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Impact of ceiling of care on mortality across waves: an inverse probability weighting analysis

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Complete List of Authors:	Pallarès, Natàlia; IGTP; Germans Trias i Pujol Research Institute and Hospital (IGTP) Videla, Sebastià; Germans Trias i Pujol University Hospital, Department of Clinical Pharmacology; University of Barcelona, Department of Pathology and Experimental Therapeutics Carratala, Jordi; Hospital Universitari de Bellvitge, Infectious Diseases; Bellvitge Biomedical Research Institute Tebé, Cristian; Germans Trias i Pujol Research Institute and Hospital (IGTP)
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Impact of ceiling of care on mortality across waves: an inverse probability weighting analysis
 Pallarès N^{1,2} MSc, Videla S^{3,4} MD PhD, Carratalà J^{5,6,7,8} MD PhD, Tebé C¹ MPH PhD, on behalf of
 the MetroSud study group and the Divine study group

1) Biostatistics Support and Research Unit, Germans Trias i Pujol Research Institute and
 Hospital (IGTP), Badalona, Barcelona, Spain

2) Department of Basic Clinical Practice, School of Medicine and Health Sciences, University of
 Barcelona, Spain

3) Clinical Research Support Area, Department of Clinical Pharmacology, Germans Trias i Pujol
 University Hospital, Badalona, Spain

4) Department of Pathology and Experimental Therapeutics, School of Medicine and Health
 Sciences, University of Barcelona, Spain

5) Department of Infectious Diseases, Bellvitge University Hospital, Barcelona, Spain

6) Bellvitge Biomedical Research Institute (IDIBELL), Barcelona, Spain

7) Centro de Investigación en Red de Enfermedades Infecciosas (CIBERINFEC), Instituto de
 Salud Carlos III, Madrid, Spain.

8) Department of Clinical Sciences, School of Medicine and Health Sciences, University of
 Barcelona, Spain

MetroSud Study group: Gabriela Abelenda-Alonso, Alexander Rombauts, Isabel Oriol,
 Antonella F. Simonetti, Alejandro Rodríguez-Molinero, Elisenda Izquierdo, Vicens Díaz-Brito,
 Carlota Gudiol, Judit Aranda-Lobo, Marta Arroyo, Carlos Pérez-López, Montserrat Sanmartí,
 Encarna Moreno, Maria C Alvarez, Ana Faura, Martha González, Paula Cruz, Mireia Colom,
 Andrea Perez, Laura Serrano.

DIVINE Study group: Mireia Besalú, Erik Cobo, Jordi Cortés, Daniel Fernández, Leire
 Garmendia, Guadalupe Gómez, Pilar Hereu, Klaus Langohr, Gemma Molist, Núria Pérez-
 Álvarez, Xavier Piulachs.

Corresponding author

Natàlia Pallarès, MSc
 Biostatistics Support and Research Unit
 Germans Trias i Pujol Research Institute and Hospital (IGTP)
 Campus Can Ruti. Carretera de Can Ruti, Camí de les Escoles s/n. 08916 Badalona, Barcelona,
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ABSTRACT

Objectives

The aim of this study was to compare in-hospital mortality across waves in patients without and with a ceiling of care at hospital admission.

Study design

Prospective cohort study.

Methods

Adult patients hospitalised for COVID-19 in five centres in Catalonia between March 2020 and August 2021 with information available on ceiling of care were included. Three models were constructed to compare in-hospital mortality by wave: a raw logistic regression model, a fully clinical adjusted logistic regression model and an inverse probability weighting logistic regression model.

Results

A total of 3982 patients without ceiling of care and 1831 patients with ceiling of care were included. The adjusted odds ratio (OR) of in-hospital mortality in the second wave were 0.57 (95%CI 0.40 to 0.80), in the third 0.56 (95%CI 0.37 to 0.84) and in the fourth 0.34 (95%CI 0.21 to 0.56) compared with the first wave in subjects without ceiling of care. The adjusted odds ratio were significantly lower in the fourth (0.38 95%CI 0.25 to 0.58) wave compared to the first wave in subjects with ceiling of care.

Conclusions

In patients without ceiling of care, mortality decreased over time suggesting better disease knowledge and management. In ceiling of care, only fourth-wave patients were less likely to die than first-wave patients. In a future infectious disease pandemic, it will be a challenge to improve the management of patients with ceiling of care.

Keywords

COVID-19, Infectious diseases, Palliative care, Epidemiology

Strengths and limitations of this study

- This is multicentric study with a large number of subjects included from four different waves of the COVID-19 pandemic.
- Several methods were used to compare in-hospital mortality between waves to increase the robustness of the estimated effects.
- Despite the inverse probability weighting analysis, there may be unobserved characteristics that lead to residual confounding.

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3 79 • The national vaccination campaign started for the elderly subjects before the
4 80 fourth wave so it could not be used in the adjustment analysis.
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INTRODUCTION

Despite the lack of definition in epidemiology, the term epidemic wave implies a natural pattern of peaks and troughs in the incidence of cases or hospitalisations due to an outbreak [1]. Epidemics often occur in local or global waves, each one with variations in severity or in transmission dynamics. For example, the 1918-1920 influenza pandemic was a global pandemic caused by the H1N1 virus. It is known to have occurred in three waves. The first wave (spring 1918) was relatively mild and resembled previous flu epidemics, the second wave (autumn 1918) was severe and much more deadly than the first. The third wave (1919) was less severe than the second but more deadly than the first [2]. In 1968, the Hong Kong flu was caused by the influenza A subtype H3N2 virus [3]. There were two waves in China (summer 1968 and June-December 1970) and, because the virus was highly contagious, it spread rapidly around the world reaching the United States and Europe by the end of December 1968. In most places, the second wave caused more deaths than the first [4].

Following a similar pattern, the COVID-19 pandemic began in Wuhan, China, in December 2019, and spread rapidly across Europe, with the first outbreak in Italy in February 2020. During the course of the pandemic, countries and regions experienced several waves with distinct peaks in cases. In Spain, 7 waves of the pandemic have been recorded between March 2020 and September 2023, with almost 14 million confirmed cases and more than 120.000 deaths [5]. Throughout this period, knowledge of the disease has progressively increased with the sequencing of the virus [6], clinical trials to assess treatments efficacy [7,8], the identification of different strains of the virus [9] and the development of vaccines [10]. All these factors, together with the natural immunity protection against COVID-19 [11], lead to a reduction in the need for hospitalisation, in-hospital mortality and complications.

In general, in a non-pandemic setting, decisions about the ceiling of care are common practice when dealing with patients with a critical prognosis and have implications for the use of life-sustaining measures such as intubation, mechanical ventilation, and cardiopulmonary resuscitation. However, in the peaks of the COVID-19 pandemic, decisions about the maximum level of care that each patient should receive, besides of the critical prognosis of the patient, were made in a scenery of emergency with excess demand for critical care and limited availability of clinical resources. Previously published data [12,13] suggest that COVID-19 hospitalised patients who had a ceiling of care were mainly older, had more comorbidities and higher incidence of in-hospital death. However, little is known about the impact of ceiling of care on mortality in hospitalised patients with COVID-19 across pandemic waves.

Our hypothesis is that in-hospital mortality should decrease over time as a result of increasing knowledge, natural immunity, the effect of new treatments and the introduction of vaccines. However, the role of the ceiling effect in this process has not been defined. The aim of this study was to compare in-hospital mortality across

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METHODS

Data source

The MetroSud study is an observational multicenter study conducted in five centers located in the southern metropolitan area of Barcelona (Catalonia, Spain), to characterise all patients with COVID-19 admitted to these hospitals during four waves of the pandemic from March 2020 to August 2021. Analised data of the first wave of COVID-19 pandemic embraced from March to April 2020, second wave from October to November 2020, third wave from January to February 2021 and fourth wave from July to August 2021. MetroSud cohort has been previously described [12].

An electronic case report form in REDCap [14] was designed in March 2020 to collect study data. Demographic data, comorbidities and other relevant findings on medical history, previous medications, clinical symptoms, vital signs, laboratory results and respiratory exploration were collected at baseline.

The presence or absence of ceiling of care was decided at the emergency room by the attending physicians according to their criteria, taking into account the patient's prognosis and the resources available (mainly intensive care unit (ICU) beds) at each participating hospital. Patients without a ceiling of care would have access to an ICU or could receive invasive mechanical ventilation (IMV). Otherwise, patients assigned to ceiling of care would have limited access to the ICU and, if they required any respiratory support, it would be non-rebreather mask, high-flow nasal cannula or NIMV. In terms of outcomes, the incidence of in-hospital mortality was defined as death by any cause during hospitalisation.

The study was approved by the Bellvitge Hospital Research Ethics Committee with medicines (CREm) in accordance with Spanish legislation and was performed in accordance with the Helsinki Declaration of 1964. The need for patient informed consent was waived by the ethics committee. Bellvitge's CREm decision was the basis for the approval of the remaining hospital centers.

Statistical methods

To describe cohort characteristics, categorical variables were presented as the number of cases and percentage, while continuous variables were expressed as the mean and standard deviation (SD) or median and interquartile range (IQR). All analyses were presented by wave and stratified by ceiling of care.

Once the variables to be used to match patients were identified, multiple imputation with chained equations (MICE) [15] was used to create five datasets with complete data. Missing data were assumed to be at random. Predictive mean matching was used to impute continuous variables and binomial logistic regression was used to impute binary variables. Information on age, sex and baseline comorbidities was used to impute missing values for obesity, body mass index (BMI), race, pneumonia severity Index (PSI), FiO₂, oxygen support, D-dimer, C-reactive protein, leukocytes,

haemoglobin and lymphocytes. Final estimates were adjusted for variability between the five imputed datasets according to the Rubin rules [16] to obtain the final model.

With the complete database, three models were constructed to study the association between in-hospital mortality and wave: 1) a crude logistic regression model using wave as a covariate, 2) a fully adjusted logistic regression model and 3) an inverse probability weighting (IPW) logistic regression model.

After discussion with clinicians, the variables included in the fully adjusted logistic regression model were baseline variables that define the patient's status at hospital admission: age, sex, race, BMI, obesity, long-term facility, comorbidities (diabetes mellitus, COPD, heart failure, hypertension, renal insufficiency, dyslipidemia, coronary heart disease, haematological neoplasm, solid neoplasm, organ transplantation, immunosuppressive treatment, chronic complex patient (PCC) and patients with advanced chronic disease (MACA), baseline laboratory values (dimer, C-reactive protein, leukocytes, haemoglobin, lymphocytes), pneumonia severity index (PSI), FiO₂ and oxygen support.

IPW [17] was used to adjust for differences in the patient baseline profile between waves. Bayesian additive regression trees, entropy balancing, generalised boosted models and generalised linear models were tested as methods for weighting individuals. In the end, we chose the method with better covariate balance between waves after weighting, which was the bayesian additive regression trees method [18]. In each imputed dataset, weights were calculated with the wave as the outcome and the variables used for the full adjusted logistic model as covariates.

To identify imbalances between waves after weighting, we estimated and described the standardised mean differences in baseline variables before and after weighting. We then fitted a logistic regression model for each imputation with in-hospital death as the outcome, using the stabilised weights and model-robust standard errors and adjusting for the variables that remained imbalanced between groups after weighting.

We used the STROBE cohort checklist [19] when writing our report. All analyses were performed with a two-sided significance level of 0.05 using R software version 4.3.0 [20]. The main R packages used for data management and analysis were flowchart [21], REDCapDM [22], mice [15], WeightIt [23], cobalt [24] and survey [25].

RESULTS

Flow chart

A total of 4417 patients without ceiling of care and 2159 patients with ceiling of care were included in the MetroSud. After excluding patients who were admitted to hospital for less than 24 hours, patients who died within the first 24 hours, patients with incomplete data on a pool of essential variables (age, sex, Charlson score, ceiling of care, and circumstances at discharge) or patients who were initially admitted to one hospital but transferred to another and treated in the latter, a total of 3982 patients without ceiling of care and a total of 1831 patients with ceiling of care were included in the analysis. All patients were followed up until in-hospital death or hospital discharge. (Figure 1, Flow Chart)

Baseline characteristics by wave

Table 1 describes the baseline characteristics of the included patients by wave and stratified by ceiling of care. Other variables included in the matching process are described in Supplementary Table 1.

TABLE 1: Patient’s most relevant characteristics according to wave and ceiling of care.

	No ceiling of care				Ceiling of care			
	Wave 1, N = 2076	Wave 2, N = 611	Wave 3, N = 605	Wave 4, N = 690	Wave 1, N = 1330	Wave 2, N = 175	Wave 3, N = 163	Wave 4, N = 163
Age	59 (49, 69)	62 (53, 71)	63 (53, 72)	49 (37, 63)	79 (72, 85)	83 (78, 88)	83 (78, 87)	85 (80, 89)
Sex								
Women	855 (41%)	222 (36%)	248 (41%)	242 (35%)	565 (42%)	75 (43%)	81 (50%)	75 (46%)
Race								
Caucasian	1206 (78%)	394 (69%)	464 (86%)	406 (65%)	795 (96%)	164 (95%)	154 (97%)	154 (98%)

	No ceiling of care				Ceiling of care			
	Wave 1, N = 2076	Wave 2, N = 611	Wave 3, N = 605	Wave 4, N = 690	Wave 1, N = 1330	Wave 2, N = 175	Wave 3, N = 163	Wave 4, N = 163
Other	341 (22%)	179 (31%)	73 (14%)	218 (35%)	32 (3.9%)	9 (5.2%)	4 (2.5%)	3 (1.9%)
Unknown	529	38	68	66	503	2	5	6
Charlson Index	2.00 (1.00, 3.00)	2.00 (1.00, 4.00)	3.00 (1.00, 4.00)	1.00 (0.00, 3.00)	5.00 (4.00, 7.00)	6.00 (5.00, 8.00)	6.00 (5.00, 7.00)	6.00 (5.00, 7.00)
PSI	62 (50, 79)	69 (56, 87)	66 (53, 84)	59 (46, 78)	97 (79, 123)	115 (95, 134)	103 (84, 124)	114 (96, 135)
Unknown	374	6	3	3	239	1	4	2

Median (IQR) for continuous variables; n (%) for categorical variables

PSI: Pneumonia severity index

Regarding age, patients with a ceiling of care were, in median, 20 years older than patients without a ceiling of care in all waves. There were no differences in the proportion of women. The most common race was Caucasian (in all waves, almost 90% of patients without ceiling of care and over 70% of patients with ceiling of care were Caucasian). Patients with a ceiling of care had a median Charlson Index more than 3 points higher than patients without a ceiling of care in all waves. PSI scores for patients with ceiling of care were more than 35 points higher in all waves (greater differences in wave 4) than PSI scores for patients without ceiling of care.

In-hospital mortality

The overall cumulative incidence of in-hospital mortality for patients with and without ceiling of care in all waves is shown in Table 2.

TABLE 2: Cumulative incidence and 95% confidence interval for in-hospital mortality according to wave and ceiling of care.

	Wave 1	Wave 2	Wave 3	Wave 4
No ceiling of care	10.50% [9.23 to 11.92]	10.15% [7.92 to 12.89]	7.60% [5.68 to 10.09]	5.22% [3.73 to 7.22]
Ceiling of care	37.07% [34.48 to 39.74]	40.00% [32.76 to 47.69]	44.79% [37.06 to 52.76]	30.06% [23.27 to 37.81]

About 1 in 10 patients without ceiling of care died in hospital in the first and second waves. In patients with a ceiling of care, about 4 in 10 patients die in hospital in the first three waves. The percentages are lower in the fourth wave (5% and 30% respectively for patients without and with a ceiling of care).

Mortality in patients without ceiling of care

Figure 2A shows the balance of covariates before and after IPW by means of the standardised mean differences (SMD) in patients without a ceiling of care. The SMD for PSI remains above 0.2. To correct for this imbalance, PSI was included in the weighted mortality models.

The odds ratios of the three models for mortality are shown in Figure 3A. The results with the three methods are consistent and show the same trend for all waves. Patients from waves 2, 3 and 4 were less likely to die in hospital than patients from wave 1 both in the raw models and in the models adjusted for covariates or adjusted with weights (OR for all models and all waves lower than 1). In addition, the value of the OR decreases across waves.

Mortality in patients with ceiling of care

Figure 2B shows the balance of covariates before and after IPW by means of the SMD in patients with a ceiling of care assigned at admission. Age, PSI and race showed a difference between waves greater than 0.2. These variables were included as adjustments in the weighted mortality model to account for these differences.

The odds ratios of the three models for mortality are shown in Figure 3B. No differences were found between 1st and 2nd wave patients or between 1st and 3rd wave patient (neither in the crude nor in the adjusted models). For wave 4th, both adjusted and IPW models showed that, given two patients with the same baseline profile, a wave 4 patient was less likely to die in hospital than a wave 1 patient.

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DISCUSSION

Our multicentre cohort study compared in-hospital mortality across COVID-19 waves between patients with and without a ceiling of care at hospital admission. We found that among patients without ceiling of care, those admitted in the first wave had worse in-hospital mortality than patients hospitalised during the other waves. Moreover, the adjusted models showed a significant decrease in mortality as the waves progressed. Among patients with a ceiling of care, no differences in in-hospital mortality were found between second and first wave patients, or between third and first wave patients. Only in the fourth wave, patients were less likely to die than first wave patients after adjustment for baseline risk. The magnitude of this effect on mortality reduction observed in patients with ceiling of care in the 4th wave was similar to the effect observed among patients without ceiling of care in the same 4th wave.

It is worth noting that if the differences in mortality between waves were only due to patient’s risk profile, the mortality rates would be similar after adjustment for baseline profile. However, this is not the case, as Figure 3A shows that in the adjusted and weighted models, mortality among patients without ceiling of care decreases as waves progress in time. The emergency situation experienced by the hospitals in the first months of the pandemic, with a lack of organization prepared to face an emergency such as COVID-19, partly explains the differences observed [26]. Besides, in the first wave, hospital resources (such as ICU beds, number of non-invasive ventilators or high-flow nasal oxygen therapy devices) and human resources were not sufficient to cope with the high demand for medical care [27]. ICU capacity is known to be an important indicator of hospital stress (health system resilience) which is associated with a reduction in quality of care and poorer patient outcomes [28]. In addition, other factors such as the increasing knowledge about the disease, facilitated by the rapid publication of clinical trials analysing new treatments [8], or the impact of public health surveillance measures, such as lockdowns [29] could explain this reduction in mortality. The harvest effect could also explain this decrease in mortality, as deaths that would have occurred anyway in subsequent waves may have been precipitated by the high mortality in the first wave of COVID-19 [30]. Similarly, the aggressiveness of SARS-CoV-2 varied between strains, and may also have played a role in the reduction in mortality [31].

As expected, mortality was higher among patients with ceiling care. In this group of patients, there are no differences in mortality in the first three waves, but there is a decrease in mortality in wave 4 (Figure 3B). In Spain, this fourth wave mainly affected young patients. Older patients, who were more likely to be assigned a ceiling of care, were already vaccinated at that time [32]. A study in nursing homes in our geographical area (Catalonia) [33] showed that vaccination was associated with a 95% reduction in mortality among nursing home residents. Studies in Italy and Switzerland also showed that the vaccine was about 95% effective against death in the general population [30,34]. These results therefore suggest that there is no improvement in medical management that affects in-hospital mortality until wave 4, which coincides with the elderly vaccination campaign. The lack of a contrafactual scenario in which people received intensive care makes it difficult to assess any potential benefit.

Further research on this topic and replication of these results in other cohorts would be needed.

The high probability of a new epidemic caused by an infectious organism merits in-depth reflection by the medical and scientific community, in particular to reach a consensus on the definition of ceiling of care and to define a guideline for the management of patients who are candidates for a ceiling of care [35]. In the event of a future pandemic caused by an infectious organism, the challenge will be to improve mortality in patients with ceiling of care. To this end, the scientific community needs to develop an action plan that will enable a rapid response in terms of both human resources (by increasing the number of trained health workers), and facilities (for example, so that the ICUs can quickly increase the number of beds) [36].

Our study has some limitations that should be acknowledged. One limitation is residual confounding. Even after using all the characteristics available at admission to make the baseline status of patients comparable, there may be unobserved characteristics that make patients different between waves. For example, we knew whether a patient had pathology or not, but we could not take into account how advanced it was. A variable that collects information on patients' frailty at baseline might also be of interest for a better risk assessment. In addition, vaccines and treatments could not be used in the matching: vaccines because they did not exist in the first wave [10] and treatments because they changed drastically between waves due to increasing knowledge about the disease [7,8]. Another limitation of the study is that we assumed that the missing values in our data were at random and imputed them using standard techniques. To account for this, the analyses were repeated only with patients who had complete information on all variables, and the results were in the same line, confirming the robustness of the analysis. Moreover, we cannot guarantee that the same criteria were used to define the therapeutic ceiling of care in all hospitals. In fact, one of the challenges in clinical practice during the COVID-19 pandemic was to define the ceiling of care for infected patients.

The strengths of our study are the large number of subjects included from different hospitals and from four different waves of the pandemic, and the availability of information on ceiling of care. In addition, the different methods used to compare in-hospital mortality by waves led to the same results, demonstrating the robustness of the analysis.

In conclusion, in-hospital mortality was not homogeneous between waves in patients with and without ceiling of care. In patients without ceiling of care, mortality decreased over time suggesting better management and knowledge of the disease. In patients with ceiling of care, mortality remained constant, except in the last wave. In the event of a future pandemic caused by an infectious organism, it will be a challenge to harmonize and improve the clinical criteria and management of patients who might be assigned a ceiling of care.

FIGURE LEGENDS

FIGURE 1: Flow-chart of the included patients without ceiling of care (left) and with ceiling of care (right).

FIGURE 2: Maximum standardized mean differences (SMD) before (Unmatched) and after weighting (Matched) across waves for patients without a ceiling of care (A) and patients with ceiling of care (B). The standardized mean difference compares the difference in means between all pairs of waves in standard deviation units.

FIGURE 3: OR for raw, adjusted and IPTW models for in-hospital mortality in patients without a ceiling of care (A) and with ceiling of care (B).

DECLARATIONS

Ethics approval and consent to participate

The study was approved by the Bellvitge Hospital Research Ethics Committee with medicines (CREm) in accordance with Spanish legislation and was performed in accordance with the Helsinki Declaration of 1964. The need for patient informed consent was waived by the ethics committee. Bellvitge's CREm decision was the basis for the approval of the remaining hospital centres.

Competing interests

Cristian Tebé has received fees for speaker lectures and talks from Gedeon Richter, outside the submitted work. The rest of authors declare that they have no competing interests.

Data availability

The datasets generated during and/or analyzed during the current study are available from the corresponding author on reasonable request.

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Author's contributions

Conceptual design was performed by SV, JCa, CT, and NP.

MetroSud cohort data was provided by SV and JCa.

Statistical analysis was performed by CT and NP.

The first draft of the manuscript was written by NP and revised by JCa and CT.

All authors commented on previous versions of the manuscript.

All authors read and approved the final version of the manuscript.

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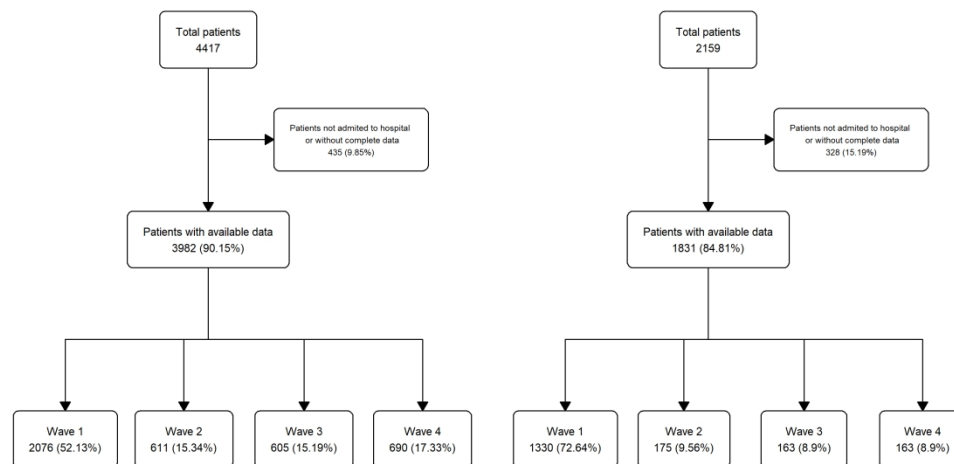


FIGURE 1: Flow-chart of the included patients without ceiling of care (left) and with ceiling of care (right).

249x199mm (300 x 300 DPI)

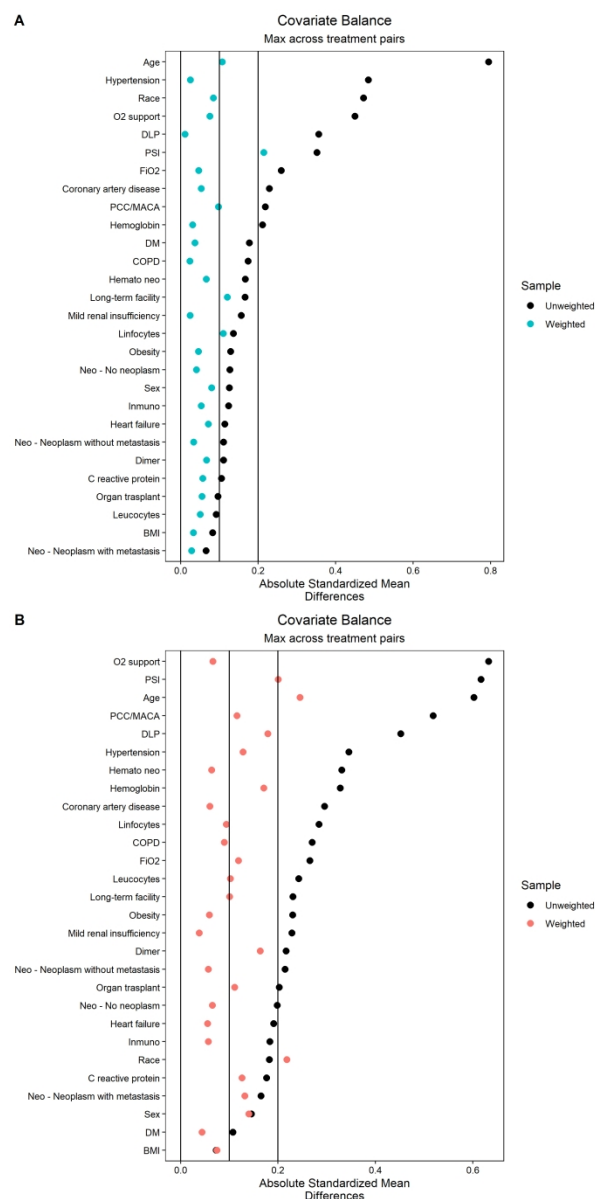


FIGURE 2: Maximum standardized mean differences (SMD) before (Unmatched) and after weighting (Matched) across waves for patients without a ceiling of care (A) and patients with ceiling of care (B). The standardized mean difference compares the difference in means between all pairs of waves in standard deviation units.

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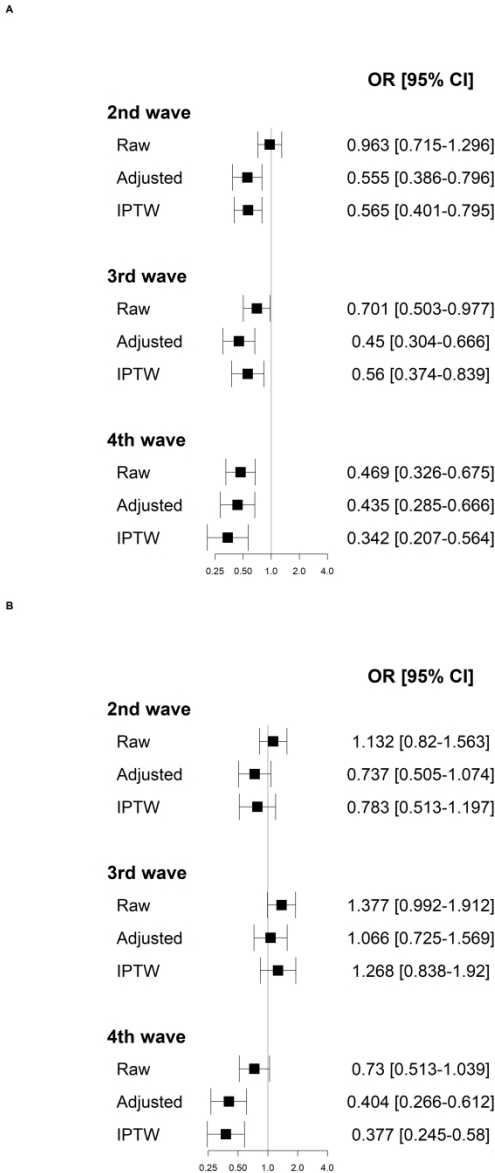


FIGURE 3: OR for raw, adjusted and IPTW models for in-hospital mortality in patients without a ceiling of care (A) and with ceiling of care (B).

299x599mm (300 x 300 DPI)

SUPPLEMENTAL MATERIAL

Table S1: Variables used in the matching procedure according to wave and ceiling of care

	No ceiling of care				Ceiling of care			
	Wave 1, N = 2076	Wave 2, N = 611	Wave 3, N = 605	Wave 4, N = 690	Wave 1, N = 1330	Wave 2, N = 175	Wave 3, N = 163	Wave 4, N = 163
Long-term facility								
Yes	64 (3.1%)	19 (3.1%)	17 (2.8%)	4 (0.6%)	223 (17%)	20 (11%)	20 (12%)	15 (9.2%)
BMI	28.9 (25.9, 32.2)	29.2 (26.4, 32.6)	29.4 (26.6, 32.9)	29.7 (26.4, 33.6)	28.1 (25.4, 31.3)	28.7 (25.8, 31.5)	29.3 (26.0, 31.5)	27.6 (24.8, 31.8)
Unknown	636	103	72	177	456	22	19	17
PCC/MACA								
PCC/MACA	93 (4.5%)	32 (5.2%)	15 (2.5%)	9 (1.3%)	290 (22%)	80 (46%)	47 (29%)	57 (35%)
Diabetes mellitus								
Yes	418 (20%)	124 (20%)	126 (21%)	96 (14%)	414 (31%)	61 (35%)	54 (33%)	59 (36%)
COPD								
Yes	274 (13%)	108 (18%)	119 (20%)	110 (16%)	325 (24%)	54 (31%)	49 (30%)	60 (37%)
Heart failure								
Yes	50 (2.4%)	18 (2.9%)	27 (4.5%)	25 (3.6%)	194 (15%)	37 (21%)	31 (19%)	36 (22%)
Hypertension								

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	No ceiling of care				Ceiling of care			
	Wave 1, N = 2076	Wave 2, N = 611	Wave 3, N = 605	Wave 4, N = 690	Wave 1, N = 1330	Wave 2, N = 175	Wave 3, N = 163	Wave 4, N = 163
Yes	792 (38%)	266 (44%)	295 (49%)	176 (26%)	881 (66%)	137 (78%)	124 (76%)	132 (81%)
Obesity								
Yes	579 (35%)	221 (36%)	248 (41%)	248 (36%)	285 (29%)	51 (29%)	65 (40%)	50 (31%)
Unknown	404	0	0	0	361	0	0	0
Dyslipidemia								
Yes	698 (34%)	223 (36%)	239 (40%)	158 (23%)	502 (38%)	105 (60%)	92 (56%)	83 (51%)
Mild renal insufficiency								
Yes	83 (4.0%)	27 (4.4%)	42 (6.9%)	25 (3.6%)	234 (18%)	42 (24%)	26 (16%)	41 (25%)
Coronary artery disease								
Yes	91 (4.4%)	33 (5.4%)	28 (4.6%)	7 (1.0%)	112 (8.4%)	26 (15%)	30 (18%)	19 (12%)
Haematological neoplasm								
Yes	12 (0.6%)	16 (2.6%)	7 (1.2%)	12 (1.7%)	27 (2.0%)	8 (4.6%)	7 (4.3%)	15 (9.2%)
Organ transplant								
Yes	20 (1.0%)	13 (2.1%)	6 (1.0%)	13 (1.9%)	12 (0.9%)	0 (0%)	3 (1.8%)	1 (0.6%)

	No ceiling of care				Ceiling of care			
	Wave 1, N = 2076	Wave 2, N = 611	Wave 3, N = 605	Wave 4, N = 690	Wave 1, N = 1330	Wave 2, N = 175	Wave 3, N = 163	Wave 4, N = 163
Immunology								
Yes	72 (3.5%)	32 (5.2%)	17 (2.8%)	32 (4.6%)	50 (3.8%)	6 (3.4%)	11 (6.7%)	5 (3.1%)
Neoplasm								
No neoplasm	1991 (96%)	578 (95%)	563 (93%)	656 (95%)	1160 (87%)	141 (81%)	136 (83%)	130 (80%)
Neoplasm without metastasis	78 (3.8%)	30 (4.9%)	37 (6.1%)	30 (4.3%)	145 (11%)	26 (15%)	22 (13%)	30 (18%)
Neoplasm with metastasis	7 (0.3%)	3 (0.5%)	5 (0.8%)	4 (0.6%)	25 (1.9%)	8 (4.6%)	5 (3.1%)	3 (1.8%)
D-dimer	570 (316, 1050)	530 (284, 970)	500 (266, 895)	365 (250, 690)	722 (378, 1608)	689 (356, 1438)	471 (280, 969)	451 (276, 895)
Unknown	488	55	62	59	384	19	29	19
C reactive protein	80 (34, 149)	84 (39, 143)	76 (39, 128)	85 (41, 144)	92 (47, 160)	86 (41, 144)	96 (44, 148)	69 (30, 155)
Unknown	161	30	54	36	102	6	24	8
Haemoglobin	13.90 (12.90, 14.90)	13.60 (12.50, 14.80)	13.90 (12.80, 15.00)	14.05 (13.10, 15.10)	13.30 (12.00, 14.43)	12.50 (11.00, 14.40)	12.80 (11.60, 13.60)	12.50 (11.20, 13.80)
Unknown	150	21	40	24	82	6	15	5
Lymphocytes	0.98 (0.72, 1.33)	0.91 (0.66, 1.26)	0.91 (0.64, 1.23)	0.93 (0.66, 1.24)	0.90 (0.63, 1.24)	0.85 (0.57, 1.18)	0.80 (0.54, 1.13)	0.90 (0.59, 1.36)
Unknown	137	21	40	28	105	6	15	5

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	No ceiling of care				Ceiling of care			
	Wave 1, N = 2076	Wave 2, N = 611	Wave 3, N = 605	Wave 4, N = 690	Wave 1, N = 1330	Wave 2, N = 175	Wave 3, N = 163	Wave 4, N = 163
Leucocytes	6.5 (5.0, 8.7)	6.5 (5.0, 9.0)	6.4 (5.0, 8.6)	6.2 (4.6, 8.6)	6.9 (5.2, 9.2)	7.2 (5.4, 9.4)	6.5 (5.0, 8.7)	6.9 (5.4, 9.6)
Unknown	109	23	40	26	71	6	15	6
FiO2	21 (21, 21)	21 (21, 28)	21 (21, 28)	21 (21, 31)	21 (21, 24)	21 (21, 28)	21 (21, 35)	24 (21, 31)
Unknown	2	0	0	0	3	0	0	0
Need for oxygen support	511 (25%)	193 (32%)	234 (39%)	316 (46%)	345 (26%)	69 (39%)	77 (47%)	92 (56%)
Unknown	2	0	0	0	3	0	0	0

Median (IQR) for continuous variables; n (%) for categorical variables

BMI: Body mass index

COPD: Chronic obstructive pulmonary disease

PCC: chronic complex patient

MACA: advanced chronic disease patient

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Impact of ceiling of care on mortality across four COVID-19 epidemic waves in Catalonia: a multicentre prospective cohort study

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Impact of ceiling of care on mortality across four COVID-19 epidemic waves in Catalonia: a multicentre prospective cohort study

Pallarès N^{1,2} MSc, Videla S^{3,4} MD PhD, Carratalà J^{5,6,7,8} MD PhD, Tebé C¹ MPH PhD, on behalf of the MetroSud study group and the Divine study group

1) Biostatistics Support and Research Unit, Germans Trias i Pujol Research Institute and Hospital (IGTP), Badalona, Barcelona, Spain

2) Department of Basic Clinical Practice, School of Medicine and Health Sciences, University of Barcelona, Spain

3) Clinical Research Support Area, Department of Clinical Pharmacology, Germans Trias i Pujol University Hospital, Badalona, Spain

4) Department of Pathology and Experimental Therapeutics, School of Medicine and Health Sciences, University of Barcelona, Spain

5) Department of Infectious Diseases, Bellvitge University Hospital, Barcelona, Spain

6) Bellvitge Biomedical Research Institute (IDIBELL), Barcelona, Spain

7) Centro de Investigación en Red de Enfermedades Infecciosas (CIBERINFEC), Instituto de Salud Carlos III, Madrid, Spain.

8) Department of Clinical Sciences, School of Medicine and Health Sciences, University of Barcelona, Spain

MetroSud Study group: Gabriela Abelenda-Alonso, Alexander Rombauts, Isabel Oriol, Antonella F. Simonetti, Alejandro Rodríguez-Molinero, Elisenda Izquierdo, Vicens Díaz-Brito, Carlota Gudíol, Judit Aranda-Lobo, Marta Arroyo, Carlos Pérez-López, Montserrat Sanmartí, Encarna Moreno, Maria C Alvarez, Ana Faura, Martha González, Paula Cruz, Mireia Colom, Andrea Perez, Laura Serrano.

DIVINE Study group: Mireia Besalú, Erik Cobo, Jordi Cortés, Daniel Fernández, Leire Garmendia, Guadalupe Gómez, Pilar Hereu, Klaus Langohr, Gemma Molist, Núria Pérez-Álvarez, Xavier Piulachs.

Corresponding author

Natàlia Pallarès, MSc
Biostatistics Support and Research Unit
Germans Trias i Pujol Research Institute and Hospital (IGTP)
Campus Can Ruti. Carretera de Can Ruti, Camí de les Escoles s/n. 08916 Badalona, Barcelona, Spain

ABSTRACT

Objective: The aim of this study was to compare in-hospital mortality across waves in patients without and with a ceiling of care at hospital admission.

Design : A multicentre prospective cohort study

Setting: Five tertiary hospitals in Catalonia, Spain, during four waves of the COVID-19 pandemic. Data from the first wave embraced from March to April 2020, second wave from October to November 2020, third wave from January to February 2021 and fourth wave from July to August 2021.

Participants: All consecutive adult subjects (older than 18 years old) admitted to any of the five aforementioned centers. All subjects had a confirmed SARS-CoV-2 infection (with a positive PCR test or antigen test) and an overnight hospital stay. Ceiling of care defined as the highest level of care that a patient will receive during medical treatment was assessed at hospital admission for all patients.

Primary measure: In-hospital mortality

Results: A total of 3982 hospitalized patients without ceiling of care and 1831 hospitalized patients with ceiling of care were included in the analysis. The adjusted odds ratio (OR) of in-hospital mortality in the second wave were 0.57 (95%CI 0.40 to 0.80), in the third 0.56 (95%CI 0.37 to 0.84) and in the fourth 0.34 (95%CI 0.21 to 0.56) compared with the first wave in subjects without ceiling of care. The adjusted odds ratio were significantly lower in the fourth (0.38 95%CI 0.25 to 0.58) wave compared to the first wave in subjects with ceiling of care.

Conclusions: In patients without ceiling of care, mortality decreased over time suggesting better disease knowledge and management. In ceiling of care, only fourth-wave patients were less likely to die than first-wave patients. In a future infectious disease pandemic, it will be a challenge to improve the management of patients with ceiling of care.

Keywords

COVID-19, Infectious diseases, Palliative care, Epidemiology

Strengths and limitations of this study

- This is multicentric study with a large number of subjects included from four different waves of the COVID-19 pandemic.
- Several methods were used to compare in-hospital mortality between waves to increase the robustness of the estimated effects.
- Despite the inverse probability weighting analysis, there may be unobserved characteristics that lead to residual confounding.

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- 82 • The national vaccination campaign started for the elderly subjects before the
83 fourth wave so it could not be used in the adjustment analysis.

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INTRODUCTION

Despite the lack of definition in epidemiology, the term epidemic wave implies a natural pattern of peaks and troughs in the incidence of cases or hospitalizations due to an outbreak [1]. Epidemics often occur in local or global waves, each one with variations in severity or in transmission dynamics [2–4].

Following a similar pattern, the COVID-19 pandemic began in Wuhan, China, in December 2019, and spread rapidly across Europe, with the first outbreak in Italy in February 2020. During the course of the pandemic, countries and regions experienced several waves with distinct peaks in cases. In Spain, 7 waves of the pandemic have been recorded between March 2020 and September 2023, with almost 14 million confirmed cases and more than 120.000 deaths [5]. Throughout this period, knowledge of the disease has progressively increased with the sequencing of the virus [6], clinical trials to assess treatments efficacy [7,8], the identification of different strains of the virus [9] and the development of vaccines [10]. All these factors, together with the natural immunity protection against COVID-19 [11], lead to a reduction in the need for hospitalization, in-hospital mortality and complications.

The therapeutic ceiling of care refers to the highest level of care that a patient will receive during medical treatment. In general, in a non-pandemic setting, decisions about the ceiling of care are common practice when dealing with patients with a critical prognosis and have implications for the use of life-sustaining measures such as intubation, mechanical ventilation, and cardiopulmonary resuscitation. However, in the peaks of the COVID-19 pandemic, decisions about the maximum level of care that each patient should receive, besides of the critical prognosis of the patient, were made in a scenery of emergency with excess demand for critical care and limited availability of clinical resources. Previously published data [12,13] suggest that COVID-19 hospitalized patients who had a ceiling of care were mainly older, had more comorbidities and higher incidence of in-hospital death. In-hospital mortality has been shown to decrease over time [14,15]. However, little is known about the impact of ceiling of care on mortality in hospitalized patients with COVID-19 across pandemic waves. Stratifying by care limitations helps to distinguish whether the reduction in mortality was due to advances in intensive care unit management, improved general hospital care, or shifts in decision making. This approach addresses a gap in previous research, which has often overlooked how changes in patient selection for intensive care can bias mortality trends. Understanding these dynamics can inform clinical decision-making and ensure optimal management for all patients, regardless of their care limitations.

Our hypothesis is that the decrease in in-hospital mortality over time is different in patients with and without ceiling of care. The aim of this study was to compare in-hospital mortality across four COVID-19 waves between patients with and without a ceiling of care at hospital admission.

METHODS

Study design and setting

The MetroSud study is an observational multicenter study conducted in five centers located in the southern metropolitan area of Barcelona (Catalonia, Spain), to characterise all patients with COVID-19 admitted to these hospitals during four waves of the pandemic from March 2020 to August 2021. Analyzed data of the first wave of COVID-19 pandemic embraced from March to April 2020, second wave from October to November 2020, third wave from January to February 2021 and fourth wave from July to August 2021 [16]. MetroSud cohort has been previously described [12].

Eligibility criteria

The MetroSud cohort included all consecutive adult subjects (older than 18 years old) admitted to any of the five aforementioned centers. All subjects had a proven SARS-CoV-2 infection (with a positive PCR test or antigen test).

Data sources and study variables

An electronic case report form in REDCap [17] was designed in March 2020 to collect study data. Demographic data (age, sex, race), comorbidities and other relevant findings on medical history, previous medications, clinical symptoms, vital signs (body temperature, FiO₂, O₂ saturation, blood pressure, pulse, and respiratory rate), laboratory results (D dimer, C-reactive protein, lactat dehydrogenase, leukocytes, and others) and respiratory exploration (wheezing, rhoncus), Pneumonia severity index (PSI) and ceiling of care were collected at baseline by the attending physicians.

The presence or absence of ceiling of care was decided at the emergency room by the attending physicians according to their criteria, taking into account the patient's potential benefit of intensive treatments. In the beginning of the first wave, due to the ICU demand and capacity, the availability of resources at each participating hospital was also taken into account. Patients without a ceiling of care would have access to an ICU or could receive invasive mechanical ventilation (IMV). Otherwise, patients assigned to ceiling of care would have limited access to the ICU and, if they required any respiratory support, it would be non-rebreather mask, high-flow nasal cannula or NIMV.

Outcome variable

The outcome variable was in-hospital mortality defined as death by any cause during hospitalization and was registered in the electronic case report form.

The study was approved by the Bellvitge Hospital Research Ethics Committee with medicines (CREm), with reference PR140/20 and code HUB-INF-COHORT-HUB-COVID, in accordance with Spanish legislation and was performed in accordance with the Helsinki Declaration of 1964. The need for patient informed consent was waived by the

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3 171 ethics committee. Bellvitge's CREm decision was the basis for the approval of the
4 172 remaining hospital centers.
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7 174 **Statistical methods**
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10 175 To describe cohort characteristics, categorical variables were presented as the number
11 176 of cases and percentage, while continuous variables were expressed as the mean and
12 177 standard deviation (SD) or median and interquartile range (IQR). All analyses were
13 178 presented by wave and stratified by ceiling of care.
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15 179 A pool of essential variables to describe the baseline profile of patients was defined.
16 180 This pool included age, sex, Charlson score, ceiling of care, and circumstances at
17 181 discharge. Patients who had incomplete data on this pool of variables were discarded
18 182 from the analysis.
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21 183 Once the variables to be used to match patients were identified, multiple imputation
22 184 with chained equations (MICE) [18] was used to create five datasets with complete
23 185 data. Missing data were assumed to be at random. Predictive mean matching was used
24 186 to impute continuous variables and binomial logistic regression was used to impute
25 187 binary variables. Information on age, sex and baseline comorbidities (completed for all
26 188 patients after exclusions) was used to impute missing values for obesity, body mass
27 189 index (BMI), race, pneumonia severity Index (PSI), FiO₂, oxygen support, D-dimer, C-
28 190 reactive protein, leukocytes, haemoglobin and lymphocytes. Final estimates were
29 191 adjusted for variability between the five imputed datasets according to the Rubin rules
30 192 [19] to obtain the final model.
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33 193 With the database with all the missing data imputed, three models were constructed
34 194 to study the association between in-hospital mortality and wave: 1) a crude logistic
35 195 regression model using wave as a covariate, 2) a fully adjusted logistic regression
36 196 model and 3) an inverse probability weighting (IPW) logistic regression model.
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38 197 After discussion with clinicians, the variables included in the fully adjusted logistic
39 198 regression model to minimize confounding and make patients comparable between
40 199 waves were baseline variables that define the patient's status at hospital admission:
41 200 age, sex, race, BMI, obesity, long-term facility, comorbidities (diabetes mellitus, COPD,
42 201 heart failure, hypertension, renal insufficiency, dyslipidemia, coronary heart disease,
43 202 haematological neoplasm, solid neoplasm, organ transplantation, immunosuppressive
44 203 treatment, chronic complex patient (PCC) and patients with advanced chronic disease
45 204 (MACA), baseline laboratory values (dimer, C-reactive protein, leukocytes,
46 205 haemoglobin, lymphocytes), pneumonia severity index (PSI), FiO₂ and oxygen support.
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48 206 IPW [20] was used to adjust for differences in the patient baseline profile between
49 207 waves. Bayesian additive regression trees, entropy balancing, generalised boosted
50 208 models and generalised linear models were tested as methods for weighting
51 209 individuals. In the end, we chose the method with better covariate balance between
52 210 waves after weighting, which was the bayesian additive regression trees method [21].
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In each imputed dataset, weights were calculated with the wave as the outcome and the variables used for the full adjusted logistic model as covariates.

To identify imbalances between waves after weighting, we estimated and described the standardised mean differences in baseline variables before and after weighting. We then fitted a logistic regression model for each imputation with in-hospital death as the outcome, using the stabilised weights and model-robust standard errors and adjusting for the variables that remained imbalanced between groups after weighting.

To overcome the limitation of assuming missing at random, a sensitivity analysis was performed by repeating the analyses using only those patients who had complete information on all variables.

We used the STROBE cohort checklist [22] when writing our report. All analyses were performed with a two-sided significance level of 0.05 using R software version 4.3.0 [23]. The main R packages used for data management and analysis were flowchart [24], REDCapDM [25], mice [18], WeightIt [26], cobalt [27] and survey [28].

Patient and public involvement

There was no patient or public involvement in the development of the research design or in conducting the study.

RESULTS

Flow chart

A total of 4417 patients without ceiling of care and 2159 patients with ceiling of care were included in the MetroSud. After excluding patients who were admitted to hospital for less than 24 hours, patients who died within the first 24 hours, patients with incomplete data on a pool of essential variables (age, sex, Charlson score, ceiling of care, and circumstances at discharge) or patients who were initially admitted to one hospital but transferred to another and treated in the latter, a total of 3982 patients without ceiling of care and a total of 1831 patients with ceiling of care were included in the analysis. All patients were followed up until in-hospital death or hospital discharge. (Figure 1, Flow Chart)

Baseline characteristics by wave

Table 1 describes the baseline characteristics of the included patients by wave and stratified by ceiling of care. Other variables included in the matching process are described in Supplementary Table 1.

TABLE 1: Patient’s most relevant characteristics according to wave and ceiling of care.

	No ceiling of care				Ceiling of care			
			Wave 4					
	Wave 1	Wave 2	Wave 3	(July-Aug	Wave 1	Wave 2	Wave 3	Wave 4
	(Mar-Apr	(Oct-Nov	(Jan-Feb	2021)	(Mar-Apr	(Oct-Nov	(Jan-Feb	(July-Aug
	2020)	2020)	2021)	2021)	2020)	2020)	2021)	2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Age	59 (49, 69)	62 (53, 71)	63 (53, 72)	49 (37, 63)	79 (72, 85)	83 (78, 88)	83 (78, 87)	85 (80, 89)
Sex								
Women	855 (41%)	222 (36%)	248 (41%)	242 (35%)	565 (42%)	75 (43%)	81 (50%)	75 (46%)
Race								

	No ceiling of care				Ceiling of care			
	Wave 4							
	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	(July- Aug 2021)	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Caucasian	1206 (78%)	394 (69%)	464 (86%)	406 (65%)	795 (96%)	164 (95%)	154 (97%)	154 (98%)
Other	341 (22%)	179 (31%)	73 (14%)	218 (35%)	32 (3.9%)	9 (5.2%)	4 (2.5%)	3 (1.9%)
Unknown	529	38	68	66	503	2	5	6
Charlson Index	2.00 (1.00, 3.00)	2.00 (1.00, 4.00)	3.00 (1.00, 4.00)	1.00 (0.00, 3.00)	5.00 (4.00, 7.00)	6.00 (5.00, 8.00)	6.00 (5.00, 7.00)	6.00 (5.00, 7.00)
PSI	62 (50, 79)	69 (56, 87)	66 (53, 84)	59 (46, 78)	97 (79, 123)	115 (95, 134)	103 (84, 124)	114 (96, 135)
Unknown	374	6	3	3	239	1	4	2

Median (IQR) for continuous variables; n (%) for categorical variables

PSI: Pneumonia severity index

Regarding age, patients with a ceiling of care were, in median, 20 years older than patients without a ceiling of care in all waves. There were no differences in the proportion of women. The most common race was Caucasian (in all waves, almost 90% of patients without ceiling of care and over 70% of patients with ceiling of care were Caucasian). Patients with a ceiling of care had a median Charlson Index more than 3 points higher than patients without a ceiling of care in all waves. PSI scores for patients with ceiling of care were more than 35 points higher in all waves (greater differences in wave 4) than PSI scores for patients without ceiling of care.

In-hospital mortality

The overall cumulative incidence of in-hospital mortality for patients with and without ceiling of care in all waves is shown in Table 2.

TABLE 2: Cumulative incidence and 95% confidence interval for in-hospital mortality according to wave and ceiling of care.

	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
No ceiling of care	10.50% [9.23 to 11.92]	10.15% [7.92 to 12.89]	7.60% [5.68 to 10.09]	5.22% [3.73 to 7.22]
Ceiling of care	37.07% [34.48 to 39.74]	40.00% [32.76 to 47.69]	44.79% [37.06 to 52.76]	30.06% [23.27 to 37.81]

About 1 in 10 patients without ceiling of care died in hospital in the first and second waves. In patients with a ceiling of care, about 4 in 10 patients die in hospital in the first three waves. The percentages are lower in the fourth wave (5% and 30% respectively for patients without and with a ceiling of care).

Mortality in patients without ceiling of care

Figure 2A shows the balance of covariates before and after IPW by means of the standardised mean differences (SMD) in patients without a ceiling of care. The SMD for PSI remains above 0.2. To correct for this imbalance, PSI was included in the weighted mortality models.

The odds ratios of the three models for mortality are shown in Figure 3A. The results with the three methods are consistent and show the same trend for all waves. Patients from waves 2, 3 and 4 were less likely to die in hospital than patients from wave 1 both in the raw models and in the models adjusted for covariates or adjusted with weights (OR for all models and all waves lower than 1). In addition, the value of the OR decreases across waves.

Mortality in patients with ceiling of care

Figure 2B shows the balance of covariates before and after IPW by means of the SMD in patients with a ceiling of care assigned at admission. Age, PSI and race showed a difference between waves greater than 0.2. These variables were included as adjustments in the weighted mortality model to account for these differences.

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The odds ratios of the three models for mortality are shown in Figure 3B. No differences were found between 1st and 2nd wave patients or between 1st and 3rd wave patient (neither in the crude nor in the adjusted models). For wave 4th, both adjusted and IPW models showed that, given two patients with the same baseline profile, a wave 4 patient was less likely to die in hospital than a wave 1 patient.

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DISCUSSION

Our multicentre cohort study compared in-hospital mortality across COVID-19 waves between patients with and without a ceiling of care at hospital admission. We found that among patients without ceiling of care, those admitted in the first wave had worse in-hospital mortality than patients hospitalized during the other waves. Moreover, the adjusted models showed a significant decrease in mortality as the waves progressed. Among patients with a ceiling of care, no differences in in-hospital mortality were found between second and first wave patients, or between third and first wave patients. Only in the fourth wave, patients were less likely to die than first wave patients after adjustment for baseline risk. The magnitude of this effect on mortality reduction observed in patients with ceiling of care in the 4th wave was similar to the effect observed among patients without ceiling of care in the same 4th wave.

It is worth noting that if the differences in mortality between waves were only due to patient's risk profile, the mortality rates would be similar after adjustment for baseline profile. However, this is not the case, as Figure 3A shows that in the adjusted and weighted models, mortality among patients without ceiling of care decreases as waves progress in time (OR decreasing from 0.56 (second and third wave) to 0.34 (fourth wave) when comparing with first wave)). The emergency situation experienced by the hospitals in the first months of the pandemic, with a lack of organization prepared to face an emergency such as COVID-19, partly explains the differences observed [29]. Besides, in the first wave, hospital resources (such as ICU beds, number of non-invasive ventilators or high-flow nasal oxygen therapy devices) and human resources were not sufficient to cope with the high demand for medical care [30]. ICU capacity is known to be an important indicator of hospital stress (health system resilience) which is associated with a reduction in quality of care and poorer patient outcomes [31]. In addition, other factors such as the increasing knowledge about the disease, facilitated by the rapid publication of clinical trials analysing new treatments [8], or the impact of public health surveillance measures, such as lockdowns [32] could explain this reduction in mortality. The harvest effect could also explain this decrease in mortality, as deaths that would have occurred anyway in subsequent waves may have been precipitated by the high mortality in the first wave of COVID-19 [33]. Similarly, the aggressiveness of SARS-CoV-2 varied between strains, and may also have played a role in the reduction in mortality [34].

As expected, mortality was higher among patients with ceiling care. In this group of patients, there are no differences in mortality in the first three waves, but there is a decrease in mortality in wave 4 (OR 0.38 95%CI 0.25 to 0.58) (Figure 3B). In Spain, this fourth wave mainly affected young patients. Older patients, who were more likely to be assigned a ceiling of care, were already vaccinated at that time [35]. A study in nursing homes in our geographical area (Catalonia) [36] showed that vaccination was associated with a 95% reduction in mortality among nursing home residents. Studies in Italy and Switzerland also showed that the vaccine was about 95% effective against death in the general population [33,37]. These results therefore suggest that there is no improvement in medical management that affects in-hospital mortality until wave 4, which coincides with the elderly vaccination campaign. The lack of a contrafactual

scenario in which people received intensive care makes it difficult to assess any potential benefit. Further research on this topic and replication of these results in other cohorts would be needed. Moreover, it will be of interest to study the management of ceiling of care in other cultural settings. It would also be interesting to investigate whether the impact of ceiling of care is the same on other outcomes, such as complications or length of hospital stay.

The high probability of a new epidemic caused by an infectious organism merits in-depth reflection by the medical and scientific community, in particular to reach a consensus on the definition of ceiling of care and to define a guideline for the management of patients who are candidates for a ceiling of care [38]. In the event of a future pandemic caused by an infectious organism, the challenge will be to improve mortality in patients with ceiling of care. To this end, the scientific community needs to develop an action plan that will enable a rapid response in terms of both human resources (by increasing the number of trained health workers), and facilities (for example, so that the ICUs can quickly increase the number of beds) [39].

Our study has some limitations that should be acknowledged. One limitation is residual confounding. Even after using all the characteristics available at admission to make the baseline status of patients comparable, there may be unobserved characteristics that make patients different between waves. For example, we knew whether a patient had pathology or not, but we could not take into account how advanced it was. A variable that collects information on patients' frailty at baseline might also be of interest for a better risk assessment. In addition, vaccines and treatments could not be used in the matching: vaccines because they did not exist in the first wave [10] and treatments because they changed drastically between waves due to increasing knowledge about the disease [7,8]. Moreover, we do not have data on the follow-up of patients with regard to treatments received during hospitalization, which could help to understand some of the differences in mortality. Another limitation of the study is that we assumed that the missing values in our data were at random and imputed them using standard techniques. To account for this, a sensitivity analysis was performed repeating the analysis only with patients who had complete information on all variables, and the results were in the same line, confirming the robustness of the analysis. Moreover, we cannot guarantee that the same criteria were used to define the therapeutic ceiling of care in all hospitals. In fact, one of the challenges in clinical practice during the COVID-19 pandemic was to define the ceiling of care for infected patients. Even though the definition of ceiling of care is not a standardized one, the definition in the MetroSud cohort was a pragmatic one which would be readable and understood by clinician teams involved in reaching these decisions. Our definition is consistent with that used in the Leeds cohort [13] and with the one used in a multicentre study to identify factors influencing ceiling of treatment in an Emergency Department [40]. In addition, when the study protocol was written, little was known about COVID-19, including the lack of immunity and the possibility of reinfection. Before the emergence of the Omicron strain, the incidence of COVID-19 reinfection leading to hospitalization was very low (<1%) [41]. Our last wave included patients from July to August 2021, when the Omicron strain had not yet reach Spain and the incidence of reinfection was still very low. However, we could not rule out the possibility that some subjects from the 3rd or 4th wave had previously been included

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in the study. Due to data protection laws, we did not have access to patients' clinical records, which prevented verification. Nevertheless, participating physicians were aware of this possibility and took measures to avoid case duplication.

The strengths of our study are the large number of subjects included from different hospitals and from four different waves of the pandemic, and the availability of information on ceiling of care. In addition, the different methods used to compare in-hospital mortality by waves led to the same results, demonstrating the robustness of the analysis.

In conclusion, knowing that the evolution of in-hospital mortality through waves is different in patients with and without ceiling of care could help the scientific community to address the management of patients with ceiling of care to improve their outcomes in a new pandemic scenario. The lessons learned from the COVID-19 pandemic could help health-care professional and health policy-makers to face future pandemics.

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FIGURE LEGENDS

FIGURE 1: Flow-chart of the included patients without ceiling of care (left) and with ceiling of care (right).

FIGURE 2: Maximum standardized mean differences (SMD) before (Unmatched) and after weighting (Matched) across waves for patients without a ceiling of care (A) and patients with ceiling of care (B). The standardized mean difference compares the difference in means between all pairs of waves in standard deviation units.

FIGURE 3: OR for raw, adjusted and IPTW models for in-hospital mortality in patients without a ceiling of care (A) and with ceiling of care (B).

DECLARATIONS

Ethics approval and consent to participate

The study was approved by the Bellvitge Hospital Research Ethics Committee with medicines (CREm) in accordance with Spanish legislation and was performed in accordance with the Helsinki Declaration of 1964. The need for patient informed consent was waived by the ethics committee. Bellvitge's CREm decision was the basis for the approval of the remaining hospital centres.

Competing interests

Cristian Tebé has received fees for speaker lectures and talks from Gedeon Richter, outside the submitted work. The rest of authors declare that they have no competing interests.

Data availability

The datasets generated during and/or analyzed during the current study are available from the corresponding author on reasonable request.

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3 447 **Author’s contributions**

4 448 Conceptual design was performed by SV, JCa, CT, and NP.

5 449 MetroSud cohort data was provided by SV and JCa.

6 450 Statistical analysis was performed by CT and NP.

7 451 The first draft of the manuscript was written by NP and revised by JCa and CT.

8 452 All authors commented on previous versions of the manuscript.

9 453 All authors read and approved the final version of the manuscript.

10 454 CT is the guarantor.

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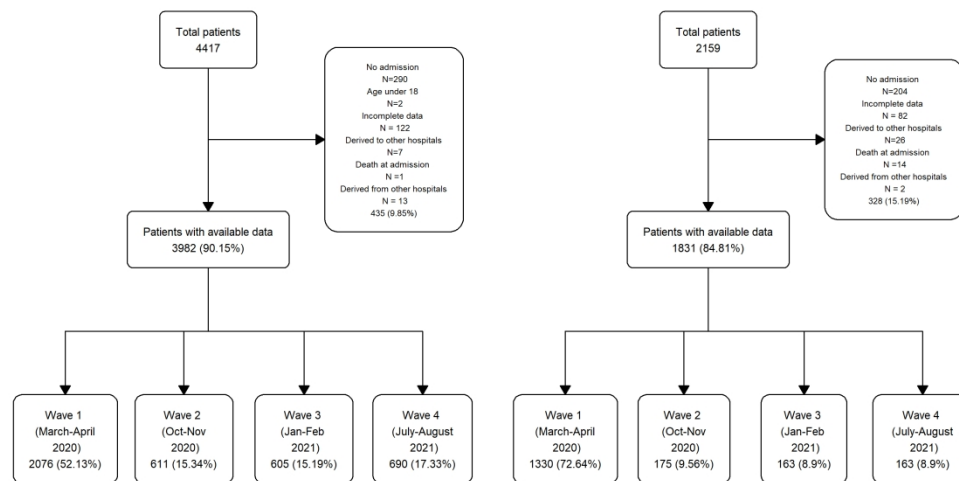


FIGURE 1: Flow-chart of the included patients without ceiling of care (left) and with ceiling of care (right).

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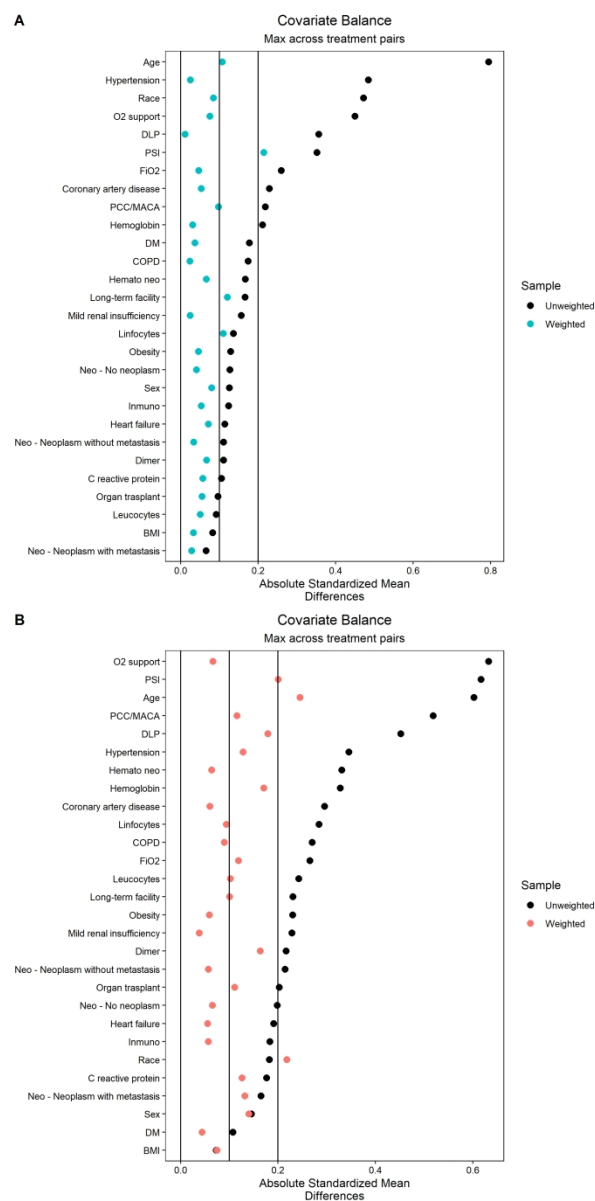


FIGURE 2: Maximum standardized mean differences (SMD) before (Unmatched) and after weighting (Matched) across waves for patients without a ceiling of care (A) and patients with ceiling of care (B). The standardized mean difference compares the difference in means between all pairs of waves in standard deviation units.

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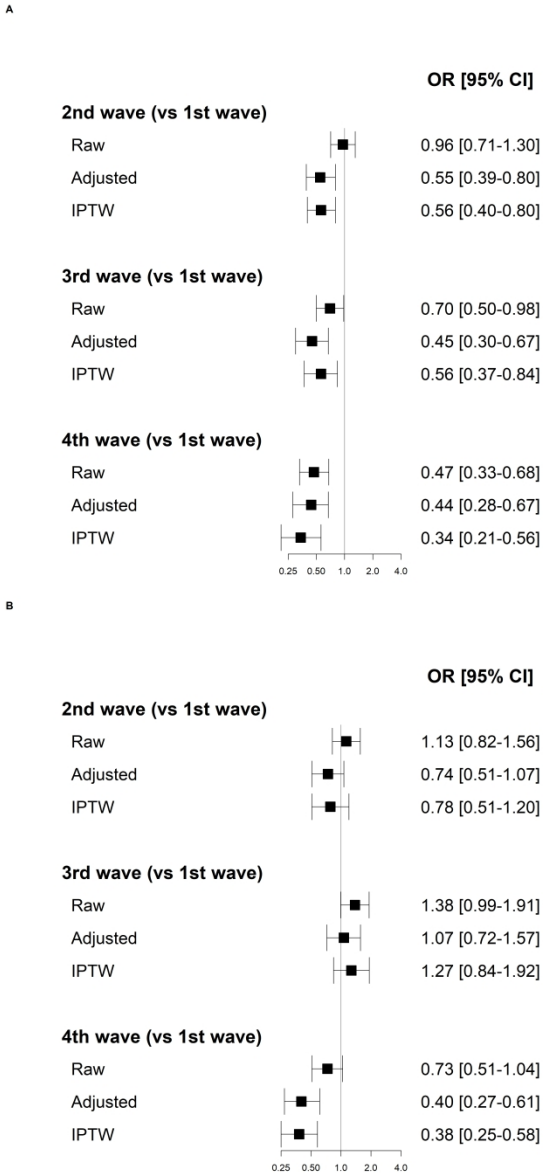


FIGURE 3: OR for raw, adjusted and IPTW models for in-hospital mortality in patients without a ceiling of care (A) and with ceiling of care (B).

299x599mm (300 x 300 DPI)

SUPPLEMENTAL MATERIAL

Table S1: Variables used in the matching procedure according to wave and ceiling of care

	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N =				N =			
	2076	N = 611	N = 605	N = 690	1330	N = 175	N = 163	N = 163
Long-term facility								
Yes	64 (3.1%)	19 (3.1%)	17 (2.8%)	4 (0.6%)	223 (17%)	20 (11%)	20 (12%)	15 (9.2%)
BMI	28.9 (25.9, 32.2)	29.2 (26.4, 32.6)	29.4 (26.6, 32.9)	29.7 (26.4, 33.6)	28.1 (25.4, 31.3)	28.7 (25.8, 31.5)	29.3 (26.0, 31.5)	27.6 (24.8, 31.8)
Unknown	636	103	72	177	456	22	19	17
PCC/MACA								
PCC/MACA	93 (4.5%)	32 (5.2%)	15 (2.5%)	9 (1.3%)	290 (22%)	80 (46%)	47 (29%)	57 (35%)
Diabetes mellitus								
Yes	418 (20%)	124 (20%)	126 (21%)	96 (14%)	414 (31%)	61 (35%)	54 (33%)	59 (36%)
COPD								
Yes	274 (13%)	108 (18%)	119 (20%)	110 (16%)	325 (24%)	54 (31%)	49 (30%)	60 (37%)
Heart failure								

	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar- Apr 2020)	Wave 2 (Oct- Nov 2020)	Wave 3 (Jan- Feb 2021)	Wave 4 (July- Aug 2021)	(Mar- Apr 2020)	Wave 2 (Oct- Nov 2020)	Wave 3 (Jan- Feb 2021)	Wave 4 (July- Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Yes	50 (2.4%)	18 (2.9%)	27 (4.5%)	25 (3.6%)	194 (15%)	37 (21%)	31 (19%)	36 (22%)
Hypertension								
Yes	792 (38%)	266 (44%)	295 (49%)	176 (26%)	881 (66%)	137 (78%)	124 (76%)	132 (81%)
Obesity								
Yes	579 (35%)	221 (36%)	248 (41%)	248 (36%)	285 (29%)	51 (29%)	65 (40%)	50 (31%)
Unknown	404	0	0	0	361	0	0	0
Dyslipidemia								
Yes	698 (34%)	223 (36%)	239 (40%)	158 (23%)	502 (38%)	105 (60%)	92 (56%)	83 (51%)
Mild renal insufficiency								
Yes	83 (4.0%)	27 (4.4%)	42 (6.9%)	25 (3.6%)	234 (18%)	42 (24%)	26 (16%)	41 (25%)
Coronary artery disease								
Yes	91 (4.4%)	33 (5.4%)	28 (4.6%)	7 (1.0%)	112 (8.4%)	26 (15%)	30 (18%)	19 (12%)

	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Haematological neoplasm								
Yes	12 (0.6%)	16 (2.6%)	7 (1.2%)	12 (1.7%)	27 (2.0%)	8 (4.6%)	7 (4.3%)	15 (9.2%)
Organ transplant								
Yes	20 (1.0%)	13 (2.1%)	6 (1.0%)	13 (1.9%)	12 (0.9%)	0 (0%)	3 (1.8%)	1 (0.6%)
Immunology								
Yes	72 (3.5%)	32 (5.2%)	17 (2.8%)	32 (4.6%)	50 (3.8%)	6 (3.4%)	11 (6.7%)	5 (3.1%)
Neoplasm								
No neoplasm	1991 (96%)	578 (95%)	563 (93%)	656 (95%)	1160 (87%)	141 (81%)	136 (83%)	130 (80%)
Neoplasm without metastasis	78 (3.8%)	30 (4.9%)	37 (6.1%)	30 (4.3%)	145 (11%)	26 (15%)	22 (13%)	30 (18%)
Neoplasm with metastasis	7 (0.3%)	3 (0.5%)	5 (0.8%)	4 (0.6%)	25 (1.9%)	8 (4.6%)	5 (3.1%)	3 (1.8%)
D-dimer								
	570 (316, 1050)	530 (284, 970)	500 (266, 895)	365 (250, 690)	722 (378, 1608)	689 (356, 1438)	471 (280, 969)	451 (276, 895)
Unknown	488	55	62	59	384	19	29	19

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	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar- Apr 2020)	Wave 2 (Oct- Nov 2020)	Wave 3 (Jan- Feb 2021)	Wave 4 (July- Aug 2021)	(Mar- Apr 2020)	Wave 2 (Oct- Nov 2020)	Wave 3 (Jan- Feb 2021)	Wave 4 (July- Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
C reactive protein	80 (34, 149)	84 (39, 143)	76 (39, 128)	85 (41, 144)	92 (47, 160)	86 (41, 144)	96 (44, 148)	69 (30, 155)
Unknown	161	30	54	36	102	6	24	8
Haemoglobin	13.90 (12.90, 14.90)	13.60 (12.50, 14.80)	13.90 (12.80, 15.00)	14.05 (13.10, 15.10)	13.30 (12.00, 14.43)	12.50 (11.00, 14.40)	12.80 (11.60, 13.60)	12.50 (11.20, 13.80)
Unknown	150	21	40	24	82	6	15	5
Lymphocytes	0.98 (0.72, 1.33)	0.91 (0.66, 1.26)	0.91 (0.64, 1.23)	0.93 (0.66, 1.24)	0.90 (0.63, 1.24)	0.85 (0.57, 1.18)	0.80 (0.54, 1.13)	0.90 (0.59, 1.36)
Unknown	137	21	40	28	105	6	15	5
Leucocytes	6.5 (5.0, 8.7)	6.5 (5.0, 9.0)	6.4 (5.0, 8.6)	6.2 (4.6, 8.6)	6.9 (5.2, 9.2)	7.2 (5.4, 9.4)	6.5 (5.0, 8.7)	6.9 (5.4, 9.6)
Unknown	109	23	40	26	71	6	15	6
FiO2	21 (21, 21)	21 (21, 28)	21 (21, 28)	21 (21, 31)	21 (21, 24)	21 (21, 28)	21 (21, 35)	24 (21, 31)
Unknown	2	0	0	0	3	0	0	0
Need for oxygen support	511 (25%)	193 (32%)	234 (39%)	316 (46%)	345 (26%)	69 (39%)	77 (47%)	92 (56%)
Unknown	2	0	0	0	3	0	0	0

Median (IQR) for continuous variables; n (%) for categorical variables

No ceiling of care					Ceiling of care			
Wave 1					Wave 1			
(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)		(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
N = 2076	N = 611	N = 605	N = 690		N = 1330	N = 175	N = 163	N = 163

BMI: Body mass index

COPD: Chronic obstructive pulmonary disease

PCC: chronic complex patient

MACA: advanced chronic disease patient

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Impact of ceiling of care on mortality across four COVID-19 epidemic waves in Catalonia: a multicentre prospective cohort study

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Keywords:	COVID-19, INFECTIOUS DISEASES, PALLIATIVE CARE, Epidemiology < INFECTIOUS DISEASES

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Impact of ceiling of care on mortality across four COVID-19 epidemic waves in Catalonia: a multicentre prospective cohort study

Pallarès N^{1,2} MSc, Videla S^{3,4} MD PhD, Carratalà J^{5,6,7,8} MD PhD, Tebé C¹ MPH PhD, on behalf of the MetroSud study group and the Divine study group

1) Biostatistics Support and Research Unit, Germans Trias i Pujol Research Institute and Hospital (IGTP), Badalona, Barcelona, Spain

2) Department of Basic Clinical Practice, School of Medicine and Health Sciences, University of Barcelona, Spain

3) Clinical Research Support Area, Department of Clinical Pharmacology, Germans Trias i Pujol University Hospital, Badalona, Spain

4) Department of Pathology and Experimental Therapeutics, School of Medicine and Health Sciences, University of Barcelona, Spain

5) Department of Infectious Diseases, Bellvitge University Hospital, Barcelona, Spain

6) Bellvitge Biomedical Research Institute (IDIBELL), Barcelona, Spain

7) Centro de Investigación en Red de Enfermedades Infecciosas (CIBERINFEC), Instituto de Salud Carlos III, Madrid, Spain.

8) Department of Clinical Sciences, School of Medicine and Health Sciences, University of Barcelona, Spain

MetroSud Study group: Gabriela Abelenda-Alonso, Alexander Rombauts, Isabel Oriol, Antonella F. Simonetti, Alejandro Rodríguez-Molinero, Elisenda Izquierdo, Vicens Díaz-Brito, Carlota Gudíol, Judit Aranda-Lobo, Marta Arroyo, Carlos Pérez-López, Montserrat Sanmartí, Encarna Moreno, Maria C Alvarez, Ana Faura, Martha González, Paula Cruz, Mireia Colom, Andrea Perez, Laura Serrano.

DIVINE Study group: Mireia Besalú, Erik Cobo, Jordi Cortés, Daniel Fernández, Leire Garmendia, Guadalupe Gómez, Pilar Hereu, Klaus Langohr, Gemma Molist, Núria Pérez-Álvarez, Xavier Piulachs.

Corresponding author

Natàlia Pallarès, MSc
Biostatistics Support and Research Unit
Germans Trias i Pujol Research Institute and Hospital (IGTP)
Campus Can Ruti. Carretera de Can Ruti, Camí de les Escoles s/n. 08916 Badalona, Barcelona, Spain

ABSTRACT

Objective: The aim of this study was to compare in-hospital mortality across waves in patients without and with a ceiling of care at hospital admission.

Design: A multicentre prospective cohort study

Setting: Five tertiary hospitals in Catalonia, Spain, during four waves of the COVID-19 pandemic. Data from the first wave embraced from March to April 2020, second wave from October to November 2020, third wave from January to February 2021 and fourth wave from July to August 2021.

Participants: All consecutive adult subjects (older than 18 years old) admitted to any of the five aforementioned centers. All subjects had a confirmed SARS-CoV-2 infection (with a positive PCR test or antigen test) and an overnight hospital stay. Ceiling of care defined as the highest level of care that a patient will receive during medical treatment was assessed at hospital admission for all patients.

Primary measure: In-hospital mortality

Results: A total of 3982 hospitalized patients without ceiling of care and 1831 hospitalized patients with ceiling of care were included in the analysis. The adjusted odds ratio (OR) of in-hospital mortality in the second wave were 0.57 (95%CI 0.40 to 0.80), in the third 0.56 (95%CI 0.37 to 0.84) and in the fourth 0.34 (95%CI 0.21 to 0.56) compared with the first wave in subjects without ceiling of care. The adjusted odds ratio were significantly lower in the fourth (0.38 95%CI 0.25 to 0.58) wave compared to the first wave in subjects with ceiling of care.

Conclusions: In patients without ceiling of care, mortality decreased over time suggesting better disease knowledge and management. In ceiling of care, only fourth-wave patients were less likely to die than first-wave patients. In a future infectious disease pandemic, it will be a challenge to improve the management of patients with ceiling of care.

Keywords

COVID-19, Infectious diseases, Palliative care, Epidemiology

Strengths and limitations of this study

- This is multicentric study with a large number of subjects included from four different waves of the COVID-19 pandemic.
- Several methods were used to compare in-hospital mortality between waves to increase the robustness of the estimated effects.
- Despite the inverse probability weighting analysis, there may be unobserved characteristics that lead to residual confounding.

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3 81 • The national vaccination campaign started for the elderly subjects before the
4 82 fourth wave so it could not be used in the adjustment analysis.
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INTRODUCTION

Despite the lack of definition in epidemiology, the term epidemic wave implies a natural pattern of peaks and troughs in the incidence of cases or hospitalizations due to an outbreak [1]. Epidemics often occur in local or global waves, each one with variations in severity or in transmission dynamics [2–4].

Following a similar pattern, the COVID-19 pandemic began in Wuhan, China, in December 2019, and spread rapidly across Europe, with the first outbreak in Italy in February 2020. During the course of the pandemic, countries and regions experienced several waves with distinct peaks in cases. In Spain, 7 waves of the pandemic have been recorded between March 2020 and September 2023, with almost 14 million confirmed cases and more than 120.000 deaths [5]. Throughout this period, knowledge of the disease has progressively increased with the sequencing of the virus [6], clinical trials to assess treatments efficacy [7,8], the identification of different strains of the virus [9] and the development of vaccines [10]. All these factors, together with the natural immunity protection against COVID-19 [11], lead to a reduction in the need for hospitalization, in-hospital mortality and complications.

The therapeutic ceiling of care refers to the highest level of care that a patient will receive during medical treatment. In general, in a non-pandemic setting, decisions about the ceiling of care are common practice when dealing with patients with a critical prognosis and have implications for the use of life-sustaining measures such as intubation, mechanical ventilation, and cardiopulmonary resuscitation. However, in the peaks of the COVID-19 pandemic, decisions about the maximum level of care that each patient should receive, besides of the critical prognosis of the patient, were made in a scenery of emergency with excess demand for critical care and limited availability of clinical resources. Previously published data [12,13] suggest that COVID-19 hospitalized patients who had a ceiling of care were mainly older, had more comorbidities and higher incidence of in-hospital death. In-hospital mortality has been shown to decrease over time [14,15]. However, little is known about the impact of ceiling of care on mortality in hospitalized patients with COVID-19 across pandemic waves. Stratifying by care limitations helps to distinguish whether the reduction in mortality was due to advances in intensive care unit management, improved general hospital care, or shifts in decision making. This approach addresses a gap in previous research, which has often overlooked how changes in patient selection for intensive care can bias mortality trends. Understanding these dynamics can inform clinical decision-making and ensure optimal management for all patients, regardless of their care limitations.

Our hypothesis is that the decrease in in-hospital mortality over time is different in patients with and without ceiling of care. The aim of this study was to compare in-hospital mortality across four COVID-19 waves between patients with and without a ceiling of care at hospital admission.

METHODS

Study design and setting

The MetroSud study is an observational multicenter study conducted in five centers located in the southern metropolitan area of Barcelona (Catalonia, Spain), to characterise all patients with COVID-19 admitted to these hospitals during four waves of the pandemic from March 2020 to August 2021. COVID-19 epidemic waves followed a pattern of peaks and troughs in the incidence of cases or hospitalizations. There is no official date for the start or end of a wave in Catalonia, but recruitment in the MetroSud cohort occurred during peaks in the incidence of cases or hospitalizations in our hospitals. The Infectious Diseases Unit of Bellvitge's Hospital developed the protocol for this study as soon as the first cases appeared. After the first wave, and with the experience gained, the protocol and data collection were reactivated when early epidemiological indicators signaled the arrival of the second wave. This approach was continued in subsequent waves. Analyzed data of the first wave of COVID-19 pandemic embraced from March to April 2020, second wave from October to November 2020, third wave from January to February 2021 and fourth wave from July to August 2021 [16]. MetroSud cohort has been previously described [12].

Eligibility criteria

The MetroSud cohort included all consecutive adult subjects (older than 18 years old) admitted to any of the five aforementioned centers. All subjects had a proven SARS-CoV-2 infection (with a positive PCR test or antigen test).

Data sources and study variables

An electronic case report form (eCRF) in REDCap [17] was designed in March 2020 to collect study data: in-hospital mortality as main outcome, ceiling of care and epidemic wave as main independent variables, and subjects clinical profile to adjust for potential confounding.

Demographic data (age, sex, race), comorbidities and other relevant findings on medical history, previous medications, clinical symptoms, vital signs (body temperature, FiO₂, O₂ saturation, blood pressure, pulse, and respiratory rate), laboratory results (D dimer, C-reactive protein, lactat dehydrogenase, leukocytes, and others) and respiratory exploration (wheezing, rhoncus), Pneumonia severity index (PSI) and ceiling of care were collected at baseline by the attending physicians. Patient status at hospital discharge was also recorded in the eCRF. No variables were transformed, and ranges of plausible values for continuous variables were indicated in the eCRF to ensure data quality.

The presence or absence of ceiling of care was decided at the emergency room by the attending physicians according to their criteria, taking into account the patient's potential benefit of intensive treatments. In the beginning of the first wave, due to the ICU demand and capacity, the availability of resources at each participating hospital was also taken into account. Patients without a ceiling of care would have access to an

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3 171 ICU or could receive invasive mechanical ventilation (IMV). Otherwise, patients
4 172 assigned to ceiling of care would have limited access to the ICU and, if they required
5 173 any respiratory support, it would be non-rebreather mask, high-flow nasal cannula or
6 174 NIMV. Information about ceiling of care was registered in the eCRF at hospital
7 175 admission. Patients without information on ceiling of care assigned were excluded
8 176 from the analysis.
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12 178 **Outcome variable**
13 179 The outcome variable was in-hospital mortality defined as death by any cause during
14 180 hospitalization and was registered in the eCRF. Patients without information on in-
15 181 hospital mortality were excluded from the analysis.
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18 183 The study was approved by the Bellvitge Hospital Research Ethics Committee with
19 184 medicines (CREm), with reference PR140/20 and code HUB-INF-COHORT-HUB-COVID,
20 185 in accordance with Spanish legislation and was performed in accordance with the
21 186 Helsinki Declaration of 1964. The need for patient informed consent was waived by the
22 187 ethics committee. Bellvitge's CREm decision was the basis for the approval of the
23 188 remaining hospital centers.
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27 190 **Statistical methods**
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29 191 To describe cohort characteristics, categorical variables were presented as the number
30 192 of cases and percentage, while continuous variables were expressed as the mean and
31 193 standard deviation (SD) or median and interquartile range (IQR). All analyses were
32 194 presented by wave and stratified by ceiling of care.
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35 195 A pool of essential variables to describe the baseline profile of patients was defined.
36 196 This pool included age, sex, Charlson score, ceiling of care, and circumstances at
37 197 discharge. Patients who had incomplete data on this pool of variables were discarded
38 198 from the analysis.
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41 199 Once the variables to be used to match patients were identified, multiple imputation
42 200 with chained equations (MICE) [18] was used to create five datasets with complete
43 201 data. Missing data were assumed to be at random. Predictive mean matching was used
44 202 to impute continuous variables and binomial logistic regression was used to impute
45 203 binary variables. Information on age, sex and baseline comorbidities (completed for all
46 204 patients after exclusions) was used to impute missing values for obesity, body mass
47 205 index (BMI), race, pneumonia severity Index (PSI), FiO₂, oxygen support, D-dimer, C-
48 206 reactive protein, leukocytes, haemoglobin and lymphocytes. Final estimates were
49 207 adjusted for variability between the five imputed datasets according to the Rubin rules
50 208 [19] to obtain the final model.
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55 209 With the database with all the missing data imputed, three models were constructed
56 210 to study the association between in-hospital mortality and wave: 1) a crude logistic
57 211 regression model using wave as a covariate, 2) a fully adjusted logistic regression
58 212 model and 3) an inverse probability weighting (IPW) logistic regression model.
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After discussion with clinicians, the variables included in the fully adjusted logistic regression model to minimize confounding and make patients comparable between waves were baseline variables that define the patient's status at hospital admission: age, sex, race, BMI, obesity, long-term facility, comorbidities (diabetes mellitus, COPD, heart failure, hypertension, renal insufficiency, dyslipidemia, coronary heart disease, haematological neoplasm, solid neoplasm, organ transplantation, immunosuppressive treatment, chronic complex patient (PCC) and patients with advanced chronic disease (MACA), baseline laboratory values (dimer, C-reactive protein, leukocytes, haemoglobin, lymphocytes), pneumonia severity index (PSI), FiO2 and oxygen support.

IPW [20] was used to adjust for differences in the patient baseline profile between waves. Bayesian additive regression trees, entropy balancing, generalised boosted models and generalised linear models were tested as methods for weighting individuals. In the end, we chose the method with better covariate balance between waves after weighting, which was the bayesian additive regression trees method [21]. In each imputed dataset, weights were calculated with the wave as the outcome and the variables used for the full adjusted logistic model as covariates.

To identify imbalances between waves after weighting, we estimated and described the standardised mean differences in baseline variables before and after weighting. We then fitted a logistic regression model for each imputation with in-hospital death as the outcome, using the stabilised weights and model-robust standard errors and adjusting for the variables that remained imbalanced between groups after weighting.

To overcome the limitation of assuming missing at random, a sensitivity analysis was performed by repeating the analyses using only those patients who had complete information on all variables.

We used the STROBE cohort checklist [22] when writing our report. All analyses were performed with a two-sided significance level of 0.05 using R software version 4.3.0 [23]. The main R packages used for data management and analysis were flowchart [24], REDCapDM [25], mice [18], WeightIt [26], cobalt [27] and survey [28].

Patient and public involvement

There was no patient or public involvement in the development of the research design or in conducting the study.

RESULTS

Flow chart

A total of 4417 patients without ceiling of care and 2159 patients with ceiling of care were included in the MetroSud. Patients who were discharged or died within 24 hours of admission (N=494 and N=15, respectively) were not considered hospitalized for the purposes of this study, in accordance with the study protocol. For those discharged within 24 hours, it was assumed that their clinical condition may have been more appropriately managed in an outpatient setting. In both cases, however, key study variables were often unavailable or incomplete, limiting the ability to include them in the analysis. Patients with incomplete data on a pool of essential variables (age, sex, Charlson score, ceiling of care, and circumstances at discharge) (N=204) or patients who were initially admitted to one hospital but transferred to another and treated in the latter (N=48) were also excluded. After exclusions, a total of 3982 patients without ceiling of care and a total of 1831 patients with ceiling of care were included in the analysis. All patients were followed up until in-hospital death or hospital discharge. (Figure 1, Flow Chart)

Baseline characteristics by wave

Table 1 describes the baseline characteristics of the included patients by wave and stratified by ceiling of care. Other variables included in the matching process are described in Supplementary Table 1.

TABLE 1: Patient’s most relevant characteristics according to wave and ceiling of care.

	No ceiling of care				Ceiling of care			
	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Age	59 (49, 69)	62 (53, 71)	63 (53, 72)	49 (37, 63)	79 (72, 85)	83 (78, 88)	83 (78, 87)	85 (80, 89)
Sex								

	No ceiling of care				Ceiling of care			
	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Women	855 (41%)	222 (36%)	248 (41%)	242 (35%)	565 (42%)	75 (43%)	81 (50%)	75 (46%)
Race								
Caucasian	1206 (78%)	394 (69%)	464 (86%)	406 (65%)	795 (96%)	164 (95%)	154 (97%)	154 (98%)
Other	341 (22%)	179 (31%)	73 (14%)	218 (35%)	32 (3.9%)	9 (5.2%)	4 (2.5%)	3 (1.9%)
Unknown	529	38	68	66	503	2	5	6
Charlson Index								
	2.00 (1.00, 3.00)	2.00 (1.00, 4.00)	3.00 (1.00, 4.00)	1.00 (0.00, 3.00)	5.00 (4.00, 7.00)	6.00 (5.00, 8.00)	6.00 (5.00, 7.00)	6.00 (5.00, 7.00)
PSI								
	62 (50, 79)	69 (56, 87)	66 (53, 84)	59 (46, 78)	97 (79, 123)	115 (95, 134)	103 (84, 124)	114 (96, 135)
Unknown	374	6	3	3	239	1	4	2

Median (IQR) for continuous variables; n (%) for categorical variables

PSI: Pneumonia severity index

Regarding age, patients with a ceiling of care were, in median, 20 years older than patients without a ceiling of care in all waves. There were no differences in the proportion of women. The most common race was Caucasian (in all waves, almost 90% of patients without ceiling of care and over 70% of patients with ceiling of care were Caucasian). Patients with a ceiling of care had a median Charlson Index more than 3 points higher than patients without a ceiling of care in all waves. PSI scores for patients

with ceiling of care were more than 35 points higher in all waves (greater differences in wave 4) than PSI scores for patients without ceiling of care.

In-hospital mortality

The overall cumulative incidence of in-hospital mortality for patients with and without ceiling of care in all waves is shown in Table 2.

TABLE 2: Cumulative incidence and 95% confidence interval for in-hospital mortality according to wave and ceiling of care.

	Wave 1 (Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
No ceiling of care	10.50% [9.23 to 11.92]	10.15% [7.92 to 12.89]	7.60% [5.68 to 10.09]	5.22% [3.73 to 7.22]
Ceiling of care	37.07% [34.48 to 39.74]	40.00% [32.76 to 47.69]	44.79% [37.06 to 52.76]	30.06% [23.27 to 37.81]

About 1 in 10 patients without ceiling of care died in hospital in the first and second waves. In patients with a ceiling of care, about 4 in 10 patients die in hospital in the first three waves. The percentages are lower in the fourth wave (5% and 30% respectively for patients without and with a ceiling of care).

Mortality in patients without ceiling of care

Figure 2A shows the balance of covariates before and after IPW by means of the standardised mean differences (SMD) in patients without a ceiling of care. The SMD for PSI remains above 0.2. To correct for this imbalance, PSI was included in the weighted mortality models.

The odds ratios of the three models for mortality are shown in Figure 3A. The results with the three methods are consistent and show the same trend for all waves. Patients from waves 2, 3 and 4 were less likely to die in hospital than patients from wave 1 both in the raw models and in the models adjusted for covariates or adjusted with weights (OR for all models and all waves lower than 1). In addition, the value of the OR decreases across waves.

Mortality in patients with ceiling of care

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Figure 2B shows the balance of covariates before and after IPW by means of the SMD in patients with a ceiling of care assigned at admission. Age, PSI and race showed a difference between waves greater than 0.2. These variables were included as adjustments in the weighted mortality model to account for these differences.

The odds ratios of the three models for mortality are shown in Figure 3B. No differences were found between 1st and 2nd wave patients or between 1st and 3rd wave patient (neither in the crude nor in the adjusted models). For wave 4th, both adjusted and IPW models showed that, given two patients with the same baseline profile, a wave 4 patient was less likely to die in hospital than a wave 1 patient.

Sensitivity analysis

To account for the assumption of missing at random, we perform a sensitivity analysis using only those patients who had all the variables completed. The results were in the same direction as for the imputed database. In patients with ceiling of care, the effect of the wave in mortality was the same in patients with and without complete data. In patients without ceiling of care the effect was also similar, but as the sample size of the cohort with complete data was smaller, the odds ratio for the 2nd and 3rd waves did not reach statistical significance.

DISCUSSION

Our multicentre cohort study compared in-hospital mortality across COVID-19 waves between patients with and without a ceiling of care at hospital admission. We found that among patients without ceiling of care, those admitted in the first wave had worse in-hospital mortality than patients hospitalized during the other waves. Moreover, the adjusted models showed a significant decrease in mortality as the waves progressed. Among patients with a ceiling of care, no differences in in-hospital mortality were found between second and first wave patients, or between third and first wave patients. Only in the fourth wave, patients were less likely to die than first wave patients after adjustment for baseline risk. The magnitude of this effect on mortality reduction observed in patients with ceiling of care in the 4th wave was similar to the effect observed among patients without ceiling of care in the same 4th wave.

It is worth noting that if the differences in mortality between waves were only due to patient's risk profile, the mortality rates would be similar after adjustment for baseline profile. However, this is not the case, as Figure 3A shows that in the adjusted and weighted models, mortality among patients without ceiling of care decreases as waves progress in time (OR decreasing from 0.56 (second and third wave) to 0.34 (fourth

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wave) when comparing with first wave)). The emergency situation experienced by the hospitals in the first months of the pandemic, with a lack of organization prepared to face an emergency such as COVID-19, partly explains the differences observed [29]. Besides, in the first wave, hospital resources (such as ICU beds, number of non-invasive ventilators or high-flow nasal oxygen therapy devices) and human resources were not sufficient to cope with the high demand for medical care [30]. ICU capacity is known to be an important indicator of hospital stress (health system resilience) which is associated with a reduction in quality of care and poorer patient outcomes [31]. In addition, other factors such as the increasing knowledge about the disease, facilitated by the rapid publication of clinical trials analysing new treatments [8], or the impact of public health surveillance measures, such as lockdowns [32] could explain this reduction in mortality. The harvest effect could also explain this decrease in mortality, as deaths that would have occurred anyway in subsequent waves may have been precipitated by the high mortality in the first wave of COVID-19 [33]. Similarly, the aggressiveness of SARS-CoV-2 varied between strains, and may also have played a role in the reduction in mortality [34].

As expected, mortality was higher among patients with ceiling care. In this group of patients, there are no differences in mortality in the first three waves, but there is a decrease in mortality in wave 4 (OR 0.38 95%CI 0.25 to 0.58) (Figure 3B). In Spain, this fourth wave mainly affected young patients. Older patients, who were more likely to be assigned a ceiling of care, were already vaccinated at that time [35]. A study in nursing homes in our geographical area (Catalonia) [36] showed that vaccination was associated with a 95% reduction in mortality among nursing home residents. Studies in Italy and Switzerland also showed that the vaccine was about 95% effective against death in the general population [33,37]. These results therefore suggest that there is no improvement in medical management that affects in-hospital mortality until wave 4, which coincides with the elderly vaccination campaign. The lack of a contrafactual scenario in which people received intensive care makes it difficult to assess any potential benefit. Further research on this topic and replication of these results in other cohorts would be needed. Moreover, it will be of interest to study the management of ceiling of care in other cultural settings. It would also be interesting to investigate whether the impact of ceiling of care is the same on other outcomes, such as complications or length of hospital stay.

The high probability of a new epidemic caused by an infectious organism merits in-depth reflection by the medical and scientific community, in particular to reach a consensus on the definition of ceiling of care and to define a guideline for the management of patients who are candidates for a ceiling of care [38]. In the event of a future pandemic caused by an infectious organism, the challenge will be to improve mortality in patients with ceiling of care. To this end, the scientific community needs to develop an action plan that will enable a rapid response in terms of both human resources (by increasing the number of trained health workers), and facilities (for example, so that the ICUs can quickly increase the number of beds) [39].

Our study has some limitations that should be acknowledged. Excluding subjects who were discharged or died within 24 hours of admission may introduce selection bias by systematically omitting individuals with atypically short hospital stays. This limitation

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should be considered when interpreting the generalizability of the study findings. Moreover, we could have residual confounding because even after using all the characteristics available at admission to make the baseline status of patients comparable, there may be unobserved characteristics that make patients different between waves. For example, we knew whether a patient had pathology or not, but we could not take into account how advanced it was. A variable that collects information on patients' frailty at baseline might also be of interest for a better risk assessment. In addition, vaccines and treatments could not be used in the matching: vaccines because they did not exist in the first wave [10] and treatments because they changed drastically between waves due to increasing knowledge about the disease [7,8]. Moreover, we do not have data on the follow-up of patients with regard to treatments received during hospitalization, which could help to understand some of the differences in mortality. Another limitation of the study is that we assumed that the missing values in our data were at random and imputed them using standard techniques. To account for this, a sensitivity analysis was performed repeating the analysis only with patients who had complete information on all variables, and the results were in the same line, confirming the robustness of the analysis. Moreover, we cannot guarantee that the same criteria were used to define the therapeutic ceiling of care in all hospitals. In fact, one of the challenges in clinical practice during the COVID-19 pandemic was to define the ceiling of care for infected patients. Even though the definition of ceiling of care is not a standardized one, the definition in the MetroSud cohort was a pragmatic one which would be readable and understood by clinician teams involved in reaching these decisions. Our definition is consistent with that used in the Leeds cohort [13] and with the one used in a multicentre study to identify factors influencing ceiling of treatment in an Emergency Department [40]. In addition, when the study protocol was written, little was known about COVID-19, including the lack of immunity and the possibility of reinfection. Before the emergence of the Omicron strain, the incidence of COVID-19 reinfection leading to hospitalization was very low (<1%) [41]. Our last wave included patients from July to August 2021, when the Omicron strain had not yet reach Spain and the incidence of reinfection was still very low. However, we could not rule out the possibility that some subjects from the 3rd or 4th wave had previously been included in the study. Due to data protection laws, we did not have access to patients' clinical records, which prevented verification. Nevertheless, participating physicians were aware of this possibility and took measures to avoid case duplication.

The strengths of our study are the large number of subjects included from different hospitals and from four different waves of the pandemic, and the availability of information on ceiling of care. In addition, the different methods used to compare in-hospital mortality by waves led to the same results, demonstrating the robustness of the analysis.

In conclusion, knowing that the evolution of in-hospital mortality through waves is different in patients with and without ceiling of care could help the scientific community to address the management of patients with ceiling of care to improve their outcomes in a new pandemic scenario. The lessons learned from the COVID-19

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434 pandemic could help health-care professional and health policy-makers to face future
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DECLARATIONS

Ethics approval and consent to participate

The study was approved by the Bellvitge Hospital Research Ethics Committee with medicines (CREm), with reference PR140/20 and code HUB-INF-COHORT-HUB-COVID , in accordance with Spanish legislation and was performed in accordance with the Helsinki Declaration of 1964. The need for patient informed consent was waived by the ethics committee. Bellvitge's CREm decision was the basis for the approval of the remaining hospital centres.

Competing interests

Cristian Tebé has received fees for speaker lectures and talks from Gedeon Richter, outside the submitted work. The rest of authors declare that they have no competing interests.

Data availability

The datasets generated during and/or analyzed during the current study are available from the corresponding author on reasonable request.

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Author's contributions

Conceptual design was performed by SV, JCa, CT, and NP.

MetroSud cohort data was provided by SV and JCa.

Statistical analysis was performed by CT and NP.

The first draft of the manuscript was written by NP and revised by JCa and CT.

All authors commented on previous versions of the manuscript.

All authors read and approved the final version of the manuscript.

CT is the guarantor.

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FIGURE LEGENDS

FIGURE 1: Flow-chart of the included patients without ceiling of care (left) and with ceiling of care (right).

FIGURE 2: Maximum standardized mean differences (SMD) before (Unmatched) and after weighting (Matched) across waves for patients without a ceiling of care (A) and patients with ceiling of care (B). The standardized mean difference compares the difference in means between all pairs of waves in standard deviation units.

FIGURE 3: OR for raw, adjusted and IPTW models for in-hospital mortality in patients without a ceiling of care (A) and with ceiling of care (B).

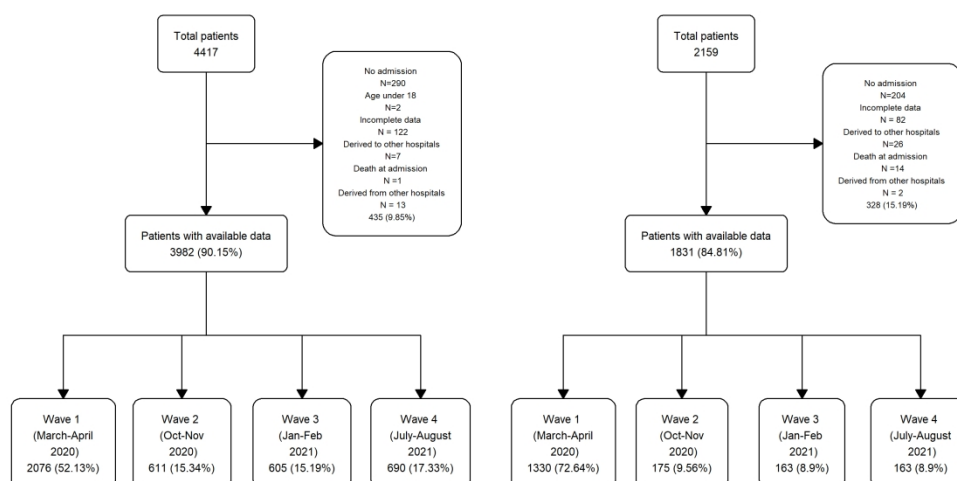


FIGURE 1: Flow-chart of the included patients without ceiling of care (left) and with ceiling of care (right).

249x199mm (300 x 300 DPI)

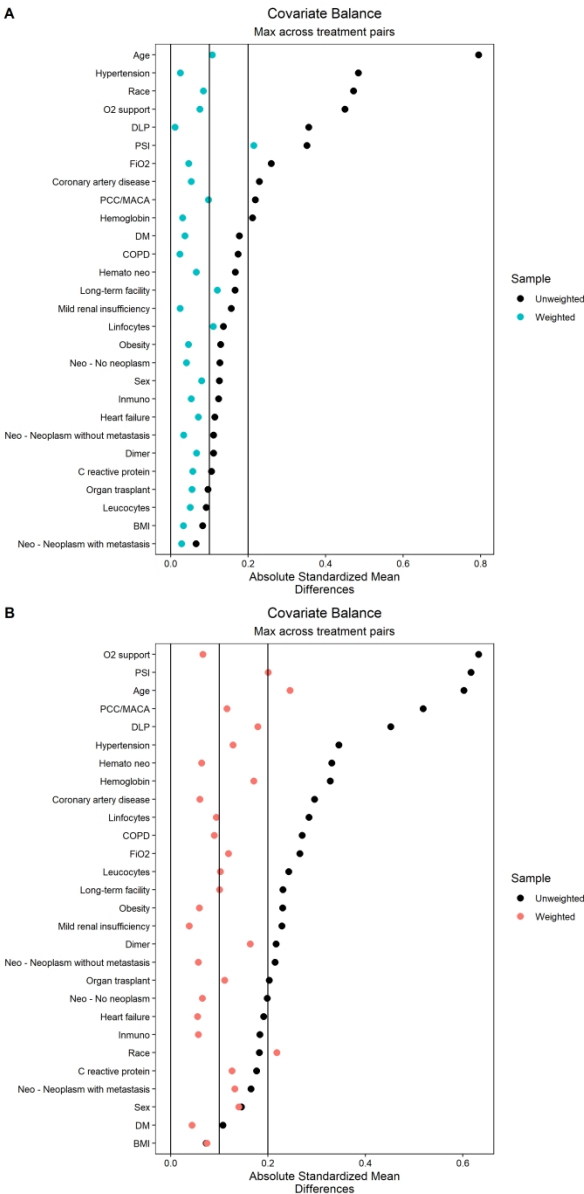


FIGURE 2: Maximum standardized mean differences (SMD) before (Unmatched) and after weighting (Matched) across waves for patients without a ceiling of care (A) and patients with ceiling of care (B). The standardized mean difference compares the difference in means between all pairs of waves in standard deviation units.

199x399mm (300 x 300 DPI)

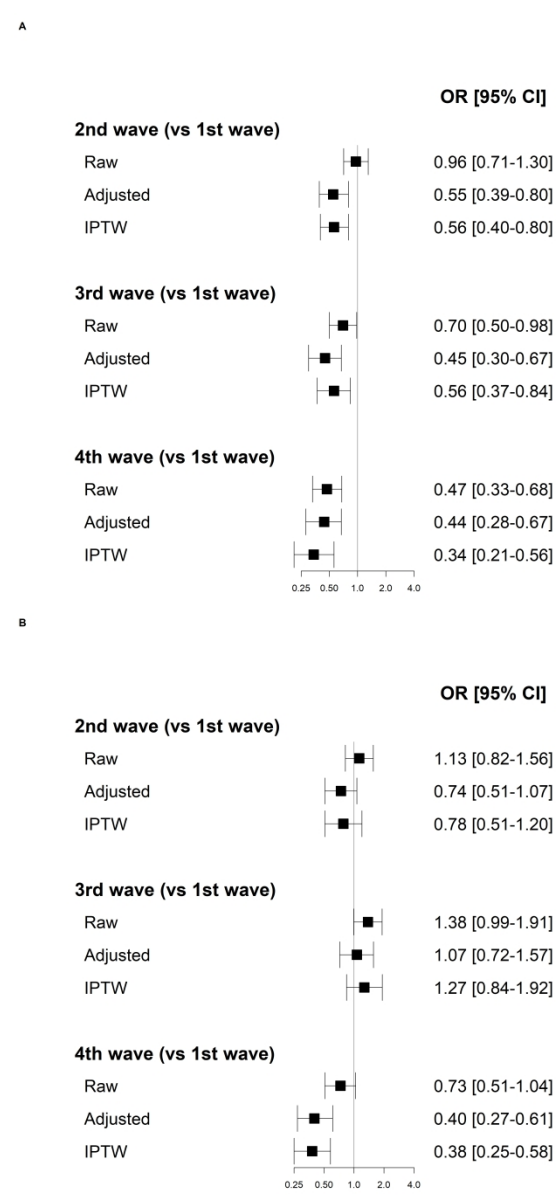


FIGURE 3: OR for raw, adjusted and IPTW models for in-hospital mortality in patients without a ceiling of care (A) and with ceiling of care (B).

299x599mm (300 x 300 DPI)

SUPPLEMENTAL MATERIAL

Table S1: Variables used in the matching procedure according to wave and ceiling of care

	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N =				N =			
	2076	N = 611	N = 605	N = 690	1330	N = 175	N = 163	N = 163
Long-term facility								
Yes	64 (3.1%)	19 (3.1%)	17 (2.8%)	4 (0.6%)	223 (17%)	20 (11%)	20 (12%)	15 (9.2%)
BMI	28.9 (25.9, 32.2)	29.2 (26.4, 32.6)	29.4 (26.6, 32.9)	29.7 (26.4, 33.6)	28.1 (25.4, 31.3)	28.7 (25.8, 31.5)	29.3 (26.0, 31.5)	27.6 (24.8, 31.8)
Unknown	636	103	72	177	456	22	19	17
PCC/MACA								
PCC/MACA	93 (4.5%)	32 (5.2%)	15 (2.5%)	9 (1.3%)	290 (22%)	80 (46%)	47 (29%)	57 (35%)
Diabetes mellitus								
Yes	418 (20%)	124 (20%)	126 (21%)	96 (14%)	414 (31%)	61 (35%)	54 (33%)	59 (36%)
COPD								
Yes	274 (13%)	108 (18%)	119 (20%)	110 (16%)	325 (24%)	54 (31%)	49 (30%)	60 (37%)
Heart failure								

	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Yes	50 (2.4%)	18 (2.9%)	27 (4.5%)	25 (3.6%)	194 (15%)	37 (21%)	31 (19%)	36 (22%)
Hypertension								
Yes	792 (38%)	266 (44%)	295 (49%)	176 (26%)	881 (66%)	137 (78%)	124 (76%)	132 (81%)
Obesity								
Yes	579 (35%)	221 (36%)	248 (41%)	248 (36%)	285 (29%)	51 (29%)	65 (40%)	50 (31%)
Unknown	404	0	0	0	361	0	0	0
Dyslipidemia								
Yes	698 (34%)	223 (36%)	239 (40%)	158 (23%)	502 (38%)	105 (60%)	92 (56%)	83 (51%)
Mild renal insufficiency								
Yes	83 (4.0%)	27 (4.4%)	42 (6.9%)	25 (3.6%)	234 (18%)	42 (24%)	26 (16%)	41 (25%)
Coronary artery disease								
Yes	91 (4.4%)	33 (5.4%)	28 (4.6%)	7 (1.0%)	112 (8.4%)	26 (15%)	30 (18%)	19 (12%)

	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
Haematological neoplasm								
Yes	12 (0.6%)	16 (2.6%)	7 (1.2%)	12 (1.7%)	27 (2.0%)	8 (4.6%)	7 (4.3%)	15 (9.2%)
Organ transplant								
Yes	20 (1.0%)	13 (2.1%)	6 (1.0%)	13 (1.9%)	12 (0.9%)	0 (0%)	3 (1.8%)	1 (0.6%)
Immunology								
Yes	72 (3.5%)	32 (5.2%)	17 (2.8%)	32 (4.6%)	50 (3.8%)	6 (3.4%)	11 (6.7%)	5 (3.1%)
Neoplasm								
No neoplasm	1991 (96%)	578 (95%)	563 (93%)	656 (95%)	1160 (87%)	141 (81%)	136 (83%)	130 (80%)
Neoplasm without metastasis	78 (3.8%)	30 (4.9%)	37 (6.1%)	30 (4.3%)	145 (11%)	26 (15%)	22 (13%)	30 (18%)
Neoplasm with metastasis	7 (0.3%)	3 (0.5%)	5 (0.8%)	4 (0.6%)	25 (1.9%)	8 (4.6%)	5 (3.1%)	3 (1.8%)
D-dimer								
	570 (316, 1050)	530 (284, 970)	500 (266, 895)	365 (250, 690)	722 (378, 1608)	689 (356, 1438)	471 (280, 969)	451 (276, 895)
Unknown	488	55	62	59	384	19	29	19

	No ceiling of care				Ceiling of care			
	Wave 1				Wave 1			
	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)	(Mar-Apr 2020)	Wave 2 (Oct-Nov 2020)	Wave 3 (Jan-Feb 2021)	Wave 4 (July-Aug 2021)
	N = 2076	N = 611	N = 605	N = 690	N = 1330	N = 175	N = 163	N = 163
C reactive protein	80 (34, 149)	84 (39, 143)	76 (39, 128)	85 (41, 144)	92 (47, 160)	86 (41, 144)	96 (44, 148)	69 (30, 155)
Unknown	161	30	54	36	102	6	24	8
Haemoglobin	13.90 (12.90, 14.90)	13.60 (12.50, 14.80)	13.90 (12.80, 15.00)	14.05 (13.10, 15.10)	13.30 (12.00, 14.43)	12.50 (11.00, 14.40)	12.80 (11.60, 13.60)	12.50 (11.20, 13.80)
Unknown	150	21	40	24	82	6	15	5
Lymphocytes	0.98 (0.72, 1.33)	0.91 (0.66, 1.26)	0.91 (0.64, 1.23)	0.93 (0.66, 1.24)	0.90 (0.63, 1.24)	0.85 (0.57, 1.18)	0.80 (0.54, 1.13)	0.90 (0.59, 1.36)
Unknown	137	21	40	28	105	6	15	5
Leucocytes	6.5 (5.0, 8.7)	6.5 (5.0, 9.0)	6.4 (5.0, 8.6)	6.2 (4.6, 8.6)	6.9 (5.2, 9.2)	7.2 (5.4, 9.4)	6.5 (5.0, 8.7)	6.9 (5.4, 9.6)
Unknown	109	23	40	26	71	6	15	6
FiO2	21 (21, 21)	21 (21, 28)	21 (21, 28)	21 (21, 31)	21 (21, 24)	21 (21, 28)	21 (21, 35)	24 (21, 31)
Unknown	2	0	0	0	3	0	0	0
Need for oxygen support	511 (25%)	193 (32%)	234 (39%)	316 (46%)	345 (26%)	69 (39%)	77 (47%)	92 (56%)
Unknown	2	0	0	0	3	0	0	0

Median (IQR) for continuous variables; n (%) for categorical variables

No ceiling of care					Ceiling of care			
Wave 1					Wave 1			
(Mar- Apr 2020)	Wave 2 (Oct- Nov 2020)	Wave 3 (Jan- Feb 2021)	Wave 4 (July- Aug 2021)		(Mar- Apr 2020)	Wave 2 (Oct- Nov 2020)	Wave 3 (Jan- Feb 2021)	Wave 4 (July- Aug 2021)
N =					N =			
2076	N = 611	N = 605	N = 690		1330	N = 175	N = 163	N = 163

BMI: Body mass index

COPD: Chronic obstructive pulmonary disease

PCC: chronic complex patient

MACA: advanced chronic disease patient