

Construction of a Demand and Capacity Model for Intensive Care and Hospital Ward Beds, and Mortality From Covid-19

Christopher Martin (✉ chris.martin@crystallise.com)

Crystallise Ltd

Stuart McDonald

Lloyds Banking Group (United Kingdom)

Steve Bale

Munich Re UK Life Branch

Michiel Luteijn

Hannover Re UK Life Branch

Rahul Sarkar

Medway NHS Foundation Trust

Research Article

Keywords: COVID-19, demand, capacity, model, mortality, public health, infectious diseases

Posted Date: March 6th, 2021

DOI: <https://doi.org/10.21203/rs.3.rs-274009/v1>

License:  This work is licensed under a Creative Commons Attribution 4.0 International License.

[Read Full License](#)

1 Title page

2 **Construction of a demand and capacity**
3 **model for intensive care and hospital ward**
4 **beds, and mortality from COVID-19**

5 **Christopher Martin^{1,2}, Stuart McDonald³, Steve Bale⁴, Michiel Luteijn⁵, Rahul Sarkar⁶,**

6 ***Correspondence:** chris.martin@crystallise.com

7 ¹Director of Modelling at Crystallise, Unit 19, Saffron Court, Southfields Business Park, Basildon, Essex,
8 SS15 6SS.

9 **Abstract**

10 **Background**

11 This paper describes a model for estimating COVID-19 related excess deaths that are a direct
12 consequence of insufficient hospital ward bed and intensive care unit (ICU) capacity.

13 **Methods**

14 Compartmental models were used to estimate deaths under different combinations of ICU and ward
15 care required and received in England up to late April 2021. Model parameters were sourced from
16 publicly available government information and organisations collating COVID-19 data. A sub-model was

17 used to estimate the mortality scalars that represent increased mortality due to insufficient ICU or
18 general ward bed capacity. Three illustrative scenarios for admissions numbers, 'Optimistic', 'Middling'
19 and 'Pessimistic', were modelled and compared with the subsequent observations to the 3rd February.

20 Results

21 The key output was the demand and capacity model described. There were no excess deaths from a lack
22 of capacity in the 'Optimistic' scenario.

23 Several of the 'Middling' scenario applications resulted in excess deaths - up to 597 deaths (0.6%
24 increase) with a 20% reduction compared to best estimate ICU capacity.

25 All the 'Pessimistic' scenario applications resulted in excess deaths, ranging from 49,178 (17.0%
26 increase) for a 20% increase in ward bed availability, to 103,735 (35.8% increase) for a 20% shortfall in
27 ward bed availability. These scenarios took no account of the emergence of the new, more
28 transmissible, variant of concern (b.1.1.7).

29 Conclusions

30 Mortality is increased when hospital demand exceeds available capacity.

31 No excess deaths from breaching capacity would be expected under the 'Optimistic' scenario. The
32 'Middling' scenario could result in some excess deaths - up to a 0.7% increase relative to the total
33 number of deaths. The 'Pessimistic' scenario would have resulted in significant excess deaths. Our
34 sensitivity analysis indicated a range between 49,178 (17% increase) and 103,735 (35.8% increase).

35 Given the new variant, the pessimistic scenario appeared increasingly likely and could have resulted in a
36 substantial increase in the number of COVID-19 deaths. In the event, it would appear that capacity was

37 not breached at any stage at a national level with no excess deaths. it will remain unclear if minor local
38 capacity breaches resulted in any small number of excess deaths.

39 *Keywords*

40 COVID-19, demand, capacity, model, mortality, public health, infectious diseases

41 **Background**

42 The number of COVID-19 deaths in the UK was 74,125 deaths and the number of known cases was
43 2,542,069 as of the 2nd January 2021 representing a case fatality rate (CFR) of 2.9%. The Office for
44 National Statistics (ONS) Coronavirus (COVID-19) Infection Survey estimated that 8.7% of people in
45 England still had antibodies at detectable levels based upon serological testing.(1) Assuming that, as of
46 the 2nd January 2021, around 10% to 15% of the population of England has been infected (taking into
47 account those with antibodies no longer at detectable levels, the lag time and the small proportion of
48 false negative serology), then this would suggest that between 6.7 million and 10 million people had
49 already been infected representing an infection fatality rate (IFR) of 0.7% to 1.1%.

50 IFR estimates are typically made in the context of adequate capacity of health care services including
51 hospital beds and ICU beds. Should the demand for ICU beds exceed the supply, then the IFR would be
52 expected to rise. In this paper we describe a demand and capacity model designed to estimate the
53 number of COVID-19 deaths that would directly arise from a lack of ICU and ward beds. Such a model
54 currently does not exist for the UK in the public domain.

55 In November 2020 we began development of a model to estimate what increases in deaths could be
56 expected as a direct consequence of a lack of hospital bed capacity, and in particular ICU beds. Due to
57 availability of data, the model was restricted to England.

58 Estimating the ICU capacity in England is complicated, as the actual capacity at a given point of time
59 varies between usual bedbase and the maximum surge capacity created to handle increasing demand
60 during a pandemic. There were estimated to be 4,114 ICU beds in England pre-pandemic.(2) However,
61 there are plans in place to allow surges in ICU capacity when pandemics occur which entails repurposing
62 other hospital resources including anaesthetic rooms, operating theatres and parts of accident and
63 emergency departments. The aim was to be able to increase ICU bed capacity across the country by
64 100% in the event of a pandemic.(3) It is likely that the increase in capacity will vary by institution and
65 one case study managed to increase capacity by 236% during the first wave.(4) In the event of a
66 pandemic, the NHS is expected to respond through a whole system approach, which has been outlined
67 in the UK CRITCON scoring system. The basis for the system is that the same non-pandemic ethical
68 standards are applied to treat patients and allocate resources, unless in the extreme scenario (CRITCON
69 4). The system was formulated in 2009 for H1N1 and has been updated in 2020 in the context of the
70 COVID-19 pandemic. The levels described above extend from a normal capacity at CRITCON 0 to
71 CRITCON 4 when the system is overwhelmed. A “mutual aid” system exists to facilitate inter-hospital or
72 regional transfer of patients when the local intensive care capacity is breached in order to support the
73 principle that no patient should be deprived of the appropriate care if there is systemwide capacity
74 available. Eighteen intensive care networks in the England and Northern Ireland manage the capacity
75 and allocation to intensive care in the event of increasing demand.

76 At the start of the pandemic a series of field hospitals were constructed at eight sites across England.
77 The capacity provided by these field hospitals is flexible according to circumstance, but it is estimated
78 that this would potentially provide an additional 8,000 general ward beds and 500 ICU beds though
79 there may be staffing constraints that limit this number.(2)

80 In this paper we describe a model for estimating the number of additional deaths that occur from a lack
81 of capacity under different scenarios. These are illustrative scenarios and not forecasts.
82 Parameterization of the model can be modified according to available information and updated over
83 time.

84 Methods

85 Model description

86 The model can be found on GitHub: https://github.com/Crystallize/COVID19_ExceedingCapacityModel.

87 The model is a series of static compartmental models that estimates the number of COVID-19 related
88 deaths in England up to late April 2021 under three conditions: availability of both general ward and ICU
89 care; availability of general ward care but no ICU care; and no availability of either general ward or ICU
90 care.

91 The model was developed as an Excel workbook as Excel allows rapid iteration and development with
92 transparency for other developers and reviewers as it is a widely available platform.

93 General outline

94 The general outline of the model is shown in Figure 1 Outline of the structure of the model. It operates
95 using a scenario of COVID-19 hospital admission demand in weekly time steps.

96 First, we modelled the expected age distribution of admissions. This is an important input as hospital
97 and intensive care admission rates and COVID-19 mortality vary substantially by age. Subsequently, the
98 weekly demand for intensive care and ward beds was modelled by age. The weekly availability of beds
99 was calculated using estimates of the number of free beds usually available in ICUs in England plus any

100 surge capacity and additional capacity freed by the cancellation of routine surgery, field hospitals and
101 the use of private facilities adjusted for the average duration of stay on ICU and in the ward in general.

102 We then compared the weekly demand for ward and ICU beds with the maximum number of beds
103 available. This allowed for the number of weekly admissions that fall within or outside of capacity.

104 Next, we calculated an estimate of the multiplier of the mortality rates when an ICU bed is not available
105 to someone who needs it, and similarly for general ward care. This was done by compartmentalizing
106 ward and ICU patients to categories of care for which estimates were made and then aggregated up to
107 the ward or ICU level again. The assumption is that the ICU mortality rate is multiplied by 1.99 if there is
108 no ICU bed available, but there is a ward bed, and by 9.02 if there is neither a ward nor ICU bed
109 available. For general ward patients, the multiplier is 3.69 if there are no ward beds available. More
110 detail on the derivation of this is provided in the section on the “Excess Mortality Model”. Sensitivity
111 testing indicated that results were not especially sensitive to small changes in these mortality multipliers
112 (+/- 20%).

113 Once COVID-19 mortality rates for ward and ICU patients both within and outside of capacity were
114 calibrated, the number of deaths occurring ‘within capacity’ and the number that occur directly as a
115 result of being out of capacity could be calculated. The number of ‘out of capacity’ deaths does not
116 include the number of deaths that would have been expected to occur if normal care had been received,
117 and so represent ‘excess’ deaths occurring because the capacity limit was breached.

118 [Description of each step in the modelling](#)

119 We will now describe the purpose and workings of the model in more detail.

120 *Principle parameters*

121 The pre-pandemic (i.e., before March 2020) spare bed availability in the general wards in England is
122 estimated at 9,769 and the ICU spare bed capacity as 817 beds from a study of hospital capacity in the
123 COVID-19 pandemic by the Medical Research Council, Public Health England, The National Institute of
124 Health Research and Imperial College.(2) The same study estimated an additional 1,810 ICU beds and
125 52,498 beds could be acquired using field hospitals, the cancellation of routine care and the use of
126 private hospital facilities. In addition, there are pandemic response plans in place to increase ICU
127 capacity with a surge in demand. This surge capacity would be intended to increase ICU beds by 100% by
128 using anaesthetic rooms, operating theatres and other hospital resources.(3) Altogether this would
129 increase spare ICU capacity to 3,444 beds and spare ward capacity to 62,267 beds. These 62,267 ward
130 beds can provide for 87,174 patients weekly. These figures are based on an estimated length of stay on
131 ICU of 7 days and length of stay on the general ward of 5 days.(5,6)

132 [\[AdditionalFile_1_Figure_1_ModelStructure.pptx\]](#).

133 Figure 1 Outline of the structure of the model.

134 Estimating COVID-19 Admissions by Age

135 The age distribution of admissions was required in 5-year age bands up to the age of 80 years with one
136 category for those of age 80-years and over (if this data is available it can be applied directly in the
137 model removing the need for this estimation step). The age distribution in ICUs is lower than on the
138 wards with an average age of 62 years and with only a small proportion of those over 80-years being fit
139 enough for invasive ventilation.(5) However, we estimated the distribution of admissions for each age-
140 band for each calendar week by fitting an exponential curve to age-binned data on admission rates for
141 adults, then interpolating admission rates for each year of age. Using population estimates from 2019

142 for each year of age, estimates of the numbers of admissions were made.(7) These were then
143 recalibrated so that the exact number of admissions in each age group matched the observed numbers
144 in each corresponding age-band.

145 *New COVID-19 patients in hospital*

146 In this step, we applied the distribution of admissions by age to the weekly admission demand from the
147 scenarios to calculate the percentage of admissions accounted for by each year of age.

148 *Splitting COVID-19 Admissions into ICU and Ward*

149 Next, we calculated the proportion of admissions that require intensive care and ward care by age for
150 each 5-year age band, with one age band for those aged 80 years or over. This is done using data on the
151 age breakdown of admissions and total admission numbers to ICU from the ICNARC report of the 18th
152 December 2020, and the total number of COVID-19 hospital admissions up to the 18th December 2020
153 from the Gov.UK COVID-19 data dashboard.(8,9)

154 *COVID-19 Ward Demand*

155 In this step, we separated the weekly demand for general ward care by age into two tables; one for
156 those cared for within the expected hospital capacity, and a second for those who fail to receive any
157 hospital care when required because of lack of capacity.

158 *COVID-19 ICU Demand*

159 In this step we separated the weekly demand for ICU care by age into three tables; one for those cared
160 for within the expected ICU capacity, a second for those who receive only ward care when ICU care is
161 required because of no capacity, and a third for those who fail to receive any hospital care when
162 required.

163 *Excess COVID-19 Mortality Sub-Model*

164 *This is a critical step in the modelling as it is here that the estimate of the increase in COVID-19 mortality in the absence of a*
165 *ward or ITU bed is determined. It would be very challenging to directly make a meaningful estimate of the increase in mortality*
166 *overall, given the diverse range of severity of COVID-19 and the distribution of people with COVID-19 across the different*
167 *pathways that unfold. In order to reduce uncertainty, it is necessary to map the flow of individuals through treatment pathways*
168 *and consider the impact on mortality that would arise from depriving any individual requiring that element of care. The*
169 *uncertainty in estimating the impact of lack of capacity is greatly reduced by considering each element of care in the pathway*
170 *separately than when considering the impact across the pathway as a whole. Consequently, we have taken the approach of*
171 *compartmentalizing the care-pathway for COVID-19, identifying the most critical element of the pathway for different*
172 *individuals and the volume of people in those compartments from observed data. For example, for people who never need more*
173 *than high-flow oxygen on a general ward (>35% O₂), they are allocated to this compartment and their risk is not related to*
174 *lower-risk care compartments like ‘general ward care’. Similarly, for individuals who pass through general ward care and high*
175 *flow oxygen to ITU admission, continuous positive airways pressure’ support (CPAP) and eventually invasive mechanical*
176 *ventilation (IMV), then they are allocated to riskiest of these compartments – the IMV. In the latter case, we can observe the*
177 *survival from data in this group and also make a reasonably reliable judgment that if the option of IMV is removed, then the*
178 *survival will be close to zero. There will be greater uncertainty in the estimation of the impact of removing lower levels of care,*
179 *but structure can be applied to estimating the impact of depriving this care by considering the fatal events that are averted by*
180 *the care and their risk in the absence of that care. The evidence and assumptions used in setting the mortality rates in each*
181 *compartment can be found in*

182 Table 2. All required judgements were made by CM, a former clinician and the principal architect of this
183 sub-model in conjunction with RS, a physician in respiratory medicine and critical care.

184 We estimated the multiplier of COVID-19 mortality risk for ward and ICU care COVID-19 patients when
185 there is 1) no ICU capacity and 2) when there is neither ward nor ICU capacity in comparison to the in-
186 capacity mortality. The output from this sub-model is only dependent on the ‘hospitalised’ compartment
187 and the modelling is not age dependent.

188 The proportion of patients requiring intensive care was taken from the results of the modelling step
189 “Splitting Admissions into ICU and Ward” (8%). Those receiving general ward care or ICU care were
190 segmented into the categories shown in **Error! Reference source not found.**. The distribution of ward
191 care patients across the care categories was populated using observation data from the Nottingham
192 Universities Hospitals Trust.(10) The proportion of ICU patients only requiring supportive care or high
193 flow oxygen was also take from the Nottingham data. The rest of the ICU categories were populated in
194 line with published data from University Hospital Southampton for similar group of patients.(11)

195

196 *Table 1. Hospital care compartments, proportion occupancy and the mortality rates under three capacity scenarios.*

Proportion	Care segment	Proportion in each group	COVID-19 Mortality: in capacity	COVID-19 Mortality: no ICU	COVID-19 Mortality: no ward or ICU
92%	General ward care	0.6	0.01	0.01	0.03
	Ward care: O ₂ >35%	0.17	0.01	0.01	0.5
	Ward care: O ₂ >35% (Ceiling)	0.23	0.4	0.4	0.95
8%	ICU: Supportive/HFO	0.48	0.01	0.02	0.75
	ICU: NIV/CPAP	0.16	0.01	0.02	0.95
	ICU: NIV/CPAP (Ceiling)	0.12	0.83	0.9	1
	ICU: IMV+/-ECMO	0.24	0.4	1	1

197 Legend for **Error! Reference source not found.**: ICU – intensive care unit, HFO – high flow oxygen, NIV –

198 non-invasive ventilation, CPAP – continuous positive airways pressure, IMV – invasive mechanical

199 ventilation, ECMO – extracorporeal membrane oxygenation, Ceiling – this refers to the most intensive

200 level of care that is appropriate to each patient depending on their level of frailty and usually

201 determined by the Clinical Frailty Score (CFS).

202 The expected numbers of deaths in each category calculated using the assumed mortality rate and

203 weighted by the proportions in the categories using the formulae below.

204
$$q = \frac{\sum_{c=1}^{c=n} deaths_c}{hosp}$$

205 Where:

206 q = probability of dying in this episode of COVID-19

207 $deaths_c$ = the number of deaths in the care category 'c'

208 $hosp$ = the number of people hospitalised with COVID-19

209
$$m = -(1 - LN(1 - q))$$

210 Where:

211 m = hazard rate for death corresponding to the probability of dying in the scenario 'q', and

212
$$HR = \frac{m_s}{m_b}$$

213 where HR = hazard ratio applied to the base-case mortality rate m_b to find the mortality rate in
214 scenario 's'.

215 From the Excess Mortality Sub-model, we estimated a hazard ratio of 1.99 for COVID-19 mortality in
216 those needing ICU care when no ICU bed was available but there was a ward bed, and 9.02 when there
217 was neither an ICU nor ward bed available. For those needing ward care only, the hazard ratio is 3.69
218 when no hospital bed is available. Sensitivity testing indicated that results were not especially sensitive
219 to small changes in these hazard ratios (+/- 20%).

220 *Ward Mortality – Treated*

221 In this step, we estimated the in-capacity COVID-19 mortality rate by age in 5-year age bands
222 interpolating from data from the cumulative COVID-19 daily deaths report on the NHS England website

223 and using the age distributions for admissions calculated in the modelling step “Estimating Admissions
224 by Age”.(12) Data was harvested from the 6th November 2020 dataset as this precedes the peak of the
225 second wave when pressure may already have been building on internal hospital resources. After
226 estimating the mid-points of the age-bands, an exponential model is fitted to the mortality rates for the
227 three age bands from age 40 upwards. This fitted model is then used to interpolate the mortality rates
228 for the 5-year age bands required.

229 *ICU Mortality – Treated*

230 In this step, we calculated the in-capacity COVID-19 mortality rates by 5-year age bands in ICU. Data was
231 taken from the 6th November 2020 ICNARC report on COVID-19 in intensive care.(13) The mortality rate
232 for each age band was calculated from the sum of the product of the number of admissions for that age
233 band, the 28-day in-hospital mortality rate and the total number of admissions for that age band. An
234 exponential model was then fitted to the data to which age bands had been applied in order to allow
235 interpolation and re-categorising by 5-year age-bands.

236 *COVID-19 Mortality Rates*

237 In this step we took the COVID-19 mortality rates by age calculated in the previous two modelling steps
238 and the hazard ratios from the ‘Excess Mortality Model’ step to calculate the mortality rates when no
239 ICU beds and no ward beds are available respectively. The mortality rate under the scenario ‘s’ is
240 calculated using the equation:

$$241 \quad q_s = 1 - (1 - q)^{HR_s}$$

242 Where:

243 q = the mortality rate in the in-capacity scenario.

244 HR_s = the hazard ratio in scenario ‘s’

245 q_s = the mortality rate in the scenario 's'

246

247 Table 2 Expert judgements made on mortality rates by care category in the excess mortality sub-model.

248 [Note to authors – this table is supplied as a separate file

249 'BMC_IFOA_COVID_1_AdditionalFile_2_Table_2_AssumptionsExcessMortality.pptx' as it is too large to

250 fit here.]

251 Legend for Table 2: ICU – intensive care unit, HFO – high flow oxygen, NIV – non-invasive ventilation,

252 CPAP – continuous positive airways pressure, IMV – invasive mechanical ventilation, ECMO –

253 extracorporeal membrane oxygenation, CFS – clinical frailty score, Ceiling – this refers to the most

254 intensive level of care that is appropriate to each patient depending on their level of frailty and usually

255 determined by the Clinical Frailty Score (CFS).

256 *COVID-19 Deaths - In capacity*

257 In this step, we calculated the number of COVID-19 deaths each week from the corresponding number

258 of admissions generated by the 'Ward Demand' and 'ICU Demand' steps and the in-capacity mortality

259 rates from the previous 'Mortality Rates' step. Observed data is used up to the 9th December 2020, with

260 ICU deaths calculated as 15% of the total number of deaths (as there are significant delays in reporting

261 the deaths in the ICNARC data).

262 *Deaths - Outside capacity*

263 In this step, we calculated the numbers of deaths arising each week as a direct result of failing to get an

264 ICU or ward bed. Observed data is used up to the 9th December 2020. A series of seven calculations were

265 used:

- 266 1. The number of deaths in ICU within capacity as the dot product of the vector of ICU demand
267 within ICU capacity by age from step 'ICU Demand' and the vector of mortality rates from step
268 'Mortality Rates'.
- 269 2. The number of deaths in patients requiring ICU care where only ward care was available, as the
270 dot product of the vector of ICU demand outside ICU by age from step 'ICU Demand' and the
271 vector of mortality rates from step 'Mortality Rates'.
- 272 3. The number of deaths in patients requiring ICU care where neither ward nor ICU care was
273 available as the dot product of the vector of ICU demand outside both ICU and ward by age from
274 step 'ICU Demand' and the vector of mortality rates from step 'Mortality Rates'.
- 275 4. The number of deaths in patients requiring ward care only as the dot product of the vector of
276 ward demand within ward capacity by age from step 'Ward Demand' and the vector of mortality
277 rates from step 'Mortality Rates'.
- 278 5. The number of deaths in patients requiring ward care where ward care was not available as the
279 dot product of the vector of ward demand outside ward capacity by age from step 'Ward
280 Demand' and the vector of mortality rates from step 'Mortality Rates'.
- 281 6. The total number of deaths each week under the current scenario including both ICU and ward
282 deaths as the sum of all results in steps 1-5.
- 283 7. The total number of deaths in each calendar week in both the ICU and ward care groups
284 calculated in the step 'Deaths - In capacity' was taken and subtracted from the capacity
285 constrained total in step 6 to provide the number of deaths arising as a direct result of the lack
286 of an ICU or ward bed.

287 *Scenarios*

288 We applied three scenarios for admission demand from the week beginning 16th December 2020
289 through to the week beginning 28th April 2021 (see Figure 2).

290 [\[AdditionalFile_3_Figure_2_AdmissionsDemandScenario\]](#).

291 In this model, the three scenarios are referred to as “Pessimistic”, “Middling” and “Optimistic” and were
292 arbitrary but based on the trajectory of the increase in the multiplier of admissions from one week to
293 the next. The trajectory was calculated as a linear regression of the last three weeks of the known data
294 taken from the GOV.UK Coronavirus dashboard.(8) They are illustrative scenarios and not forecasts.
295 With the passage of time, further data through to the week of the 3rd of February was obtained for
296 comparison with the scenarios. The peak in admissions actually occurred around the week of the 13th of
297 January.

298 In the Middling scenario, the multiplier continued to increase following the rate over the previous 3
299 weeks up to three weeks post the introduction of more vigorous countermeasures on 19th December
300 2020. This is to be expected as the incubation period is nearly a week, so the enhanced
301 countermeasures would usually take at least a week to result in any observed change. The multiplier
302 falls below 1.0 after 4 more weeks. The emergence of a new variant of SARS-CoV-2 with higher
303 transmission rates that became more prevalent from the beginning of December has driven more rapid
304 growth of infection rates. As the new variant makes up a greater proportion of new cases, the rate of
305 growth will continue to rise unless effectively mitigated. It remains uncertain, at the time of writing, how
306 it will respond to tighter controls, but there is no reason to assume the mortality rates will differ at
307 present. This new variant has only recently been identified and has not been factored into the analysis.
308 However, initial indications are that this variant could open up a variety of more pessimistic scenarios in
309 the short term.

310 We now have effective vaccines and it is expected that their use will prevent a large number of
311 infections and many deaths (14).

312 In the Optimistic scenario, the multiplier begins to decelerate from week beginning 16th December,
313 continues to decelerate after the 16th December and falls below 1.0 after 3 weeks.

314 In the Pessimistic scenario, the multiplier continued to increase for 4 weeks after week beginning 16th
315 December before decelerating at the same rate as the other two scenarios. This is not intended as a
316 worst-case scenario, but one that is somewhat worse than the 'Middling' scenario.

317 Figure 2 Graph showing three scenarios of projected admission demand from the 9th December 2020.

318 *Sensitivity analysis*

319 We performed an analysis to assess the sensitivity of the model to the choice of capacity and hazard
320 ratio parameter values and scenario. Five input parameters were varied independently in turn and the
321 effect on the model output recorded. The five parameters varied were the ICU and ward capacities and
322 the hazard ratios for mortality in ward patients when there are no ward beds available and ICU patients
323 when there are no ICU beds and ICU patients when there are no ward beds available. The hazard ratios
324 were adjusted using the equation.

325
$$HR^* = 1 + (HR - 1) \cdot (1 + d)$$

326 Where:

327 HR = the within-capacity hazard ratio.

328 HR^* = the outside-capacity hazard ratio.

329 d = the adjustment to be applied as a proportion.

330 This model has been populated with data specific to England, it could be applied in other geographies
331 and we have added ways of estimating some data like the age distribution of admissions or the ICU
332 admission rates by age. Where this data is directly available it can be substituted.

333 *Assumptions*

334 There are several general underlying assumptions to the model.

- 335 • There is no change in the expected mortality as demand grows until the bed capacity is
336 breached.
- 337 • Only bed capacity affects excess mortality rates and not other resource constraints such as
338 staffing, equipment, ambulance availability or other finite resources, although actual system
339 capacity may be a product of all of the above.
- 340 • Health care policy and conscious or unconscious clinician behaviour that may affect the
341 compartments of care through which the patients flow does not change as the pandemic waxes
342 and wanes.
- 343 • Surge capacity will be applied before the use of field hospitals or the commandeering of private
344 facilities.
- 345 • All the surge capacity, field hospital and private facility beds will be utilised for COVID-19 rather
346 than any other demand.
- 347 • The age distribution of admissions remains constant over time.
- 348 • There is perfect redistribution of bed capacity and resources across the country immediately
349 according to demand.
- 350 • Weekly bed availability remains constant.
- 351 • Capacity estimates are fixed throughout the model when, in fact, new resources may be
352 recruited over time, or capacity may fall in response to disruption of staffing and supplies.

- 353 • The mortality rates, including by age, are the same for the old COVID strain and novel
354 mutations.

355 Results

356 The key output of our collaboration was the model itself rather than the results of any of the scenarios.

357 The model allows a user to understand the excess COVID-19 mortality impact arising as a direct
358 consequence of ward and/or ICU capacity being breached under various scenarios or forecasts of
359 hospital admissions. The scenarios described in this paper are illustrative and are not forecasts.

360 Table 3. Numbers of deaths with the unadjusted middling scenario from 16th of December (bold font).

361 [Insert Table 3 here – Table 3 at end of file as per BMC guidelines for tables larger than 1A4 in length]

362 Sensitivity analyses

363 The results of the sensitivity analysis are shown in Table 4 and Figure 3. None of the Optimistic scenario
364 adjustments resulted in any excess deaths. In the Middling scenario adjustments, the excess deaths
365 attributable to lack of capacity alone ranged from 0 to 597 with a 20% reduction in the expected number
366 of ICU beds. In the Pessimistic scenario the number of excess deaths attributable directly to lack of
367 capacity range from 49,178 with a 20% increase in the number of ward beds, and 103,735 with a 20%
368 reduction in the expected ward bed availability.

369 Figure 3. Results of the sensitivity analysis.

370 [Additional File 4_Figure_3_SensitivityAnalysis]

371 The results can be explained by considering how capacity evolves in each of the scenarios. In the
372 Middling scenario, whilst ICU capacity may be approached and even possibly breached, there remains
373 sufficient ward capacity to take lives who need either ward or ICU support, keeping excess deaths

374 relatively low. However, the Pessimistic scenario sees ward capacity breached, and in many scenarios
375 for a period of several weeks, resulting in much higher mortality in those lives who require care but do
376 not receive it. ICU capacity is much lower than ward capacity and only a small proportion of all
377 hospitalized patients need ICU care so the number of deaths from breaches of ward capacity are
378 proportionally larger than breaches for ICU care. ICU care is assumed to reach capacity before ward
379 care, so with marginal breaches of capacity, ICU breaches are the source of the excess deaths, but with
380 large breaches of capacity, the majority will arise from breaching of ward capacity.

381 The number of excess deaths is most sensitive to ward bed availability in the pessimistic scenario with a
382 difference of 54,556 excess deaths between the 20% increase and 20% decrease in the bed availability
383 compared to 5,129 with the same variation in ICU bed availability. However, in the Middling scenario,
384 the excess deaths are most sensitive to ICU bed capacity with a difference of 597 with +/- 20% variation
385 in the ICU bed capacity estimate, with no difference arising from the ward bed capacity estimate.

386 These results are shown graphically in Additional File 3_SensitivityAnalysis (note the different y-axis
387 scales). Excess deaths are more sensitive to the availability of ICU or ward beds than to the adjustments
388 in the hazard ratios used here. The greatest loss of life occurs with a 20% reduction in the estimate of
389 ward availability in the pessimistic scenario with 103,735 (35.8% increase) excess deaths.

390 The actual observations including data from the 16th December through to the week of the 3rd of
391 February showed a peak weekly admission rate of 29,447 in the week of the 13th of January before
392 declining (Figure 2). There would have been no excess deaths due to lack of capacity in the observed
393 cases through to February under this set of assumptions.

394 Discussion

395 Table 4. Results of the sensitivity analysis.

396 [Insert Table 4 here – Table 4 at end of file as per BMC guidelines for tables larger than 1A4 in length]

397

398 The new variant of SARS-CoV-2 that emerged in the UK (B.1.1.7) appears to be about 56% (95% CI 50%-
399 74%) more transmissible than the existing variants and appears to have a higher case-fatality rate(15,16)
400 Alternative explanations for its rise in prevalence and the increased rate of transmission observed since
401 its appearance at the beginning of October were investigated including “immune escape” where
402 individuals previously infected return to susceptibility as a result of mutation of key antigens, increased
403 susceptibility amongst children, and a shorter generation time. None of these alternative explanations
404 fitted the data as well as increased infectiousness. In addition, a variant has emerged from South Africa
405 (501Y.V2) which also appears to have greater transmissibility and an increased viral load.(17) These
406 unsettling developments increases uncertainty in the future trajectory of hospital demand and open up
407 a variety of significantly more pessimistic scenarios in the short term that have not been possible to
408 explore here.

409 In the longer term, there are now three vaccines approved for use in the UK, as of the 19th January 2021
410 (18). It is expected they will have a substantial impact on admissions and deaths, though not until after
411 the majority of deaths in our scenarios are modelled as occurring (19). Consequently, we have not
412 factored the application of vaccination into the scenarios as they are operating in the short-term.

413 Limitations

414 Here we list and discuss the limitations of the model:

- 415
- The estimate of bed availability is determined using the mean length of stay on the ward or in
- 416 ICU. The distribution of occupants by length of stay will change over time which may result in a
- 417 slow consumption of capacity that is not captured in this model. For example, a higher
- 418 proportion of afflicted younger patients in hospital may lead to longer bed occupancy both in
- 419 ward and critical care, as withdrawal of active treatment in view of futility is less likely to take
- 420 place in this group.
- Mortality rates are affected by constraints other than just bed availability including staffing and
- 421 equipment.
- The model will not capture the transition between low mortality with full capacity and the high
- 422 mortality from lack of a bed that arises from stressing of the system before capacity is
- 423 absent.(20)
- Capacity estimates are fixed throughout the model when, in fact, new resources may be
- 424 recruited over time, or capacity may fall in response to disruption of staffing and supplies.
- Age specific case fatality rates are assumed to be static, but in fact may change, either due to
- 425 changes in virulence, improvement in care or dilution of standards of care during a surge
- 426 scenario.
- The oxygen cut-off of 35% is based on clinical opinion, and there is no empirical study in support
- 427 of this threshold for obvious reasons. However, often clinical practice may be to err on the side
- 428 of caution in a ward setting and to keep a patient on a slightly higher fractionated oxygen than
- 429 s/he needs and therefore a patient on a lower than 35% FiO₂ may actually need even smaller
- 430 amounts of oxygen, and therefore the assumption is that s/he would withstand the lack of
- 431 medical oxygen with a degree of success.
- 432
- 433
- 434
- 435
- 436
- 437

438 Intensive care resources are constrained not only by bed availability but also by equipment and staffing.
439 A combination of all three is required for optimal care in ICU. In reality there is not a simple binary state
440 of presence or absence of these factors and skills. A degradation of equipment maintenance, training
441 around new equipment and distribution as well as reductions in the effectiveness of staff, either
442 because staff-to-patient ratios fall as demand rises, or because of staff sickness due to COVID-19 or
443 simply the physical and emotional fatigue as the pandemic continues is also seen. An analysis of within-
444 capacity mortality rates in ICU by bed occupancy has found that there is an almost linear increase in
445 mortality with no excess mortality at 0% bed occupancy to a 92% increase in mortality by 100%
446 occupancy.(20)

447 The modelling by McCabe et al suggests that ICU capacity is first constrained by bed availability, though
448 lack of nurses and junior doctors is close behind.(2) They included sickness absence rates taken from
449 surveys of union members suggesting 15% of doctors were off sick in the first wave and may under-
450 estimate the impact of sickness on staffing overall at peak times in the pandemic.(21) As the pandemic
451 progresses, higher than anticipated absence due to sickness in these groups could result in lack of
452 limiting capacity before the lack of beds, both in ICU and in wards. Reorganisation within hospitals may
453 mitigate this by training other staff to support ICU work and thereby increasing ICU staff to bed
454 ratios.(22) One hospital managed to meet demand in the first wave, but reduced ICU nurse to bed ratios
455 from the normal pre-pandemic of 1:1 to 1:4.(4) Current guidance on nursing staff ratios during the
456 COVID-19 pandemic advocates a ratio of 1:2.(22)

457 Decision making may change in the face of increasing demand either with formal revisions of treatment
458 thresholds as resources become increasingly scarce or with the introduction of triaging. On a less formal
459 level, the heuristics used by clinicians in their everyday management of patients may vary as competing
460 pressures rise. For example, at times of abundant capacity, the thresholds for transferring patients into

461 ICU for a trial to see if a patient with poor chance of survival recovers with a short stay, may be lower
462 than when there is severe limitation on capacity.

463 One of the most uncertain parts of the modelling is the determination of the multipliers of risk in the
464 section 'Excess Mortality Submodel'. We estimated that the multiplier for the scenario of no ICU bed
465 when one was required was 1.99. A recent observational study of 4,032 ICU admissions across 114
466 hospital trusts found a rising mortality with bed occupancy (20). The estimated excess mortality rate was
467 estimated to reach 92% at 100% capacity. This is very close to our estimate of 99% and is a reassuring
468 triangulation of the method.

469 An important but unrealistic assumption in the model is that there is perfect distribution of resources
470 with respect to demand across the country and that no patient is refused care whilst there remains a
471 bed anywhere in England. ICUs are organized into networks which facilitate the transfer of patients from
472 one hospital to another when ICU beds in a hospital run out, but even a delay of a few hours in
473 transferring a patient to an ICU bed can influence outcome. Furthermore, even in the case of bed
474 availability in a different hospital, it cannot be assumed that the patient in the referring and at-capacity
475 hospital may be fit enough to be able to be transferred safely. All the ICUs in a network are likely to have
476 correlated demand and may all run out of space at the same time. In these circumstances, transfers
477 would need to be arranged between networks or regions and this would entail yet further delay and
478 challenges with consequent impact on outcomes.

479 It is notable that the sensitivity of the excess deaths to ICU capacity is much greater than it is to ward
480 bed capacity at marginal breaches of capacity, but that large scale breaches of capacity are more
481 sensitive to ward bed capacity. There are far more ward beds than ICU beds, and there is an assumption
482 that ICU capacity would be exhausted long before ward bed capacity.

483 *Conclusions*

484 Here we describe a demand and capacity model for general hospital and intensive care beds in the
485 context of the COVID-19 pandemic in England. The model allows a user to understand the excess COVID-
486 19 mortality impact arising as a direct consequence of capacity being breached under various scenarios
487 or forecasts of hospital admissions. The scenarios described in this paper are illustrative and are not
488 forecasts.

489 We estimated the number of excess COVID-19 deaths up to the end of April 2021 that would arise from
490 lack of ward and ICU capacity under different demand assumptions from December 16th, 2020. No
491 excess deaths from excess capacity would be expected under the 'Optimistic' assumptions of demand
492 but would reach between 49,178 and 103,735 under the 'Pessimistic' scenario. Without the new variant,
493 exceeding capacity for hospital and ICU beds was not the most likely outcome but given the new variant
494 it appeared more plausible and could have result in a substantial increase in the number of COVID-19
495 deaths. In the event, it would appear that capacity was not breached at a national level any stage, and,
496 under an assumption of a perfectly even distribution of demand and capacity, there would be no excess
497 deaths due to lack of capacity expected under this set of assumptions. However, distribution of demand
498 and capacity is imperfect. It will remain unclear if minor local capacity breaches resulted in any small
499 number of excess deaths.

500 *List of abbreviations*

501 **CFR:** Case fatality rate

502 **CFS:** Clinical frailty score

503 **CPAP:** Continuous positive airway pressure

- 504 **ECMO:** Extracorporeal membrane oxygenation
- 505 **HFO:** High flow oxygen
- 506 **HR:** Hazard ratio
- 507 **ICNARC:** Intensive care national audit and research centre
- 508 **ICU:** Intensive care unit
- 509 **IFR:** Infection fatality rate
- 510 **IMV:** Invasive mechanical ventilation
- 511 **NIV:** Non-invasive ventilation
- 512 **ONS:** Office of national statistics
- 513 **SARS-CoV-2:** Severe acute respiratory syndrome coronavirus 2
- 514 **VOC:** Variant of concern
- 515 Declarations
- 516 [Ethics approval and consent to participate](#)
- 517 Not applicable
- 518 [Consent for publication](#)
- 519 Not applicable
- 520 [Availability of data and materials](#)
- 521 The dataset supporting the conclusions of this article is available GitHub:
- 522 [Crystallize/COVID19_ExceedingCapacityModel \(github.com\)](#).

523 [Competing interests](#)

524 RS received writing fees for healthcare reports from Crystallise UK Ltd. None of the other authors
525 declare any conflict of interest.

526 [Funding](#)

527 No funding source.

528 [Authors' contributions](#)

529 Stuart McDonald initiated and coordinated the project and was one of the principal architects of the
530 general model and contributed to the final paper.

531 Chris Martin was the principal architect of the excess mortality sub-model, participated in the
532 development of the general model and pulled together the initial draft of the paper.

533 Michiel Luteijn participated in the development of the general model and undertook ancillary modelling
534 that supported development and contributed to the final paper.

535 Steve Bale participated in the development of the general model and contributed to the final paper.

536 Rahul Sarkar contributed to the expert judgements in the modelling, advised on the organizational
537 context within which the model sits, and contributed to the final paper.

538 [Acknowledgements](#)

539 [Various](#)

540 Colin Dutkiewicz, (Global Head of Life Reinsurance Solutions, Aon)

541 Josephine Robertson, (Health & Care Actuary, Optum)

542 Dan Ryan, (Chief Science Officer at COIOS Research)

- 543 Scott Reid, (Global Protection Pricing & Product Development Actuary, Zurich)
- 544 James Robinson, (UK & Ireland Business Strategy and Analytics, Aon Benfield)
- 545 [Nottingham Universities Hospital Trust](#)
- 546 Joe West, (Professor of Epidemiology; Honorary Consultant Gastroenterologist, Faculty of Medicine &
547 Health Sciences Nottingham University)
- 548 Richard Hubbard, (GSK/British Lung Foundation Professor of Respiratory Epidemiology, Faculty of
549 Medicine & Health Sciences)
- 550 Tim Card, (Clinical Associate Professor, Nottingham Biomedical Research Centre)
- 551 Colin Crooks, (Clinical Associate Professor, Nottingham Biomedical Research Centre)
- 552 Nina Lewis, (Consultant Gastroenterologist, Nottingham University NHS Trust)
- 553 Andrew Fogarty (Clinical Associate Professor & Reader in Clinical Epidemiology, Faculty of Medicine &
554 Health Sciences)
- 555 Andrew Marshall, (Deputy Medical Director at Nottingham University Hospitals)
- 556 Thearina De Beer, (Consultant in Anaesthetics and ICM at Nottingham University Hospitals NHS Trust)
- 557 Dominic Shaw, (Professor and Honorary Consultant, Faculty of Medicine & Health Sciences)
- 558 Emma O’Dowd, (Honorary Consultant Assistant Professor, University of Nottingham)
- 559 Andrew Baraclough, (Assistant Director of Insight, Nottingham University Hospitals NHS Trust)
- 560 [Crystallise](#)
- 561 Shannon Connolly, Analysts at Crystallise Ltd.

562 Dr Will Letton, Consultant Analyst at Crystallise

563 Dr Louisa Rutherford, Senior Clinical Researcher at Crystallise Ltd.

564 **Additional files**

565 https://github.com/Crystallize/COVID19_ExceedingCapacityModel.

566 Link to model on GitHub

567 BMC_IFOA_COVID_1_AdditionalFile_1Figure_1_ModelStructure.pptx

568 Title: "Figure 1. Outline of structure of model"

569 BMC_IFOA_COVID_1_AdditionalFile_2_Table_2_AssumptionsExcessMortality.pptx

570 Title: "Table 2. Expert judgements made on mortality rates by care category in the excess mortality sub-
571 model."

572 BMC_IFOA_COVID_1_AdditionalFile_3_Figure_2_AdmissionsDemandsScenarios.pptx

573 Title: "Figure 2. Graph showing three scenarios of projected admission demand from the 9th December
574 2020"

575 BMC_IFOA_COVID_1_AdditionalFile_4_Figure_3_SensitivityAnalysis.pptx

576 Title: "Figure 3. Results of the sensitivity analysis"

577 **Author details**

578 ¹Director of Modelling at Crystallise, Unit 19, Saffron Court, Southfields Business Park, Basildon, Essex,
579 SS15 6SS.

580 ²Honorary Researcher, University College London.

581 ³Head of Demographic Assumptions and Methodology at Lloyds Banking Group, 25 Gresham Street,
582 London, EC2V 7HN

583 ⁴Senior Actuary at Munich Re UK Life Branch, 10 Fenchurch Avenue, London EC3M 5BN

584 ⁵Biometric Research Data Specialist at Hannover Re UK Life branch. 10 Fenchurch Street, London, EC3M
585 3BE.

586 ⁶Consultant Physician in Respiratory Medicine & Critical Care at Medway NHS Foundation Trust.
587 Windmill Road, Gillingham, Kent ME7 5NY

588

589 References

590 1. ONS. Coronavirus (COVID-19) Infection Survey: characteristics of people testing positive for
591 COVID-19 in England and antibody data for the UK: December 2020. Off Natl Stat.
592 2020;(December).

593 2. McCabe R, Schmit N, Christen P, D'Aeth JC, Løchen A, Rizmie D, et al. Adapting hospital capacity
594 to meet changing demands during the COVID-19 pandemic. BMC Med [Internet]. 2020 Dec
595 16;18(1):329. Available from: [https://bmcmmedicine.biomedcentral.com/articles/10.1186/s12916-](https://bmcmmedicine.biomedcentral.com/articles/10.1186/s12916-020-01781-w)
596 020-01781-w

597 3. NHS England. Management of surge and escalation in critical care services : standard operating
598 procedure for adult critical care [Internet]. 2013 [cited 2020 Feb 15]. Available from:
599 [https://www.england.nhs.uk/commissioning/wp-content/uploads/sites/12/2013/11/sop-](https://www.england.nhs.uk/commissioning/wp-content/uploads/sites/12/2013/11/sop-burns.pdf)
600 burns.pdf

601 4. Chong MSF, Hla TW, Sartori G. ICU surge capacity in a busy London district general hospital
602 during the COVID-19 pandemic [Internet]. Association of anaesthetists. 2020 [cited 2020 Dec 20].

- 603 Available from: <https://anaesthetists.org/Home/Resources-publications/COVID-19-guidance/ICU->
- 604 [surge-capacity-in-a-busy-London-district-general-hospital-during-the-COVID-19-pandemic](https://anaesthetists.org/Home/Resources-publications/COVID-19-guidance/ICU-surge-capacity-in-a-busy-London-district-general-hospital-during-the-COVID-19-pandemic)
- 605 5. ICNARC. ICNARC report on COVID-19 in critical care: England, Wales and Northern Ireland 24
- 606 December 2020 [Internet]. 2020. Available from: [https://www.icnarc.org/Our-](https://www.icnarc.org/Our-Audit/Audits/Cmp/Reports)
- 607 [Audit/Audits/Cmp/Reports](https://www.icnarc.org/Our-Audit/Audits/Cmp/Reports)
- 608 6. Rees EM, Nightingale ES, Jafari Y, Waterlow NR, Clifford S, Carl CA, et al. COVID-19 length of
- 609 hospital stay: A systematic review and data synthesis. *BMC Med.* 2020;18(1).
- 610 7. Human Mortality Database [Internet]. 2020. Available from: <https://www.mortality.org/>
- 611 8. GOV.UK Coronavirus (COVID-19) in the UK [Internet]. 2020 [cited 2020 Dec 22]. Available from:
- 612 <https://coronavirus.data.gov.uk/>
- 613 9. ICNARC. ICNARC report on COVID-19 in critical care: England, Wales and Northern Ireland. 18
- 614 December 2020. 2020;(18th december). Available from:
- 615 [file:///C:/Users/ChrisMartin/AppData/Local/Temp/MicrosoftEdgeDownloads/9372fc02-31b0-](file:///C:/Users/ChrisMartin/AppData/Local/Temp/MicrosoftEdgeDownloads/9372fc02-31b0-41b4-9570-ff5bdb63fd1e/ICNARC_COVID-19_Report_2020-12-18.pdf)
- 616 [41b4-9570-ff5bdb63fd1e/ICNARC_COVID-19_Report_2020-12-18.pdf](file:///C:/Users/ChrisMartin/AppData/Local/Temp/MicrosoftEdgeDownloads/9372fc02-31b0-41b4-9570-ff5bdb63fd1e/ICNARC_COVID-19_Report_2020-12-18.pdf)
- 617 10. Crooks CJ, West J, Fogarty A, Morling JR, Grainge MJ, Gonem S, et al. Predicting the need for
- 618 escalation of care or death from repeated daily clinical observations and laboratory results in
- 619 patients with SARS-CoV-2 during 2020: a retrospective population-based cohort study from the
- 620 United Kingdom. *MedRxiv* [Internet]. 2020; Available from:
- 621 <https://www.medrxiv.org/content/10.1101/2020.12.14.20248181v1>
- 622 11. Sivaloganathan AA, Nasim-Mohi M, Brown MM, Abdul N, Jackson A, Fletcher S V., et al.
- 623 Noninvasive ventilation for COVID-19-associated acute hypoxaemic respiratory failure:
- 624 experience from a single centre. *Br J Anaesth.* 2020;125(4):e368–71.

- 625 12. NHS England. COVID 19 total announced deaths 23 December 2020 [Internet]. 2020 [cited 2020
626 Dec 23]. Available from: [https://www.england.nhs.uk/statistics/statistical-work-areas/covid-19-](https://www.england.nhs.uk/statistics/statistical-work-areas/covid-19-daily-deaths/)
627 [daily-deaths/](https://www.england.nhs.uk/statistics/statistical-work-areas/covid-19-daily-deaths/)
- 628 13. ICNARC. Report on COVID-19 in critical care. Intensive Care Natl Audit Res Cent [Internet].
629 2020;(April):1–24. Available from: [https://www.icnarc.org/About/Latest-](https://www.icnarc.org/About/Latest-News/2020/03/22/Report-On-196-Patients-Critically-Ill-With-Covid-19)
630 [News/2020/03/22/Report-On-196-Patients-Critically-Ill-With-Covid-19](https://www.icnarc.org/About/Latest-News/2020/03/22/Report-On-196-Patients-Critically-Ill-With-Covid-19)
- 631 14. Voysey M, Clemens SAC, Madhi SA, Weckx LY, Folegatti PM, Aley PK, et al. Safety and efficacy of
632 the ChAdOx1 nCoV-19 vaccine (AZD1222) against SARS-CoV-2: an interim analysis of four
633 randomised controlled trials in Brazil, South Africa, and the UK. *Lancet* [Internet]. 2021
634 Jan;397(10269):99–111. Available from:
635 <https://linkinghub.elsevier.com/retrieve/pii/S0140673620326611>
- 636 15. Davies NG, Barnard RC, Jarvis CI, Kucharski AJ, Munday J, Pearson CA, et al. Estimated
637 transmissibility and severity of novel SARS-CoV-2 Variant of Concern 202012/01 in England.
638 Github [Internet]. 2020; Available from: [https://cmmid.github.io/topics/covid19/reports/uk-](https://cmmid.github.io/topics/covid19/reports/uk-novel-variant/2020_12_23_Transmissibility_and_severity_of_VOC_202012_01_in_England.pdf)
639 [novel-variant/2020_12_23_Transmissibility_and_severity_of_VOC_202012_01_in_England.pdf](https://cmmid.github.io/topics/covid19/reports/uk-novel-variant/2020_12_23_Transmissibility_and_severity_of_VOC_202012_01_in_England.pdf)
- 640 16. Iacobucci G. Covid-19: New UK variant may be linked to increased death rate, early data indicate.
641 *BMJ* [Internet]. 2021 Jan 26;n230. Available from:
642 <https://www.bmj.com/lookup/doi/10.1136/bmj.n230>
- 643 17. SARS-CoV-2 Variants [Internet]. World Health Organisation. 2020 [cited 2021 Jan 4]. Available
644 from: [https://www.who.int/csr/don/31-december-2020-sars-cov2-variants/en/#:~:text=On 18](https://www.who.int/csr/don/31-december-2020-sars-cov2-variants/en/#:~:text=On%2018%20December,national%20authorities,because%20of%20a%20N501Y%20mutation.)
645 [December%2C national authorities,because of a N501Y mutation.](https://www.who.int/csr/don/31-december-2020-sars-cov2-variants/en/#:~:text=On 18)
- 646 18. Moderna vaccine becomes third COVID-19 vaccine approved by UK regulator [Internet]. Gov.UK.

- 647 2021 [cited 2021 Jan 19]. Available from: <https://www.gov.uk/government/news/moderna->
648 [vaccine-becomes-third-covid-19-vaccine-approved-by-uk-regulator](https://www.gov.uk/government/news/moderna-vaccine-becomes-third-covid-19-vaccine-approved-by-uk-regulator)
- 649 19. Cook TM, Roberts J V. Impact of vaccination by priority group on UK deaths, hospital admissions
650 and intensive care admissions from COVID-19. *Anaesthesia* [Internet]. 2021 Feb 11;anae.15442.
651 Available from: <https://onlinelibrary.wiley.com/doi/10.1111/anae.15442>
- 652 20. Wilde H, Mellan T, Hawryluk I, Dennis J, Denaxas S, Pagel C, et al. The association between
653 mechanical ventilator availability and mortality risk in intensive care patients with COVID-19: A
654 national retrospective cohort study. *MedRxiv* [Internet]. 2021; Available from:
655 <https://www.medrxiv.org/content/medrxiv/early/2021/01/13/2021.01.11.21249461.full.pdf>
- 656 21. Booth R. Number of NHS doctors off sick ‘may be nearly triple the official estimate.’ *The*
657 *Guardian*. 2020.
- 658 22. NHS. Advice on acute sector workforce models during COVID-19 [Internet]. 2020. Available from:
659 [https://www.england.nhs.uk/coronavirus/wp-content/uploads/sites/52/2020/12/C0833_advice-](https://www.england.nhs.uk/coronavirus/wp-content/uploads/sites/52/2020/12/C0833_advice-on-acute-sector-workforce-models-during-COVID_with-apps_10dec.pdf)
660 [on-acute-sector-workforce-models-during-COVID_with-apps_10dec.pdf](https://www.england.nhs.uk/coronavirus/wp-content/uploads/sites/52/2020/12/C0833_advice-on-acute-sector-workforce-models-during-COVID_with-apps_10dec.pdf)
- 661 [Table 3 below]

Week	Total COVID-19 Deaths (with ICU and Ward capacity limits)	COVID-19 deaths if there are no ICU and Ward capacity limits.	Excess COVID-19 deaths specifically due to ICU and Ward capacity limits
30/09/2020	298	298	0
07/10/2020	444	444	0
14/10/2020	731	731	0
21/10/2020	1,074	1,074	0
28/10/2020	1,457	1,457	0
04/11/2020	1,784	1,784	0
11/11/2020	1,897	1,897	0
18/11/2020	2,103	2,103	0
25/11/2020	2,031	2,031	0
02/12/2020	1,889	1,889	0
09/12/2020	1,960	1,960	0
16/12/2020	2,685	2,685	0
23/12/2020	3,703	3,703	0
30/12/2020	5,697	5,697	0
06/01/2021	9,677	9,677	0
13/01/2021	11,689	11,612	77
20/01/2021	11,689	11,612	77
27/01/2021	9,289	9,289	0
03/02/2021	7,432	7,432	0
10/02/2021	5,945	5,945	0

17/02/2021	4,756	4,756	0
24/02/2021	3,805	3,805	0
03/03/2021	3,044	3,044	0
10/03/2021	2,435	2,435	0
17/03/2021	1,705	1,705	0
24/03/2021	1,193	1,193	0
31/03/2021	835	835	0
07/04/2021	585	585	0
14/04/2021	409	409	0
21/04/2021	286	286	0
28/04/2021	201	201	0
Total	102,728	102,573	154

662

663 [Table 4 below]

Scenario	Ward availability	ICU availability	No ward HR	No ICU HR	No ICU or Ward HR	Excess COVID-19 deaths due to capacity breach	Total COVID-19 deaths	% increase in COVID-19 deaths
Middling	-20%	0%	0%	0%	0%	102,573	154	102,728
Middling	0%	0%	0%	0%	0%	102,573	154	102,728

Middling	20%	0%	0%	0%	0%	102,573	154	102,728
Middling	0%	-20%	0%	0%	0%	102,573	597	103,170
Middling	0%	20%	0%	0%	0%	102,573	0	102,573
Middling	0%	0%	-20%	0%	0%	102,573	154	102,728
Middling	0%	0%	20%	0%	0%	102,573	154	102,728
Middling	0%	0%	0%	-20%	0%	102,573	130	102,703
Middling	0%	0%	0%	20%	0%	102,573	176	102,750
Middling	0%	0%	0%	0%	-20%	102,573	154	102,728
Middling	0%	0%	0%	0%	20%	102,573	154	102,728
Optimistic	0%	0%	0%	0%	0%	47,073	0	47,073
Pessimistic	-20%	0%	0%	0%	0%	289,983	103,735	393,718
Pessimistic	0%	0%	0%	0%	0%	289,983	73,711	363,694
Pessimistic	20%	0%	0%	0%	0%	289,983	49,178	339,161
Pessimistic	0%	-20%	0%	0%	0%	289,983	76,382	366,365
Pessimistic	0%	20%	0%	0%	0%	289,983	71,253	361,236
Pessimistic	0%	0%	-20%	0%	0%	289,983	66,576	356,559
Pessimistic	0%	0%	20%	0%	0%	289,983	79,764	369,747
Pessimistic	0%	0%	0%	-20%	0%	289,983	73,472	363,455
Pessimistic	0%	0%	0%	20%	0%	289,983	73,926	363,909
Pessimistic	0%	0%	0%	0%	-20%	289,983	72,943	362,926
Pessimistic	0%	0%	0%	0%	20%	289,983	74,197	364,180

Figures

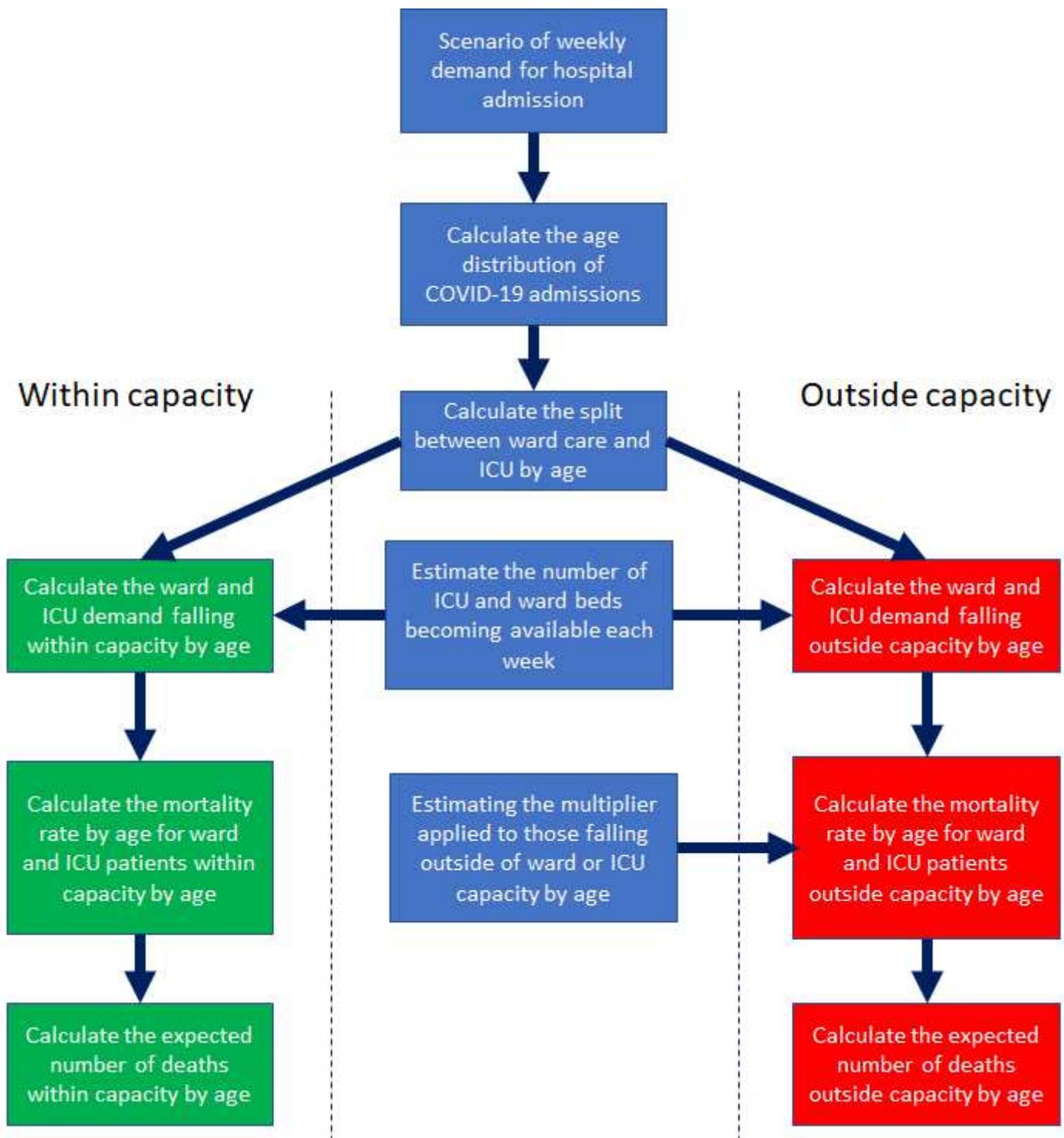


Figure 1

Outline of the structure of the model.

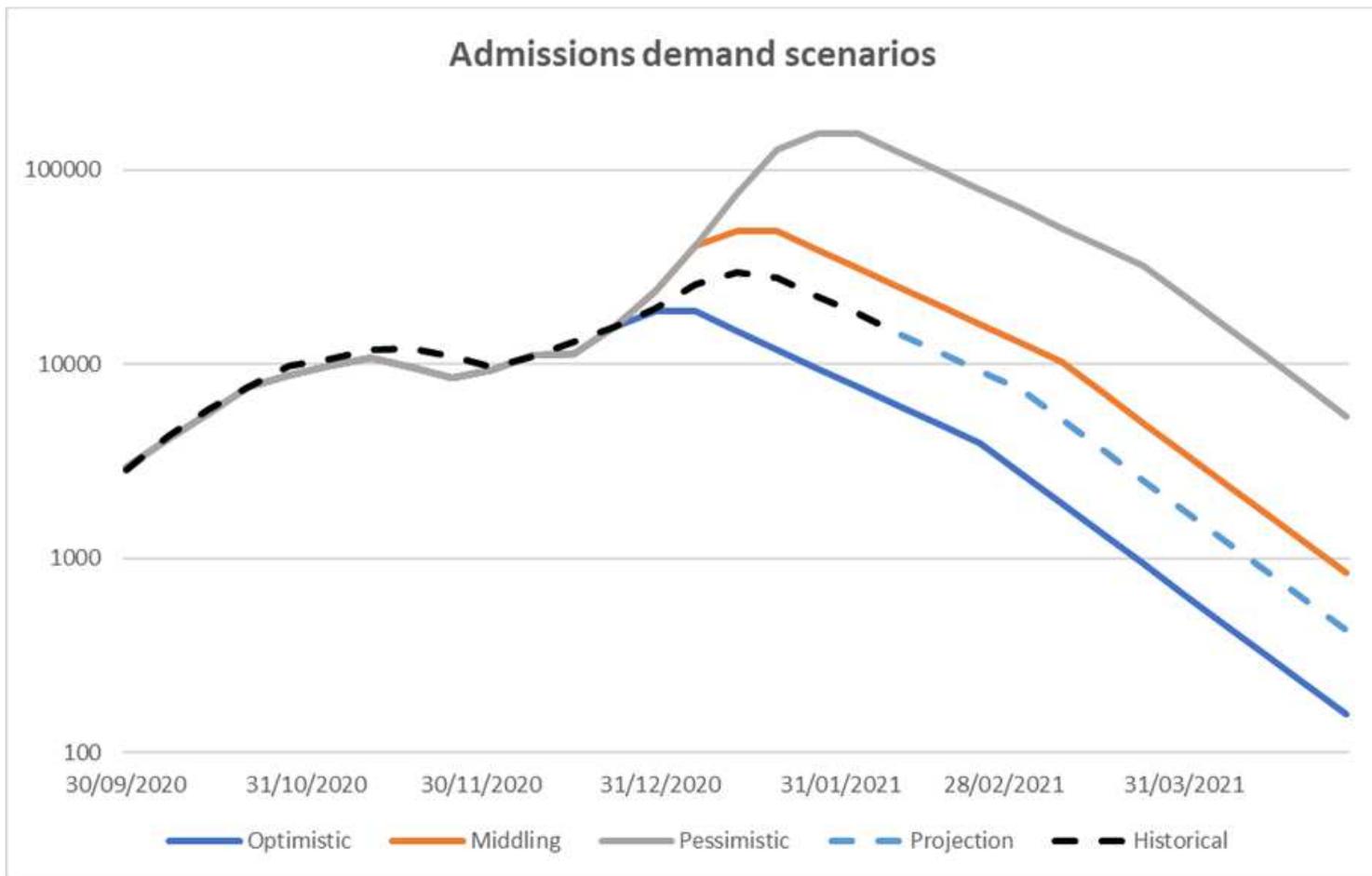


Figure 2

Graph showing three scenarios of projected admission demand from the 9th December 2020.

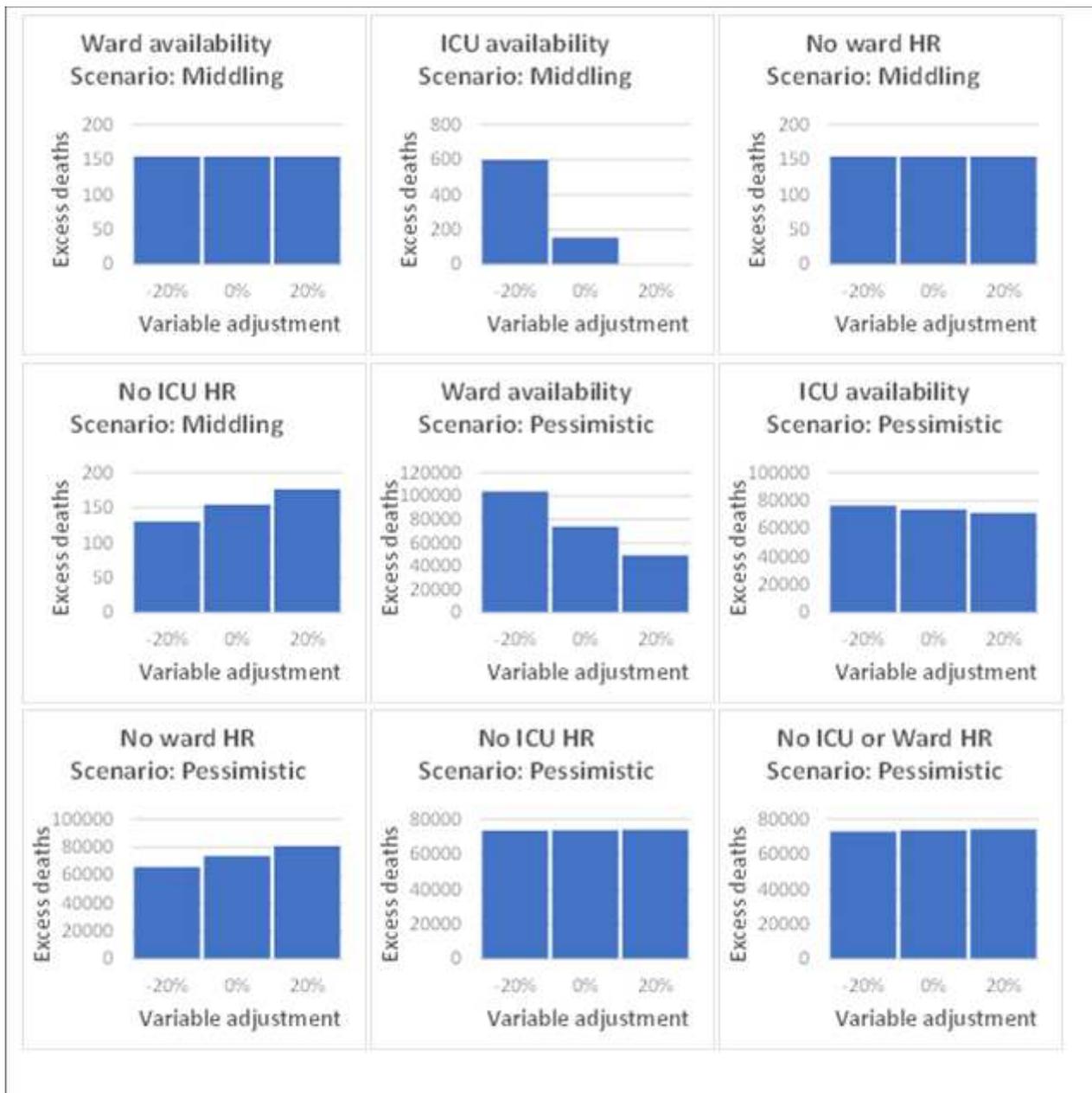


Figure 3

Results of the sensitivity analysis.

Supplementary Files

This is a list of supplementary files associated with this preprint. Click to download.

- [BMCIFOACOV19AdditionalFile2Table2AssumptionsExcessMortality.pptx](#)
- [COVID19ExceedingCapacityModelMedRxivSubmission210223.xlsx](#)