

1 Forecasting sub-national trends in COVID-19 vaccine uptake in the UK

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12 The rollout of COVID-19 vaccines has begun to at-risk populations around the world. It is currently
13 unclear whether rejection of the vaccine will pose challenges for achieving herd/community immunity
14 either through large-scale rejection or localised pockets. Here we predict uptake of the vaccine at
15 unprecedented spatial resolution across the UK using a large-scale survey of over 17,000 individuals.
16 Although the majority of the UK population would likely take the vaccine, there is substantial
17 heterogeneity in uptake intent across the UK. Large urban areas, including London and North West
18 England, females, Black or Black British ethnicities, and Polish-speakers are among the least accepting.
19 This study helps identify areas and socio-demographic groups where vaccination levels may not reach
20 those levels required for herd immunity. Identifying clusters of non-vaccinators is extremely important
21 in the context of achieving herd immunity as vaccination “cold-spots” can amplify epidemic spread and
22 disproportionately increase vaccination levels required for herd protection.

23 24 25 Introduction

26 A vaccine against the novel coronavirus 2019 disease (COVID-19) caused by the severe acute
27 respiratory coronavirus 2 (SARS-CoV-2) will be a major step in reducing mortality, morbidity,
28 economic, and societal burdens associated with the COVID-19 pandemic. The UK’s National Health
29 Service (NHS) has begun the rollout of two vaccines approved by the Medicines and Healthcare products
30 Regulatory Authority (MHRA)¹ and has administered almost 4 million doses (week ending 17 January
31 2021).

32 A successful vaccination campaign of a safe and effective COVID-19 vaccine is contingent on several
33 factors: at-scale manufacture ensuring sufficient dosages to target populations; governments and health
34 organisations ensuring fast and equitable distribution via existing and novel supply-chain networks with
35 sufficient capacity for storage and delivery; and public acceptance. This latter factor is perhaps of
36 particular concern in the UK, which has had notable hesitancy towards vaccinating in the past², and has
37 had widely circulating false stories about a COVID-19 vaccine^{3–5}. Over the past three years, there have
38 been year-on-year decreases in uptake of routine immunisations – such as the MMR vaccine⁶ – with
39 corresponding outbreaks of vaccine-preventable diseases (and a loss of the UK’s measles-free status).^{7–}
40 ¹⁰. UK policymakers may therefore face significant public concern over a novel vaccine.

41 Although recent studies show largely positive attitudes towards a COVID-19 vaccine across the UK^{11,12},
42 no study has – to date – investigated sub-national acceptance and whether specific regions may fail to
43 meet the required vaccination levels for herd immunity (estimated at around 65% for the UK¹³). Vaccine
44 delays and refusals not only place individuals directly at risk but can contribute to lowering vaccination
45 thresholds required for herd immunity. Geographic clustering of non-vaccinators can be particularly
46 troublesome, as these “cold spots” can disproportionately increase required vaccination levels for herd
47 immunity in adjacent settings, as they serve as infection hubs amplifying the spread of disease^{14,15}. It is

48 therefore important to identify the regions – and the socio-demographic groups – at risk of vaccine
49 refusal or delay.

50 In this large-scale modelling study, intent to accept a COVID-19 vaccine is estimated for 174 sub-
51 national regions across the UK using survey data from over 16,820 individuals. Multilevel regression
52 and poststratification (MRP) – a statistical method recently used to successfully predict national general
53 election results in the UK¹⁶ – is used to obtain these sub-national estimates and to identify the socio-
54 demographic barriers of intent to accept a COVID-19 vaccine. Partial validation for this modelling
55 approach is obtained via uptake rates among over 70s across England from the start of vaccine rollout
56 to 18 February 2021.

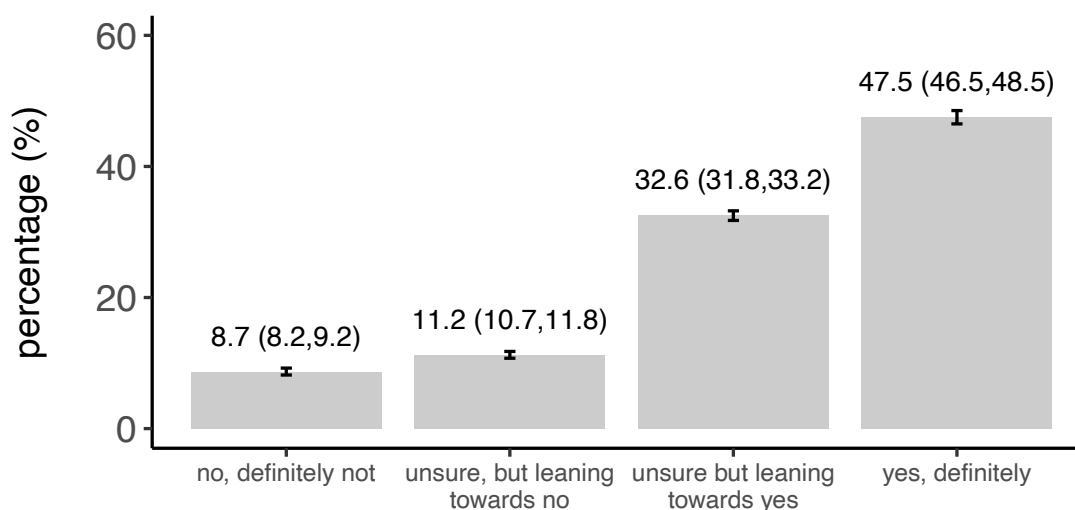
57 This study aims to provide policymakers with estimates for COVID-19 uptake rates among the UK adult
58 population and to establish socio-demographic groups at high risk of vaccine refusal and to highlight
59 regions that may pose challenges for reaching herd immunity across the UK.

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Would you accept a COVID-19 vaccine?

UK-wide MRP estimates, n=16820



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Fig. 1. National-level estimates of COVID-19 vaccine uptake intent. National-level estimates for the percentage of the UK that would: “definitely not” accept a COVID-19 vaccine, “definitely” accept a COVID-19 vaccine, or who are unsure. Uncertainty in estimates are 95% HPD intervals.

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Survey question	Values (recodes, including to align with UK census in parenthesis)	Regression baseline
Response: COVID-19 vaccination intent		
If a new coronavirus (COVID-19) vaccine became available, would you accept the vaccine for yourself?	yes, definitely; unsure, but leaning towards yes; unsure but leaning towards no no, definitely not	n/a (response variable)
Covariates: socio-demographic factors		
sex	male and female	male
age	integer value mapped to 18-24, 25-34, 35-44, 45-54, 55-64, 65-79, 80+	18-24
highest educational attainment	No academic qualifications (none/other) 0-4 GCSE, O-levels, or equivalents (level 1-3) 5+ GCSE, O-levels, 1 A level, or equivalents (level 1-3) 2+ A levels or equivalents (level 1-3) Undergraduate or postgraduate degree or other professional qualification (level 4) Apprenticeship (none/other) Other (e.g., vocational, foreign qualifications) (none/other) Do not know (none/other) Do not wish to answer (none/other)	level 1-3
religious affiliation	atheist/agnostic Christian Buddhist (other religion) Hindu Muslim other religion do not wish to answer (not given)	atheist / agnostic
work status	working full-time (including self-employed) part-time (including self-employed) unemployed student looking after the home retired (retired / disabled) unable to work (e.g., short- or long-term disability) (retired / disabled) do not wish to answer (other work status)	full-time
ethnicity	White: English/Welsh/Scottish/Northern Irish/British (White) White: Irish (White) White: Other white background (White) White and Black Caribbean (mixed) White and Black African (mixed) White and Asian or White and Asian British (mixed) Black, African, Caribbean or Black British (Black/Black British) Asian or Asian British: Indian (Asian/Asian British) Asian or Asian British: Pakistani (Asian/Asian British) Asian or Asian British: Chinese (Asian/Asian British) Asian or Asian British: Other (Asian/Asian British) other ethnicity (other ethnicity) do not wish to answer (other ethnicity)	White
language	English or Welsh Polish Punjabi (other language) Urdu (other language) Bengali (other language) Other (other language) do not wish to answer (other language)	English or Welsh

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78 **Table 1. Study data** Survey items are shown with the responses (including recodes, if any), and baselines used in the ordinal
79 logistic regressions. COVID-19 vaccination intent is the study response variable.

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85 **Results**

86 **Data collection**

87 Between 24 September and 14 October 2020, a cross-sectional online survey was administered to 17,684
88 UK residents aged 18 and over. During data collection, quality control procedures resulted in the removal
89 of 864 respondents (see *Methods*). All respondents were recruited via an online panel by ORB (Gallup)
90 International (www.orb-international.com) and informed consent was obtained before respondents
91 participated. The questionnaire is provided in the Supplementary Materials.

92
93 Respondents are asked whether they would accept a COVID-19 vaccine: “*If a new coronavirus (COVID-
94 19) vaccine became available, would you accept the vaccine for yourself?*”. Respondents could provide
95 one of four responses on an ordinal scale: “*yes, definitely*”, “*unsure, but leaning towards yes*”, “*unsure,
96 but leaning towards no*”, or “*no, definitely not*”.

97
98 Socio-demographic data was collected for each respondent to assess the relationship between these
99 characteristics and vaccine intent and to allow for the reweighting of respondents’ vaccination intent
100 according to census data (both via multilevel regression and poststratification, see below). These
101 covariate data were therefore chosen to align with the socio-demographic data collected in the latest UK
102 census. The covariate data collected for each individual was: sex, age, highest educational attainment,
103 religious affiliation, ethnicity, employment status, primary language, and outer postcode. Respondent’s
104 outer postcode was used to map respondents to one of 174 third level NUTS regions (NUTS3).
105 Descriptions for all respondent data collected and recoding are provided in table 1.

106 107 **Multilevel regression and poststratification**

108 Multilevel regression and poststratification (MRP)^{16,17} is used to estimate intent to accept a COVID-19
109 vaccine across the 174 sub-national regions across the UK and to identify the socio-demographic barriers
110 to uptake (see *Methods* for full model details).

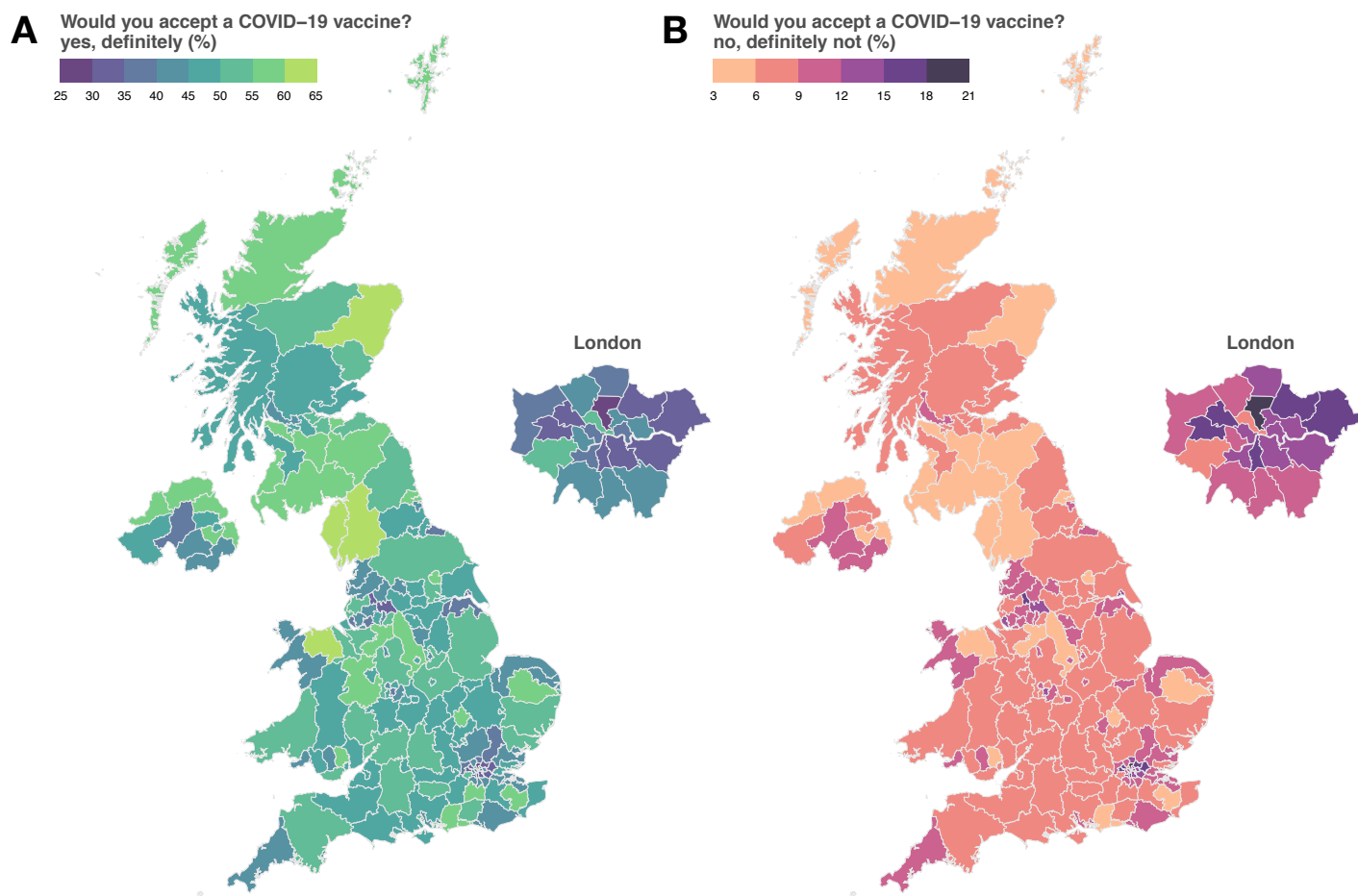
111 112 **COVID-19 vaccination intent**

113 Across the UK, just under half the population – 47.5% (95% highest posterior density interval (HPDI)
114 46.5 to 48.5%) – would “definitely” take a COVID-19 vaccine according to the MRP-based estimates
115 of uptake intent. A further 32.6% (31.8 to 33.2%) are leaning towards vaccinating but are unsure. 8.7%
116 (8.2 to 9.2%) would “definitely not” take a COVID-19 vaccine and 11.2% (10.7 to 11.8%) are unsure
117 but leaning towards no (fig 1).

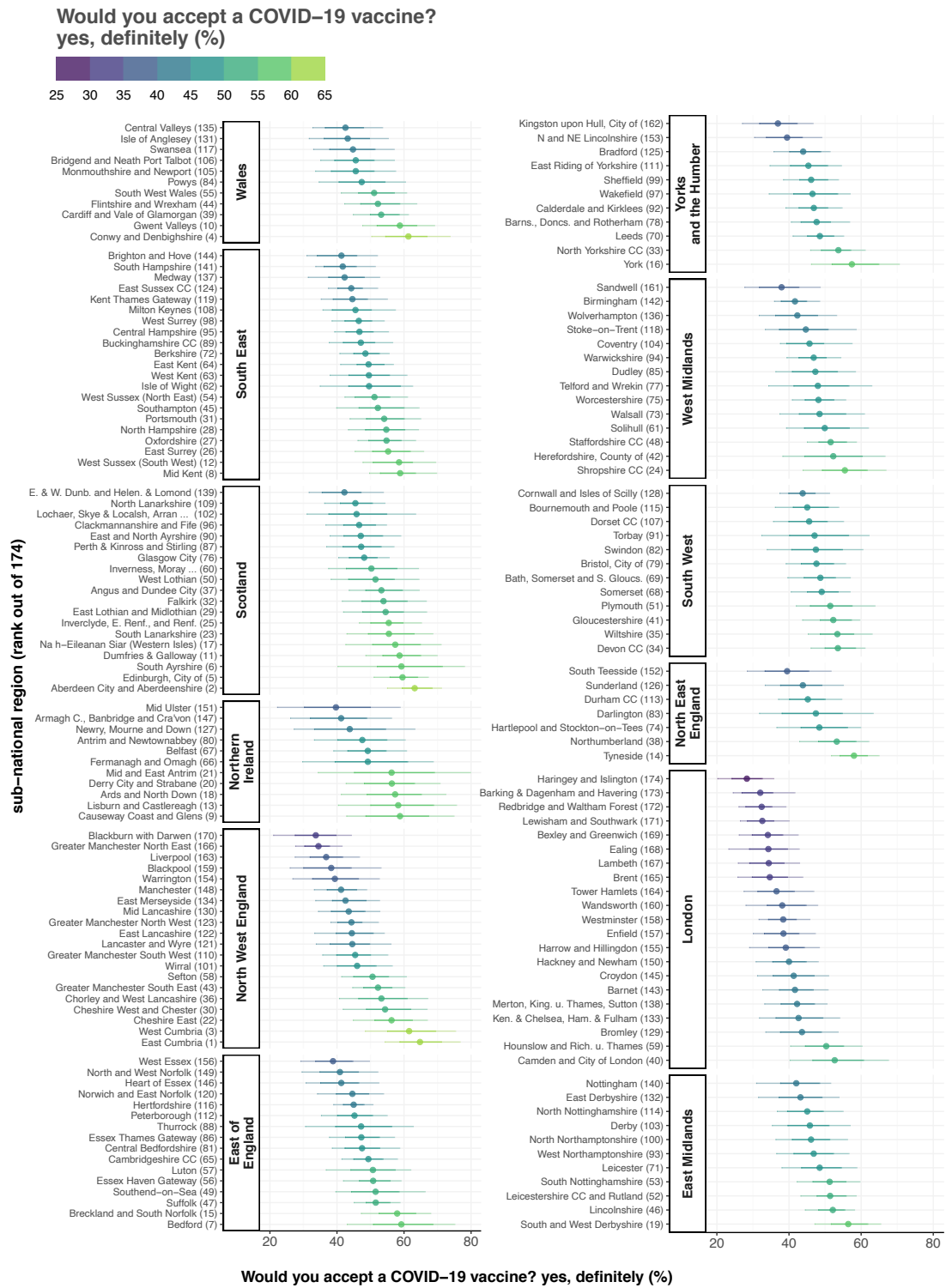
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119 Sub-national MRP estimates of the proportion of each of the UK’s 174 NUTS regions who would
120 “definitely” accept a COVID-19 vaccine are mapped in figure 2A. Estimates of the proportions who
121 would “definitely not” accept a COVID-19 vaccine are mapped in figure 2B. The values in figure 2A
122 are repeated in figure 3 with their corresponding 70% and 95% HPDIs and are ranked from lowest to
123 highest acceptance by broad UK region. (Raw values for the MRP estimates and HPDIs for each of the
124 174 sub-national regions and each of the four outcome variable options are provided in the
125 supplementary data file).

126
127 Estimates across the 174 sub-national NUTS regions of the UK vary considerably. Estimates of the
128 proportion of the public who would “definitely” accept a COVID-19 vaccine (figure 2A) range from
129 28.3% (20.1 to 35.7%) in Haringey and Islington to 64.8% (54.2 to 76.6%) in East Cumbria (fig 2 and
130 fig 3). The lowest proportions of the UK public who would “definitely” accept a COVID-19 vaccine are
131 concentrated in London, which contains 13 of the 20 lowest proportions in the UK: Haringey and
132 Islington (28.3%, 20.1 to 35.7%); Barking, and Dagenham and Havering (32.0%, 24.4 to 41.5%);
133 Redbridge and Waltham Forest (32.4%, 26.0 to 39.1%); Lewisham and Southwark (32.6%, 26.4 to
134 39.9%); Bexley and Greenwich (34.1%, 26.1 to 42.4%); Ealing (34.3%, 23.2% to 42.7%); Lambeth

135 (34.4%, 25.8 to 42.8%); Brent (34.7%, 25.7 to 43.7%); Tower Hamlets (36.5%, 27.4 to 46.8%);
136 Wandsworth (38.1%, 28.0 to 47.9%); Westminster (38.4%, 31.6 to 45.7%); Enfield (38.5%, 30.0 to
137 47.2%); and Harrow and Hillingdon (39.1%, 29.0 to 48.3%). Four of the remaining seven regions in the
138 lowest 20 are in North West England: Blackburn with Darwen (33.7%, 21.0 to 44.2%), Greater
139 Manchester North East (Bury, Oldham, and Rochdale) (34.5%, 27.5 to 41.4%), Liverpool (36.8%, 27.4
140 to 46.6%), Blackpool (38.3%, 25.9 to 53.0%). The remaining three areas in the lowest 20 are West Essex
141 (East of England, 38.8%, 29.1 to 49.6%), Sandwell (West Midlands, 37.9%, 27.6 to 48.6%), and the
142 City of Kingston upon Hull (Yorkshire and the Humber, 36.9%, 27.0 to 46.6%). (See supplementary
143 data file for all estimated values and posterior intervals.)
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149 **Fig. 2. Sub-national estimates of COVID-19 vaccine intent across the UK.** The estimated proportion of respondents in
150 each of the UK's 174 NUTS3 region who would state they would “definitely” accept a COVID-19 (A) and who would
151 “definitely not” accept a COVID-19 vaccine (B). Regional boundaries are used under the Open Government License v3.0
152 (see [https://data.gov.uk/dataset/b147a160-86b6-48e4-8dd0-f35b90981814/nuts-level-3-january-2015-super-generalised-](https://data.gov.uk/dataset/b147a160-86b6-48e4-8dd0-f35b90981814/nuts-level-3-january-2015-super-generalised-clipped-boundaries-in-england-and-wales)
153 [clipped-boundaries-in-england-and-wales](https://data.gov.uk/dataset/b147a160-86b6-48e4-8dd0-f35b90981814/nuts-level-3-january-2015-super-generalised-clipped-boundaries-in-england-and-wales) accessed 25 November 2020).
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Fig. 3. Ranked intent to accept a COVID-19 vaccine. The estimated proportion of respondents in each of the UK’s 174 NUTS3 region who would definitely accept a COVID-19 vaccine are shown and ranked within the 12 first-level NUTS regions (NUTS1). 70% and 95% highest posterior density intervals (horizontal bars) are shown around the mean estimate (dot). Each region is suffixed by its rank (out of 174) according to the estimated proportion who would “definitely” accept a COVID-19 vaccine. East Cumbria (North West England) ranks first, while Haringey and Islington (London) is last.

164 The five regions with the highest proportions of the UK public who would “definitely” accept a COVID-
165 19 vaccine are East Cumbria (64.8%, 54.2 to 76.6%), Aberdeen City and Aberdeenshire (63.2%, 54.9 to
166 71.1%), West Cumbria (61.5%, 48.3 to 75.3%), Conwy and Denbighshire (61.3%, 50.3 to 73.7%), and
167 the City of Edinburgh (59.6%, 51.0 to 67.1%). The top 20 regions disproportionately contain regions in
168 Scotland (5 regions) and Northern Ireland (4) (fig 3 and the supplementary data file). (In fact, Scotland
169 and NI have 13 of the highest-ranking regions in the top 26.)
170

171 The regions with the highest estimated proportions who would “definitely not” accept a COVID-19
172 vaccine are again predominately located in London and the North West. Haringey and Islington (19.0%
173 (12.8 to 25.6%), Blackburn with Darwen (16.6, 9.8 to 25.0), and Redbridge and Waltham Forest (16.3%,
174 11.7 to 21.1%) have the highest estimated proportions who would “definitely not” accept the vaccine,
175 while East Cumbria (3.8%, 2.0 to 5.8%), Aberdeen City and Aberdeenshire (4.1%, 2.8 to 5.9%), and
176 West Cumbria (4.5%, 2.2 to 7.5%) have the lowest. Estimates for the proportions of respondents who
177 are “unsure” about taking a COVID-19 vaccine are provided in the supplementary data file and mapped
178 in the appendix, figure S1.
179

180 **Socio-demographic determinants of vaccination intent**

181 The fixed-effects in the ordinal multilevel regression (see statistical analysis and appendix 2) – which
182 represent an “average” impact of socio-econo-demographic characteristics on vaccination intent across
183 the whole country – are shown in figure 4.
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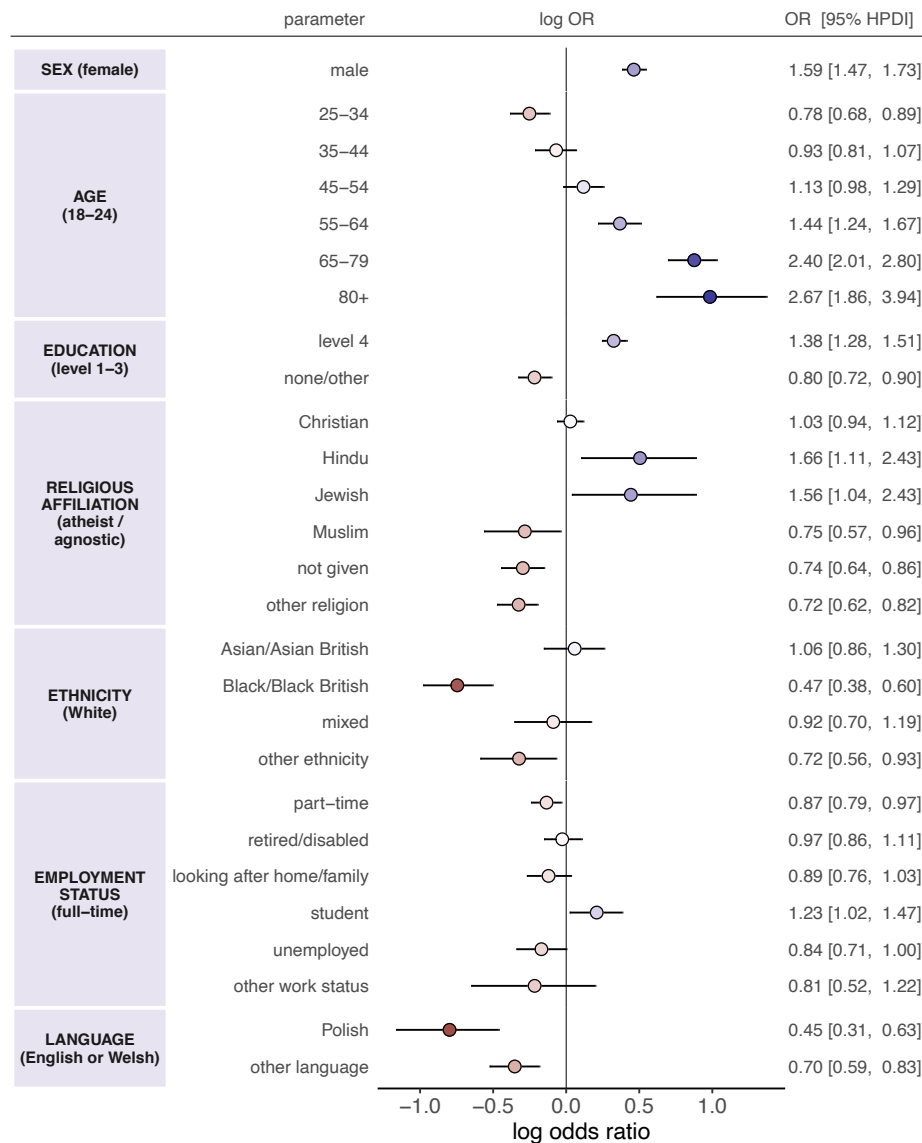
185 A number of factors are associated with COVID-19 vaccine intent. Males are more likely than females
186 (odds ratio 1.59, 95% HPDI 1.47 to 1.73) to accept a COVID-19 vaccine. Older age groups are more
187 likely to accept a COVID-19 vaccine than 18-24-year-olds, in particular 65-79 and 80+ year-olds (2.40,
188 2.01 to 2.80 and 2.67, 1.86 to 3.94, respectively). Interestingly, 25-34-year-olds are *less* likely than 18-
189 24-year-olds to accept a vaccine (0.78, 0.68 to 0.89). Individuals with undergraduate or postgraduate
190 qualifications (level 4) are more likely than those with GCSEs, A- or O-levels to accept a vaccine (1.38,
191 1.28 to 1.51) while those with no formal qualifications or other qualifications (see table 1) are less likely
192 (0.80, 0.72 to 0.90).
193

194 There is no evidence to suggest those who identify as Christian are more or less likely than atheists or
195 agnostics to accept a vaccine (1.03, 0.94 to 1.12), those reporting Hinduism or Judaism as their religion
196 are more likely than atheists or agnostics to accept a vaccine (1.66, 1.11 to 2.43 and 1.56, 1.04 to 2.43,
197 respectively). Those identifying as Muslim (0.75, 0.57 to 0.96), not providing their religion (0.74, 0.64
198 to 0.86), or stating an “other” religious affiliation (0.72, 0.62 to 0.82) are less likely to accept a COVID-
199 19 than atheists or agnostics. Ethnicity also plays a role in determining intent to accept a COVID-19
200 vaccine, independently of religion, with those identifying as Black or Black British (0.47, 0.38 to 0.60)
201 and those reporting an “other” ethnicity than those provided (0.72, 0.56 to 0.93) less likely to accept a
202 COVID-19 vaccine than Whites.
203

204 Individuals’ employment status appears to play less of a role than the other factors outlined above, with
205 odds ratios closer to one. However, there is evidence to suggest that those in part-time work (0.87, 0.79
206 to 0.97) or who are unemployed (0.84, 0.71 to 1.00) are less likely than those in full-time in employment
207 to accept a COVID-19 vaccine, while students (1.23, 1.02 to 1.47) are more likely.
208

209 Individuals who report a language other than English or Welsh as their primary language hold less intent
210 to accept to accept a vaccine than those reporting English or Welsh (Polish 0.45, 0.31 to 0.63 and “other”
211 language 0.70, 0.59 to 0.83).
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socio-econo-demographic determinants of intent to accept a COVID-19 vaccine (fixed-effects)

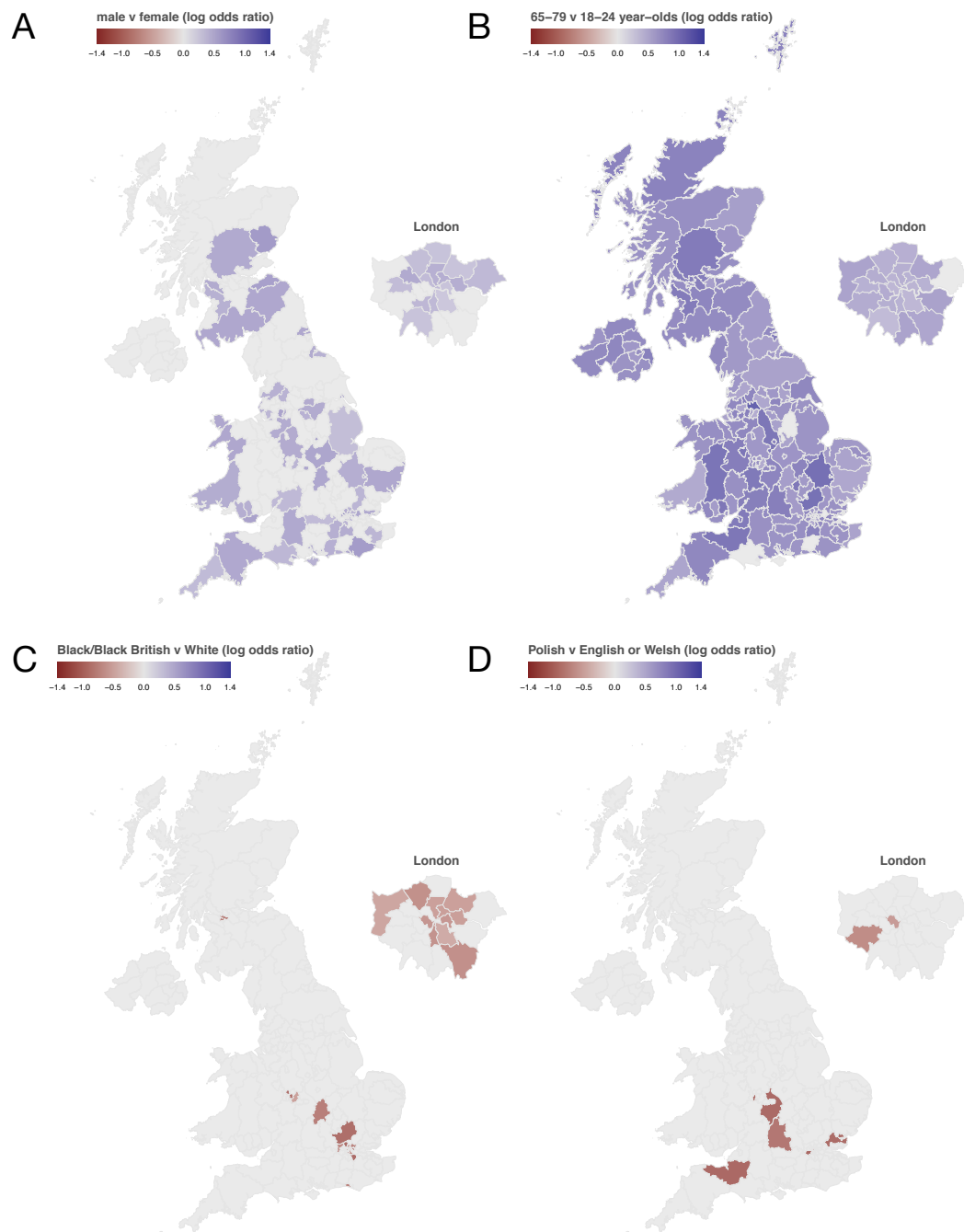


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Fig.4. The socio-demographic determinants of intent to accept a COVID-19 vaccine. Multilevel regression fixed-effect parameter log odds ratios are plotted with corresponding 95% HPDIs. These log odds ratios are coloured by effect magnitude and direction, where blues (reds) signify that the group is more (less) likely than the baseline group to accept a COVID-19 vaccine. The darker the colour the stronger the association. For each factor, the baseline group is provided in parentheses on the left. Odds ratios with 95% HPDIs are shown on the right for each parameter.

221 Variation in socio-demographic determinants of uptake across the UK are shown in figure 4 for the
222 fixed-effect parameters with the strongest overall association with uptake intent. To focus on the
223 strongest associations between socio-demographic factor and uptake, regions are coloured if the 95%
224 HPDI excludes zero and set to zero otherwise. Males are found to be more likely to accept a COVID-19
225 vaccine in 67 UK regions (figure 5A), while 65-79-year-olds are found to be more likely to accept the
226 vaccine than 18-24-year-olds in all but eight UK regions (figure 5B). There is evidence that individuals
227 identifying as Black/Black British are less likely than those identifying as White to accept the vaccine
228 in 17 UK regions mostly concentrated in London (e.g., Hackney and Newham, Tower Hamlets, Haringey
229 and Islington, Lewisham and Southwark), the West Midlands (Birmingham Sandwell, and

230 Wolverhampton), and Brighton and Hove (figure 5C). Polish speakers are found to be less likely than
231 English or Welsh speakers to accept the vaccine in seven UK regions (figure 5D). All region-specific
232 random-effect parameters with corresponding HPD intervals are provided for national and local
233 policymakers in the supplementary data file and shown in SM figure S3.
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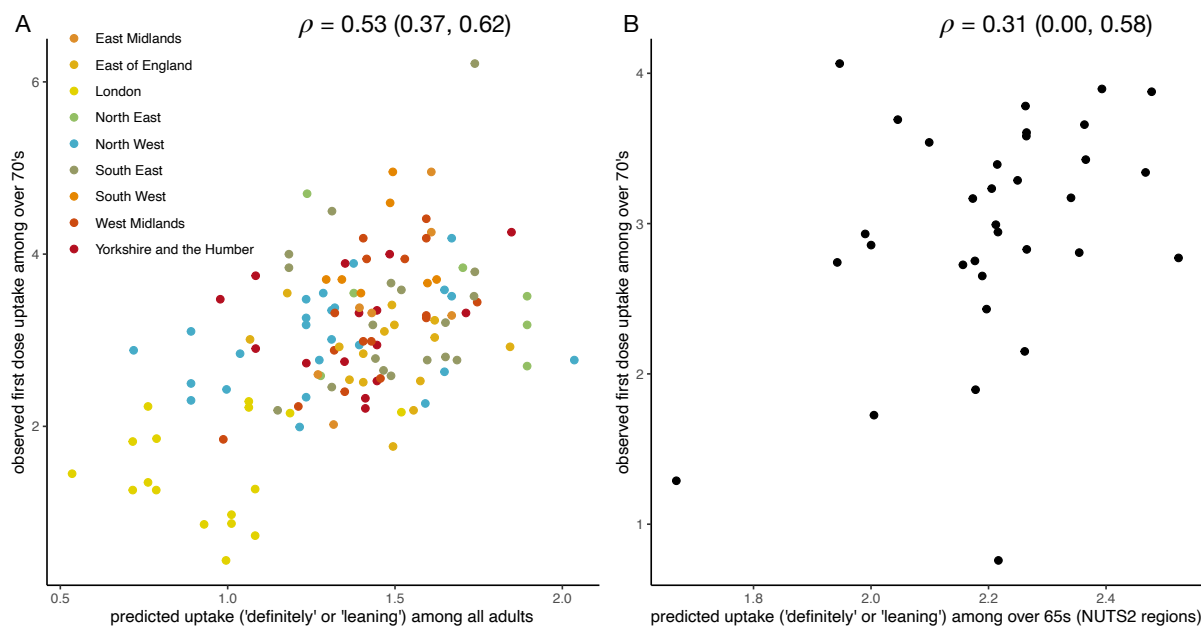


236
237 **Fig.5. Sub-national socio-demographic determinants of intent to accept a COVID-19 vaccine.** Multilevel regression
238 random-effect parameter log odds ratios are shown for the four socio-demographic sub-groups that show the strongest
239 differential association with uptake. Log odds ratios are shown for each sub-national region for A) sex (males versus females),
240 B) age (65-79 year-olds versus 18-24 year-olds), C) ethnicity (Black/Black British v White), D) language (Polish v English
241 or Welsh). Blues denote that the group has a positive association with intent to accept a COVID-19 vaccine with respect to
242 the baseline group, while holding all other covariates constant. For example, in all cases when the 95% HPDI intervals around
243 the male random-effect parameters exclude zero, males are more likely than females to state they would accept the vaccine.
244 Parameters whose 95% HPDIs exclude zero (all other parameters are set to zero for the purpose of visualisation)
245

246 Model validation against recent UK uptake data

247 Vaccination rollout began in the UK on 8 December 2020. Data are now available on the percentage of
248 first doses administered in each of England's 136 Clinical Commissioning groups (CCG) for over 70s
249 (who have now been offered a vaccine) as well as the estimated over 70 population in each of these
250 regions¹⁸. To obtain validation for the modelling approach in this study, two checks are performed.
251 Firstly, all regional forecasts across England are correlated with observed uptake among over 70s and
252 secondly, multilevel regression and poststratification is reimplemented for all individuals aged over 65
253 (census microdata records bin respondents by age, and so it was necessary to include some respondents
254 less than 70 so as to not remove respondents aged 70-74 from the analysis) at the second NUTS level.
255 As there are 3,338 individuals aged over 65 collected in the survey, MRP estimates are generated at the
256 40 second level NUTS units across the UK (of which 33 units are in England). Occasionally, regional
257 boundaries do not precisely align between CCG and NUTS units and, when this is the case, population-
258 weighted averages are taken to provide a reasonable mapping between acceptance forecasts in the NUTS
259 units and the CCGs. First dose data are used, since the UK vaccination policy involves a four- to 12-
260 week delay between of either the Pfizer-BioNTech or Oxford-AstraZeneca vaccine in the majority of
261 cases, and will therefore be more representative of uptake compared to a completed two-dose schedule¹⁹.

262
263 Predicted vaccine acceptance (the percentage of respondents stating they would "definitely" vaccinate
264 or who are "unsure, but leaning towards yes") across all England regions and for all adults correlates
265 with observed uptake among over 70s ($\rho=0.53$ (0.37, 0.62), figure 6A). Predicted uptake is lower than
266 observed uptake, because COVID-19 vaccines have not been made available to the vast majority of
267 younger age groups. Predicted acceptance among all over 65s surveyed also correlates with observed
268 uptake among over 70s, though this correlation is weaker ($\rho=0.31$ (0.00, 0.58), figure 6B). Predicted
269 uptake is again lower than observed, which is possibly due to 65–69-year-olds being included in the
270 analysis or because of changes in the UK's vaccine acceptance since survey data was collected in late
271 2020.
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275 **Fig. 6. Forecasts of vaccine acceptance correlates at sub-national levels correlates with observed first-dose COVID-19**
276 **vaccine uptake among over 70s.** (A) Observed vaccination coverage across all England Clinical Commissioning Groups
277 correlates with predicted uptake across the UK adult population. (B) Observed vaccination coverage across 33 NUTS2 levels
278 across England correlates with predicted uptake among over 65s surveyed in this study. Coverage values on both axes have
279 been scaled using the inverse logit transform.

280 Discussion

281 This study reports multiple findings of immediate relevance to clinicians and policymakers involved
282 with the delivery of a COVID-19 vaccine. This study estimates that less than half the UK public would
283 “definitely” accept a COVID-19 vaccine, with strong regional variation in estimates. Although a
284 relatively small proportion (8.7%, 8.2 to 9.2%) state that they would “definitely, not” accept a vaccine,
285 rates of rejection intent are much higher in London and the North West, where they reach as high as
286 18.0% (14.8 to 20.7%) in Haringey and Islington. Since February 2020, London and the North West
287 have experienced high disease burdens. The North West is particularly notable in this regard as four of
288 the five UK regions with the highest infection rate (correct of 20 November 2020) are all in the North
289 West: Blackburn and Darwen (6,312 per 100,000), Oldham (6,157), Rochdale (5,585), and Manchester
290 (5,539)²⁰. Interestingly, Blackburn and Darwen has the fifth lowest intent to accept a COVID-19 vaccine
291 (ranking 170 out of 174), while Oldham and Rochdale rank ninth lowest (Greater Manchester North East
292 – which contains both these towns – ranks 166 out of 174). Manchester fares a little better ranking 148
293 out of 174. These results point to an important possible interaction between high COVID-19 rates and
294 low vaccine acceptance and the effect this may have on vaccination rates required for herd/community
295 immunity in adjacent regions^{13–15}. Significant correlations are observed between sub-national forecasts
296 and first-dose vaccination uptake among over 70s, validating the modelling approach.
297

298 Socio-demographic background is strongly associated with intent to accept the vaccine. This study finds
299 strong evidence to suggest that males, and older age groups are substantially more likely to accept a
300 COVID-19 vaccine than females and 18-24-year-olds (respectively). Highest level of education,
301 ethnicity, religious affiliation, and primary language are also found to be strongly related to intent to
302 accept a COVID-19 vaccine. Most notably, individuals who identify as Black or Black British are much
303 less likely than Whites to intend to receive the vaccine, as too are those reporting Polish as their primary
304 language. These associations have been found with regards to existing immunisation programmes^{21–24},
305 as well as – more recently – with respect to vaccine acceptance of a COVID-19 vaccine specifically. A
306 study of over 30,000 adults in the UK conducted between 7 September to 5 October, found similar rates
307 of intent to reject a vaccine (14% of respondents unwilling to receive a vaccine compared to 8.7% --
308 comparison of intent to accept a vaccine is difficult due to differences in questionnaire wording and
309 socio-demographic drivers of intent¹². Notably, that females and those with education levels below
310 postgraduate degrees were less likely to accept a COVID-19 vaccine. A link between BAME groups and
311 uptake was not found at a 95% significance testing interval, however¹² (this could be because of the
312 aggregation of BAME groups and/or a different set of predictor variables used to explain variation in
313 uptake intent). Two other recent studies examining COVID-19 vaccine intent in England and Scotland,
314 however, do find that intent to accept a COVID-19 vaccine is modulated by ethnicity, with non-Whites
315 less likely to accept a COVID-19 vaccine^{25,26}. As risk of severe COVID-19 is greater in BAME
316 communities²⁷, achieving high vaccine acceptance may avert further burden within these communities.
317 (The author refers policymakers to the supplementary data file which reveals regions in which there is a
318 strong association between ethnicity and uptake intent.)
319

320 There are a number of study limitations to note. Firstly, this study maps intent to accept a COVID-19
321 vaccine across the entire population and does not assess vaccine acceptance among at-risk groups or
322 healthcare workers, who are likely to be the first groups offered a novel vaccine. Secondly, the most
323 recent census data used for probability reweighting (see *Statistical analysis* and appendix 2) is from
324 2011. Large changes in the demographic structure of the 174 regional populations could, therefore, result
325 in biased estimates of vaccine intent. Finally, the study was conducted online with a sample of panellists
326 who registered to take part in research surveys. While efforts have been made to ensure
327 representativeness via MRP, there may be a bias among respondents who have access to (and can use)
328 mobile phones or computers, through which the questionnaire would be completed.
329

330 While this study provides a comprehensive snapshot of intent to accept a vaccine across the UK in
331 September and October 2020, it predates both the Pfizer announcement that approval is being sought for
332 use in the UK and the peak of the second wave of daily new coronavirus cases. Attitudes may change
333 on short timescales. As the second wave passes, the UK public may have a decreased appreciation for
334 the importance of the vaccine through either a decrease in the perception of the seriousness of disease
335 or a belief that they have already been infected with SARS-CoV-2 (which is associated with willingness
336 to vaccinate²⁴. Fears relating to the safety of the vaccine could also grow due to the relative speed of
337 vaccine development or because vaccinating is now a reality rather than a hypothetical. Online
338 misinformation could also play a role in shaping vaccination beliefs.
339

340 Despite these limitations, this study greatly extends existing research on both COVID-19 vaccine
341 intentions and – more broadly – on the spatial resolution obtained for studies estimating nations’
342 vaccination beliefs or intentions^{28,29}. By virtue of a more granular sub-national modelling approach,
343 estimates are derived at regional scales consistent with those relevant for local policymaking or for
344 improving epidemiological projections of COVID-19 mortality in the UK³⁰. UK policymakers will need
345 to be prepared to address vaccine concerns within the communities and regions identified in this study.
346

347 **Methods**

348 **Data collection**

349
350 Between 24 September and 14 October 2020, a cross-sectional online survey (see Supplementary
351 Materials) probing acceptance of a novel COVID-19 vaccine was administered to 17,684 UK residents
352 aged 18 and over. Informed consent was obtained from all respondents before the survey commenced.
353 During data collection, quality control procedures resulted in the removal of 864 respondents (see
354 *Methods*). The initial sample size was chosen to maximise the number of observations within each of
355 the sub-national regions: this study has approximately 100 observations for each of the 174 sub-national
356 regions, which far exceeds sample sizes used in similar research³¹. Respondent quotas were set according
357 to national demographic distributions for sex, age, and sub-national region (the second level of the
358 Nomenclature of Territorial Units for Statistics, or ‘NUTS2’, see
359 <https://www.ons.gov.uk/methodology/geography/ukgeographies/eurostat> accessed 25 November 2020)
360 and which were re-adjusted based on the removal of respondents through the ongoing quality control
361 checks during data collection. These quotas ensured a geographic spread of respondents across the UK,
362 between the sexes, and across all age groups. All respondents were recruited via an online panel by ORB
363 (Gallup) International (www.orb-international.com) and informed consent was obtained before
364 respondents participated.
365

366 The initial sample size was chosen to maximise the number of observations within each of the sub-
367 national regions: this study has approximately 100 observations for each of the 174 sub-national regions,
368 which far exceeds sample sizes used in similar research³¹. Respondent quotas were set according to
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373 checks during data collection. These quotas ensured a geographic spread of respondents across the UK,
374 between the sexes, and across all age groups. All respondents were recruited via an online panel by ORB
375 (Gallup) International (www.orb-international.com) and informed consent was obtained before
376 respondents participated.
377

378 The response variable is whether a respondent would accept a COVID-19 vaccine: “*If a new coronavirus*
379 *(COVID-19) vaccine became available, would you accept the vaccine for yourself?*”, with responses on

380 a four-point ordinal scale: “yes, definitely”, “unsure, but leaning towards yes”, “unsure, but leaning
381 towards no”, or “no, definitely not”. The rationale behind this choice of responses is to elicit an explicit
382 vaccination intent rather than provide a continuous or Likert scale, from which the intent to vaccinate
383 may be less clear.

384
385 Covariate data are the socio-demographic traits collected for each individual and were chosen to align
386 with the latest UK census: sex, age, highest educational attainment, religious affiliation, ethnicity,
387 employment status, primary language, and outer postcode. Respondent’s outer postcode was used to
388 map respondents to one of 174 third level NUTS regions (NUTS3). The maximum number of surveys
389 conducted in a NUTS3 region is 293 (Hertfordshire) and the minimum is 16 (Mid and East Antrim). The
390 mean number of responses per NUTS3 unit is 96.7 (with standard deviation 52.1) and the median is 85.
391 A breakdown of the number of individuals surveyed by socio-demographic characteristic is found in
392 Supplementary Materials, figure S1 and the survey counts for each NUTS3 region can be found in the
393 supplementary data file.

394 395 **Multilevel regression and poststratification**

396 Multilevel regression and poststratification (MRP) is used to estimate opinions aggregated at sub-
397 national regions from survey data collected at the national level, via partial pooling of information
398 between these national and sub-national scales³². This pooling of information between the two levels is
399 a compromise between estimates derived via a total aggregation of data (to estimate national trends only)
400 and estimates via complete disaggregation (that is, estimating regional trends only). The former suffers
401 from a loss of information at the regional level while the latter suffers from possible low data counts and
402 the loss of statistical power. More pooling of information will occur in regions with low relative numbers
403 of surveyed individuals and less pooling in regions with high relative counts.

404
405 In brief (and relating specifically to this study), the first step of MRP is to conduct a multilevel regression
406 to estimate, for each stratum (that is, a possible combination of individual characteristics) and for each
407 region, the probability of COVID-19 vaccine acceptance. The second step is to reweight (post-stratify)
408 these strata probabilities by the frequency with which a given strata appears in a population. In this study
409 individual-level UK census data is used to perform the reweighting.

410 411 **Part 1: Multilevel regression**

412 Individual intent to accept a COVID-19 vaccine is specified as $y_{ij} \in \{1,2,3,4\}$, where 1 = “no, definitely
413 not”, 2 = “unsure, but leaning towards no”, 3 = “unsure, but leaning towards yes”, and 4 = “yes,
414 definitely” and $1 < 2 < 3 < 4$. Here, $j = 1, \dots, 174$ is one of the $J = 174$ third National Territorial Units
415 for Statistics (NUTS3) regions in the UK, and $i = 1, \dots, n_j$, where n_j is the number of individuals
416 surveyed in region j . $\sum_j n_j = 16,820$ is the total number of respondents in the survey. A breakdown of
417 the number of respondents in each region and a summary of their socio-econo-demographic status is
418 given in the supplementary data file.

419
420 Intent to accept a COVID-19 vaccine is modelled as a multilevel ordinal regression with the
421 proportional odds assumption³³,

$$422 \quad Y_{ij} | \mathbf{p}_{ij}, n_j, \sim \text{Multi}(\mathbf{p}_{ij}, 1)$$
$$423 \quad \log \frac{\Pr(Y_{ij} \leq k | X_{ij})}{\Pr(Y_{ij} > k | X_{ij})}$$
$$424 \quad = \rho_k + \beta_j^{SEX[i]} + \beta_j^{AGE[i]} + \beta_j^{EDU[i]} + \beta_j^{REL[i]} + \beta_j^{ETH[i]} + \beta_j^{EMP[i]} + \beta_j^{LAN[i]}$$

425

426 for k
 427 $= 1, \dots, 3,$
 428

429 where $\beta_j^{SEX[i]}$, $\beta_j^{AGE[i]}$, $\beta_j^{EDU[i]}$, $\beta_j^{REL[i]}$, $\beta_j^{ETH[i]}$, $\beta_j^{EMP[i]}$, and $\beta_j^{LAN[i]}$ are the random-effect varying
 430 intercepts for sex, age, highest education level, religious affiliation, ethnicity, employment status, and
 431 primary language (respectively); ρ_k are probability threshold parameters; $k \in \{1,2,3,4\}$ is the ordinal
 432 response category; $\mathbf{p}_{ij} = [\Pr(Y_{ij} = 1), \Pr(Y_{ij} = 2), \Pr(Y_{ij} = 3), \Pr(Y_{ij} = 4)]$; and X_{ij} is the
 433 covariate data for individual i in region j . The baseline group for the regression corresponds to an
 434 individual who is male, aged 18-24, has an education level 1-3, is an atheist or agnostic, is White, works
 435 full-time, and speaks English or Welsh as their primary language.

437 In line with prior recommendations for variance components in hierarchical models^{32,34}, default weakly
 438 informative priors are chosen for the random-effects regression coefficients β^1 ,

$$\begin{aligned} 440 \quad & \beta_j^c | \sigma^c \sim N(\gamma^c, (\sigma^c)^2), & \text{for } c = 1, \dots, 27 \\ 441 \quad & 1/(\sigma^c)^2 = \tau^c \sim N_+(0,1), & \text{for } c = 1, \dots, 27 \\ 442 \quad & \beta^c \sim N(0,1), & \text{for } c = 1, \dots, 27, \end{aligned}$$

444 where c indexes the regression coefficients: excluding the threshold parameters, there are 27 fixed-
 445 effect parameters: $\gamma_j^{SEX=female}$, $\gamma_j^{AGE=25-34}$, $\gamma_j^{AGE=35-44}$, $\gamma_j^{AGE=45-54}$, $\gamma_j^{AGE=55-64}$, $\gamma_j^{AGE=65-79}$,
 446 $\gamma_j^{AGE=80+}$, $\gamma_j^{EDU=level 4}$, $\gamma_j^{EDU=none}$, $\gamma_j^{REL=Christian}$,
 447 $\gamma_j^{REL=Hindu}$, $\gamma_j^{REL=Jewish}$, $\gamma_j^{REL=Muslim}$, $\gamma_j^{REL=Not given}$, $\gamma_j^{REL=Other}$, $\gamma_j^{ETH=Asian/Asian British}$,
 448 $\gamma_j^{ETH=Black/Black British}$, $\gamma_j^{ETH=Other}$, $\gamma_j^{EMP=part-time}$, $\gamma_j^{EMP=retired/disabled}$, $\gamma_j^{EMP=student}$, $\gamma_j^{EMP=other}$,
 449 $\gamma_j^{EMP=looking after home}$, $\gamma_j^{EMP=unemployed}$, $\gamma_j^{LAN=Polish}$, and $\gamma_j^{LAN=other}$. (And, thus, a total of 27×174
 450 $= 4,698$ random effect parameters.)

452 Part 2: Post-stratification

453 There are $S = 30,870$ socio-econo-demographic strata (two sexes \times seven age groups \times three education
 454 levels \times seven affiliations for religion \times five ethnicity groupings \times seven employment statuses \times three
 455 languages). Denoting the posterior probabilities of COVID-19 vaccination intent for each stratum $s =$
 456 $1, \dots, S$ and NUTS3 region $j = 1, \dots, J$ as θ_{sjk} (where, as a reminder, $k \in \{1,2,3,4\}$ denotes the response),
 457 then the MRP estimate for the intent to vaccinate within each of the UK's 174 NUTS3 regions is,

$$459 \quad \Phi_{jk} = \sum_s \frac{N_s \theta_{sjk}}{N_s}.$$

460
 461 In the main text, this quantity Φ_{jk} is computed for $k = 4$ (“yes, [I] definitely [would accept a COVID-
 462 19 vaccine]”). Estimates are computed for those who are “unsure” (“unsure, but leaning towards yes”
 463 and “unsure, but leaning towards no” have been combined) and are shown in figure S1 in the
 464 supplementary materials.

466 Model: Implementation and output

¹ Instead of an noninformative $N_+(0,100)$ distribution over the standard deviation of hierarchical variance parameters³⁴, a weakly-informative $N_+(0,1)$ prior is placed over the *precision* of these parameters, which places 95% of σ^c 's prior mass between 0.54 and 4.05.

467 The multilevel regression model detailed above is implemented using JAGS version 4.3.0 (implemented
468 via `rjags`³⁵) and R version 4.0.3. 25,000 posterior samples (excluding the first 5,000 for model burn-
469 in) was sufficient for successful convergence and all posterior draws were well-mixed. The posterior
470 draws for the fixed effects are shown in figure S4 and all look visibly well-mixed and all except “other
471 work status” ($p = 0.04$) have Geweke p -values above 0.05. There are too many posterior draws to plot
472 for all random-effects, but we show posterior draws for the first UK NUTS3 region alphabetically
473 (Hartlepool and Stockton-on-Tees) in figure S5 with a histogram of Geweke p -values for all model
474 parameters (fixed effects, random effects, and variance components) to demonstrate universally good
475 mixing and convergence in figure S3. In the computation of the Geweke statistic, the first 10% and final
476 50% of the posterior samples used for computation are used. Convergence of variance parameters is
477 shown in figure S6. A slightly larger fraction of Geweke p -values fall below 0.05 than is expected by
478 chance (0.082 compared to 0.05 by chance). Manual inspection of these chains revealed no cause for
479 concern: chains showed no ill-mixing or convergence issues.
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Ethical Approval: Approval for this study was obtained via the Imperial College Research Ethics Committee on 24 July 2020 with reference 20IC6133 and European Union GDPR guidelines were followed throughout.

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Data and materials availability: All data used in this study will be made available at <GitHub URL to be inserted>.