BOR, BTR, ALoS and BTI were used to assess the hospital performance.

Our study suggested increasing trends in ALoS before and after the COVID-19 pandemic. The rate of increase in ALoS was found to be higher after the outbreak compared to before the outbreak. The ALoS was significantly reduced in the first month following the outbreak of the disease. The significant decrease in the first month after the COVID-19 pandemic can be attributed to the fear of people getting the infection and trying to get an early discharge by both patients and providers. The increase in the ALoS in the months after the outbreak of COVID-19 can be explained by the fact that patients with mild diseases have delayed their hospitalization until after the end of the pandemic. The longer length of hospital stay of COVID-19 patients compared to patients with other diseases [29] might have contributed to the increase in the ALoS after the pandemic.

As a measure to evaluate the utilization of hospital beds, the BOR represents the efficiency in the use of hospital beds. Our results showed that BOR had a significantly de-

| Variables                               | Coefficient | SE*   | p-value - | Cl (95%) |       |
|---|-------------|-------|-----------|----------|-------|
| vanables                                |             |       |           | Lower    | Upper |
| Intercept, $\beta_0$                    | 1.00        | 0.10  | < 0.001   | 0.79     | 1.20  |
| Pre-COVID-19 slope, $\beta_1$           | 0.008       | 0.004 | 0.074     | -0.0008  | 0.02  |
| Change in slope, $\beta_2$              | 2.69        | 0.54  | < 0.001   | 1.59     | 3.78  |
| Change in trend, $\beta_3$              | -0.11       | 0.04  | 0.009     | -0.20    | -0.03 |
| Post-COVID-19 linear trend <sup>†</sup> |             |       |           |          |       |
| Linear trend, $\beta_{p1}$              | -0.10       | 0.04  | 0.01      | -0.18    | -0.02 |

Table 5. Estimated coefficients of segmented regression model for bed turnover interval (BTI, days) before and after COVID-19 pandemic in Iran

SE: standard error, CI: confidence interval.

\*Newey-West standard errors.

<sup>†</sup>This obtained from the following time trend Equation:  $Y_{pt} = \beta_{p0} + \beta_{p1} * time_{pt} + \varepsilon_t$  where  $Y_{pt}$  is the value of BTR at time *t* after the COVID-19 pandemic and *time<sub>pt</sub>* is the time trend variable which takes values between 1 (first observation after the pandemic) and 22 (last observation after the pandemic).

creasing trend in the months following the COVID-19 pandemic compared to before the pandemic. The reduction in BOR in the first month after the pandemic outbreak was about 30%. The Ministry of Health and Medical Education (MOHME) in Iran classifies the performance of hospitals into three groups based on the BOR: satisfactory performance (70% and above), somehow satisfactory performance (60-70%) and not satisfactory performance (less than 60%) [26]. The average BOR in the pre-pandemic period was about 70%, while the corresponding figure was around 50% after the COVID-19 pandemic. This indicates the under-utilization of health care resources in public hospitals in Kermanshah province during the COVID-19 pandemic as compared with the period before the pandemic. One of the main reasons for the decrease in the BOR in the hospitals during the pandemic is a decrease in the hospitalization rates for non-COVID disease [30-32]. Reduction in hospital admission was documented during the pandemic in some countries [30]. This reduction in hospital admission can be attributed to the factors such as rationing of the medical workforce and the fear people to seeking hospital care.

Our study also indicated that after the outbreak of the COVID-19 pandemic, the BTR decreased while the BTI increased. BTR measure the number of patients hospitalized per bed at a specific time (in this study, one month). BTI shows the average number of days that hospital beds are unoccupied between successive inpatients over a given period, such as month or year (in this study, month). The BTR and BTI can assess the productivity of hospital beds. The lower value of BTR and higher value of BTI in a hospital suggest that a hospital is less efficient. The results of BTR and BTI indicated a high level of inefficiency and poor performance of hospitals for beds utilization.

We also used the Pabon Lasso model to assess overall hospital efficiency before and after the COVID-19 pandemic. The Pabon Lasso model enabled us to simultaneously evaluate hospital efficiency using the three indicators of ALoS, BOR and BTR. Our study indicated that before (after) the COVID-19 pandemic, 29.4% (n = 5) of hospitals were in zone 1: an inefficient zone characterized by low BOR, high ALoS and low BTR. Hospitals in this zone experience under-utilization of hospitals beds. We also found that before the COVID-19 pandemic, 17.6% of hospitals were in zone

3 while 35.3% of hospitals were in this zone during the COVID-19 pandemic period. The zone 3 is the efficient zone that characterized by high BOR, high BTR and short ALoS. Hospitals in zone 3 have satisfactory performance with the small number of unoccupied beds [33].

The current study had some limitations that should be considered in interpreting the findings. Firstly, this study was conducted in one province, so the generalizability of the results is limited. Secondly, we only used public hospitals data to assess the impact of the COVID-19 pandemic on hospital efficiency. Since data on Social Security Organization and private hospitals are not available to include in the analysis, the study results are not generalizable for these hospitals. Thirdly, we examined the impact of the COVID-19 pandemic on indicators of hospital efficiency three years before and approximately two years after (22 months) of the pandemic. Future studies are required to investigate the long-term impact of the COVID-19 pandemic on hospital performance.

# CONCLUSION

This study aimed to investigate and compare the efficiency of public hospitals in Kermanshah province, the western part of Iran, using ITSA and the Pabon Lasso model before and after the COVID-19 pandemic in Iran. The study demonstrated that the BOR and BTR decreased substantially after the outbreak of COVID-19 disease compared with before the pandemic. In contrast, the ALoS and BTI have significantly increased following the COVID-19 pandemic. These findings suggest the inefficient use of public hospital beds in Iran. As less than a third of public hospitals were classified as efficient hospitals (zone three) based on the Pabon Lasso model before and after the outbreak of COVID-19, further studies aimed at identifying the main factors affecting inefficiency among public hospitals are needed.

# CONFLICTS OF INTEREST

The authors declare that they have no competing interests.

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