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A national survey assessing public readiness for digital health strategies against COVID-19 within the United Kingdom

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There is concern that digital public health initiatives used in the management of COVID-19 may marginalise certain population groups. There is an overlap between the demographics of groups at risk of digital exclusion (older, lower social grade, low educational attainment and ethnic minorities) and those who are vulnerable to poorer health outcomes from SARS-CoV-2. In this national survey study (n = 2040), we assessed how the UK population; particularly these overlapping groups, reported their preparedness for digital health strategies. We report, with respect to using digital information to make health decisions, that those over 60 are less comfortable (net comfort: 57%) than those between 18 and 39 (net comfort: 78%) and lower social grades are less comfortable (net comfort: 63%) than higher social grades (net comfort: 75%). With respect to a preference for digital over non-digital sources in seeking COVID-19 health information, those over 60 (net preference: 21%) are less inclined than those between 18 and 39 (net preference: 60%) and those of low educational attainment (net preference: 30%) are less inclined than those of high educational attainment (net preference: 52%). Lastly, with respect to distinguishing reliable digital COVID-19 information, lower social grades (net confidence: 55%) are less confident than higher social grades (net confidence: 68%) and those of low educational attainment (net confidence: 51%) are less confident than those of high educational attainment (net confidence: 71%). All reported differences are statistically significant (p < 0.01) following multivariate regression modelling. This study suggests that digital public health approaches to COVID-19 have the potential to marginalise groups who are concurrently at risk of digital exclusion and poor health outcomes from SARS-CoV-2.

As of 23nd December 2020, the SARS-CoV-2 virus has infected over 75.9 million people and has claimed over 1.74 million lives globally¹. Throughout, the World Health Organization has emphasised the importance of strict and prompt compliance with public health strategies as the cornerstone in addressing the COVID-19 pandemic². As such, governments have mandated nationwide and regional measures, including social distancing, quarantining, testing and contact tracing³. However, for these approaches to be effective, all sections of the population need to be included in communication efforts.

UK health bodies have been moving towards a 'digital first' strategy as a means of improving healthcare accessibility. This has led to the integration of digital technologies into various elements of national and regional public health plans. These have been especially focussed around the dissemination of critical health information, disease surveillance and digital contact tracing⁴.

Whilst digital technologies can improve the speed, reach and cost efficiency of many traditional public health measures, there are also well described barriers to their use, which can lead to the digital exclusion of population subsets. These barriers^{5–7} can be broadly categorised as:

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- 1. Access—availability and affordability of internet connection and/or equipment, such as laptops or personal computers, smartphones, tablets or smartwatches.
- 2. Skills-deficits in knowledge or ability to use digital resources.
- 3. Engagement—further factors impeding digital interaction, even in the presence of adequate access and skills (e.g., confidence, motivation or time opportunity).

According to the UK Office for National Statistics (ONS), access has steadily increased, with 96% of households with internet connectivity in 2020. Conversely, the same data suggests there remain significant disparities with respect to the skills to make use of this access^{8–10}. The need for reduced in-person contact during the COVID-19 pandemic has fast-tracked the integration and use of digital services by some sectors of the public. Those who have found themselves unable to utilise such services are at highest risk of digital exclusion. These sections of the population include those who are older, are of a lower social grade, have lower educational attainment, have disabilities and those who do not use English as a first language¹¹.

Worryingly, mortality and excess deaths from COVID-19 have been higher in the UK compared to other European countries¹². Greater susceptibility to COVID-19 in the UK has been associated with increased age, socioeconomic deprivation, comorbidity and ethnicity; predominantly those of Afro-Caribbean and South Asian origin¹³. Strikingly, there is significant overlap between these medically vulnerable groups and the aforementioned populations at the highest risk of digital exclusion. This combination of the direct health impact of COVID-19, and the transition towards a digital-first management strategy, therefore, poses a threat of deepening the digital divide thus impeding access, engagement and the efficacy of health services^{14,15}. Accordingly, the failure to account for groups at risk of digital exclusion will likely compound health and societal inequalities.

To date, research has not investigated whether members of the UK population—particularly members who identify with at-risk socio-demographic groups—are in a position to participate in digital health strategies. Do members of the population possess adequate access to digital devices and harbour sufficient confidence in digitally transmitted information for digital health strategies to be effective? Moreover, which sources of information do members of the population access, to what degree are those sources trusted, and how does the population view the particularly important information source of contact-tracing applications? To answer these questions, we conducted a national survey that asked individuals to report their access to digital devices and their perceptions about digital information relevant to the UK's digital health strategies.

Methods

Survey development. An online survey was co-designed with qualitative experts from YouGov (YouGov PLC, London, UK), a market research company. Existing frameworks were identified through a literature search to provide the foundation to the survey design. The eHealth Literacy Framework¹⁶ was the only relevant validated framework identified which covers access, education and engagement as barriers to digital inclusion. It consists of seven core domains.

Thereafter, the UK public health response to COVID-19 was assessed for features and strategies utilising a digital approach. These included delivery of information around the virus, public health messaging about social distancing and quarantine precautions, symptom tracking and contact tracing. These features were mapped to the eHealth Literacy Framework to devise a set of 17 core questions. (Appendix 1).

These were grouped into five themes in keeping with the study objectives: (1) access to personal digital devices (2) confidence to independently source and use information from digital technologies to answer health related questions, (3) identifying which sources of information are commonly used in gathering COVID-19 specific health information, (4) identifying which sources of information harbour the most trust in gathering COVID-19 specific health information and (5) quantifying public opinion regarding the use of the contact tracing apps.

Sample. A sample of 2040 adults was achieved through YouGov's non-probabilistic sampling method. YouGov employ an active sampling methodology to ensure that there is adequate socio-demographic representation within their respondents¹⁷. The proportions of demographics within the respondent panel are compared against (1) UK census data from 2011, (2) large scale random probability surveys (e.g., Labour Force Survey, The National Readership survey and the British Election Study), (3) results of the 2017 general election and 2016 referendum and (4) ONS population estimates¹⁸. This ensures that the coverage is representative of the population as a whole as opposed to those with internet or telephone access. The attained sample is retrieved from a larger panel of more than 360,000 adults, who are registered and incentivised to participate in surveys¹⁸. The sample is representative of UK adults in terms of gender, age, ethnicity, social grade, education attainment and geographical region of residence.

Data was collected between the dates of 15th June 2020 and 24th June 2020 via an online survey conducted by YouGov. A sample size calculation was not performed due to the absence of appropriate pilot data upon which a reliable power calculation may be based. Participants were identified from the YouGov panel and were sent an e-mail with a survey link. Whilst this mode of dissemination does introduce bias, there are numerous reports to suggest that the views of those with access to the internet are similar from those without¹⁹. Moreover, it has been noted that response rates for telephone polls have been sharply declining in recent years; strikingly below 10% in inner city regions¹⁸.

YouGov do not provide response rates for individual datasets, however, it is noted that their aggregate response rate is typically between 35 and 50%; a figure that varies based upon subject matter, complexity and length of survey. All invited participants are from a panel of over 800,000 adults who have registered to participate in surveys and the responding sample is weighted to the profile of the sample definition in order to provide a representative reporting sample. Of note, a Pew Research Center Report²⁰ states that YouGov 'consistently

outperformed' other vendors of nonprobability surveys with regards to accuracy of population representation. As such, given the study goal of rapidly attaining data during a pandemic period, it was felt that an online dissemination strategy, coupled with careful socio-demographic sampling, would allow for accurate yet pragmatic data collection.

Data analysis. We utilised descriptive statistics to describe the sample by gender, age, ethnicity, social grade, educational attainment and governmental office region respectively. Social grade was categorised using the National Readership Survey (NRS) classification system and dichotomised into 'middle class' (ABC1) and 'working class' (C2DE) groups²¹. Education was classified as 'low' (GCSE attainment or below), 'medium' (A-level or equivalent attainment) and 'high' (university degree attainment and above). Respondent ages were grouped into young adults (18–39 years), middle-aged (40–59 years) and elderly (60+ years). Ethnicity is classed as either Caucasian or Black, Asian and minority ethnic (BAME). Government Office regions were aggregated to Southern England (London, South East and South West), Midlands (East of England, East Midlands and West Midlands), Northern England (Yorkshire and the Humber, North East and North West) and Devolved Nations (Scotland, Wales and Northern Ireland).

Outcome. For questions with Likert-type ordinal responses, ordinal logistic regression was performed to examine the relationships between responses and the panel of demographic characteristics described above. Binary logistic regression was used for questions with binary responses. Brant tests were performed to assess the proportional odds assumption for each ordinal logistic regression model using the Stata *omodel* and *brant* commands.

In order to identify discrete response types within survey domains, K-means clustering was applied to all Likert-type ordinal response variables in each domain. Data were normalised by min-max transformation and optimal clusters sizes were determined by relative maxima in silhouette and Calinski Harabasz scores and relative minima in Davies–Bouldin scores²²⁻²⁴. The responses of each cluster and their demographic characteristics were described. All analyses were undertaken on Stata/SE 16.0 (Stata Corporation LP, College Station, Texas, United States of America). K-means clustering was performed using Python v.3.6.8 with the scikit-learn library (version 0.23.1).

Ethical approval. This study was waived by our University Research Office (Ruth Nicholson (Head of Research Governance and Integrity)), in accordance with UK HRA guidelines, as this study is a non-clinical population survey audit of public respondents (involving neither identifiable information, patients nor vulnerable individuals) that constitutes an observation of usual practice. Informed consent was attained from all participants of the survey by YouGov as part of their survey process. YouGov provided the datasets to The Institute of Global Health Innovation and the data is publicly available upon request. Patients and members of the public were not involved in the design, reporting or conduct of the study.

Results

A sample of 2040 adults (Table 1) was achieved. Figure 1 is a significance map which details the directionality and the level of significance associated with responses and the panel of pre-specified demographic characteristics. The results from the logistic regression analyses are detailed in Table 2.

Access. 99% (2024/2040) of the sample cohort have access to a personal digital device (Question 1). Smartphones and laptops/personal computers have the highest penetrance at 88% (1788/2040) and 84% (1719/2040) across the cohort respectively. 61% (1239/2040) of the cohort own tablet computers. Smartwatches (211/2040, 10%) and wearable fitness trackers (391/2040, 19%) were less frequently owned by respondents.

With respect to age, access to personal computers/laptops is stable through to the 60+ age group (651/746 (87%) in 18–39 age group compared to 522/615 (85%) in the 60+ age group). In contrast, smartphone ownership declines in the 60+ age group (702/746 (94%) in the 18–39 age group compared to 465/615 (76%) in the 60+ age group). Ownership of laptops/personal computers decline with lower social grade (508/571 (89%) in AB compared to 337/449 (75%) in DE). Smartphone ownership declines with lower educational attainment groups (587/634 (93%) in the high educational attainment group compared to 434/535 (81%) in the low educational attainment group).

836/2024 (41%) of respondents state that they have used their personal digital device to access COVID-19 specific information (Question 1.1). This figure decreases with age (372/740 (50%) between ages 18 and 39 compared to 182/609 (30%) in those aged above 60), social grades (274/568 (48%) in AB compared to 145/442 (33%) in DE) and educational attainments cohorts (329/632 (52%) in the high educational attainment group compared to 160/529 (30%) in the low educational attainment group). Of all personal digital device activities, instant messaging (1652/2024 (82%)) was the most commonly utilised function, followed by accessing the news (1476/2024 (73%)), telephone calls (1461/2024 (72%)) and then social networking (1447/2024 (71%)).

Confidence. 1423/2040 (70%) are confident at using online or app-based information to make personal health decisions (Question 2). In comparison to their reference counterparts, respondents who are female, over the age of 60 and of a lower social grade are all significantly less confident in using online or app-based information to make personal health decisions (p < 0.01) (Question 2). Those above the age of 60 are consistently significantly less confident in both sourcing and using health resources to form personal health decisions regardless of digital source (internet, apps or social media (Questions 5, 6 and 7) (p < 0.01) and would rather consult a clini-

Total	Number (n = 2040)	Percentage within YouGov sample (%)
Gender		
Male	990	49
Female	1050	51
Age		
18-29	377	18
30-39	369	18
40-49	347	17
50-59	284	15
60–69	356	18
70+	259	14
Social grade		
AB	571	28
C1	592	29
C2	428	21
DE	449	22
Educational attainment		
Low	535	26
Medium	871	43
High	634	31
Region		
North East	72	4
North West	225	11
Yorkshire and the Humber	178	9
East Midlands	162	8
West Midlands	164	8
East of England	168	8
London	268	13
South East	285	14
South West	191	9
Wales	98	5
Scotland	172	8
Northern Ireland	57	3
Ethnicity		
White	1754	86
BAME	286	14

Table 1. Survey respondent demographics table.

cian over the phone than an online or app-based telemedicine service (p < 0.01) (Question 3). Those from lower social grades and of lower educational attainment are significantly less confident at knowing where (Question 6.1) and how (Question 5.1) to use the internet to answer health questions (p < 0.01). There are no significant consistent findings with respect to either ethnicity or region for this domain of questions.

Four distinct clusters of responses for this domain of questions (Questions 3, 5 and 6) were identified. Panel A of Fig. 2 shows the responses of each cluster to each of the constituent questions on which clustering is performed. Clusters were characterised post-hoc based on their responses as 'Digitally confident and preferring online primary care' (19%), 'Digitally confident and preferring telephone primary care' (34%), 'Digitally cautious and preferring telephone primary care' (23%).

Sources of information. Respondents over the age of 40, from lower social grades and of lower educational attainment use online or app-based resources less often than their reference counterparts (p < 0.01) (Question 7). 675/2040 (34%) have not used online resources or apps to seek any COVID-19 information at all (Question 7). Over three times as many people over the age of 60 (124/259 (42%) compared to 95/746 (13%)) in the 18–39 age group would rather access health information from traditional (non-digital) media sources than relying upon digital media sources (Question 10). Those above the age of 60 are more likely to turn towards tabloid newspapers, broadsheet newspapers radio and television than their references counterparts (p < 0.01) whilst avoiding social media (p < 0.01). Those of lower social grades and educational attainment are less likely to use broadsheet newspaper sources (paper or online format) (p < 0.01) (Questions 8 and 9). Respondents of BAME background are also more likely to engage in many digital (non-NHS websites, tabloid newspaper website, broadsheet web-

		Key: Significantly more confident / lil Significantly less confident / lik	tely to use / trusting / willing to share ely to use / trusting / willing to share	$p \le 0.01$ $p \le 0.05$ $p \le 0.05$ $p \le 0.01$	Gender (ref = male)	Age (ref = 18-39)	Ethnicity (ref = white)	Social Group (ref = ABC1	Education (ref = High)	Region (ref = South)	
		Not	significant	p > 0.05	Female	40-59 60+	BAME	C2DE	Low Medium	North Midlands	Scot, Wal, NI
dence	q2 q3 q5_1 q5_2 q5_3	How comfortable are you in using o Would you rather have consult a dor I know how to use the internet to ar I know how to use apps to answer I know how to use social media to a	nline or app-based information to make stor via an app or over the phone? swer questions about my health questions about my health nswer questions about my health	personal health decisions?							
Confi	q6_1 q6_2 q6_3 q7	I know where helpful health resource I know where helpful health resource I know where helpful health resource Know where helpful health resource How often would you use online or	app-based health resources?								
of Information	q8_1 q8_2 q8_3 q8_4 q8_5 q8_6	How often, if at all, would you use the following digital sources to access COVID-19 updates?	NHS Website Other, non-NHS, healthcare websites Tabloid news websites Broadsheet news websites BBC news website Social media								
Sources	q9_1 q9_2 q9_3 q9_4	How often, if at all, would you use the following traditional sources to access COVID-19 updates? Which if either of the following sou	Television Radio Print tabloid newspapers/magazines Print broadsheet newspapers rees do you prefer to use to gather info	rmation on COVID-19?							
Trust	q11_1 q11_2 q11_3 q11_4 q11_5 q11_6	To what extent, if at all, do you trust the COVID-19 information you receive from:	NHS Website Other, non-NHS, healthcare websites Tabloid news websites Broadsheet news websites BBC news website Social media								
ø	q12 q13_1 q13_2 q13_6 q14 q15	How confident are you in telling apa If you saw information on COVID- 19, which of the following would contribute towards your trust in it? How often do you double check on How likely are you to engage with di	t reliable COVID-19 information online That it comes from the Government That it comes from scientists/scientific The source it comes from ine or app-based health information that intal resources if they were directly linke	or through apps? institutions at you receive? d to the controlling the pandemic	7]
Contact Tracing App	q16_1 q16_2 q16_3 <u>q16_4</u> q17_1 q17_2 q17_3 q17_4	How comfortable are you in sharing the following personal data with a Government COVID-19 contact tracing app? How comfortable are you in sharing the following personal data with an industry-led COVID-19 contact tracing ap2	NHS number Age Location Medical history NHS number Age Location Medical bistory								

Figure 1. A significance map detailing directionality and significance of relationships between responses and the panel of demographic characteristics.

site, social media) and traditional information sources (print tabloid and broadsheet newspapers) (p < 0.01) than reference counterparts (Questions 8 and 9).

Five distinct clusters of responses for this domain of questions (Question 9) were identified. Panel B of Fig. 2 shows the responses of each cluster to each of the constituent questions on which clustering is performed. Clusters were characterised post-hoc based on their source of information preference; 'TV, radio and broadsheets' (12.3%), 'TV and radio' (25.7%), TV and tabloids' (14.8%), 'TV only' (26.4%) and 'No traditional media' (20.7%).

Trust. 885/2040 (43%) cited 'trust in the information found' as the main barrier against the use of online/appbased information to guide personal health decisions, ahead of 'knowing where to find information' (406/2040 (20%)) and 'knowing how to action the information found' (379/2040 (19%)) (Question 4). Those above the age of 60 (p < 0.05), from lower social grades (p < 0.01) and of lower educational attainment (p < 0.01) are less confident in telling apart reliable COVID-19 information from unreliable information when encountered online or through apps (Question 12).

Amongst information sources, the NHS website has the highest trust rating (1661/2040 (81%)) whereas social media (1325/2040 (65%)) and tabloid newspapers (1303/2040 (64%)) has the highest distrust rating (Question 11). However, the NHS website is not as preferred by those in lower social grades (p < 0.01), those of low educational attainment (p < 0.05), those above 60 (p < 0.05) and those of BAME backgrounds (p < 0.05). In addition,

	How comfortable are you in using online or app-based information to make personal health decisions?									
q2		Coeff	SE	Z score	p value	95% Conf Interval	idence			
Condor	Male	Reference								
Gender	Female	0.2380	0.0850	2.80	0.005	0.0714	0.4046			
	18-39	Reference								
Age group	40-59	0.1716	0.1075	1.60	0.111	- 0.0392	0.3824			
	60 +	0.9757	0.1121	8.70	0.000	0.7560	1.1954			
Ethnicity	White	Reference								
Etimenty	BAME	0.0670	0.1268	0.53	0.597	-0.1814	0.3155			
Social group	ABC1	Reference								
Social group	C2DE	0.5096	0.0933	5.46	0.000	0.3268	0.6924			
	Low	0.1148	0.1240	0.93	0.355	-0.1283	0.3579			
Education	Medium	0.0866 0.0978 0.89 0.376 -0.1051 0.2782								
	High	Reference								
	South	Reference								
Region	North	0.0370	0.1113	0.33	0.739	-0.1811	0.2552			
Region	Midlands	0.0612	0.1085	0.56	0.573	-0.1514	0.2738			
	Scot, Wal, NI	-0.0494	0.1307	-0.38	0.706	- 0.3055	0.2067			
	Would you rat	her have con	nsult a do	ctor via an	app or ov	er the phon	e?			
q3		Coeff	SE	Z score	p value	95% Conf interval	idence			
Condor	Male	Reference								
Gender	Female	0.2540	0.1087	2.34	0.019	0.0410	0.4671			
	18-39	Reference								
Age group	40-59	0.2679	0.1265	2.12	0.034	0.0199	0.5159			
	60 +	1.2091	0.1520	7.96	0.000	0.9113	1.5070			
Fthnicity	White	Reference								
Etimetty	BAME	-0.0726	0.1490	-0.49	0.626	-0.3647	0.2194			
Social group	ABC1	Reference								
social group	C2DE	0.1437	0.1218	1.18	0.238	-0.0951	0.3825			
	Low	0.4189	0.1689	2.48	0.013	0.0879	0.7498			
Education	Medium	0.0857	0.1208	0.71	0.478	-0.1511	0.3226			
	High	Reference								
	South	Reference			1					
Region	North	0.0759	0.1442	0.53	0.599	-0.2067	0.3585			
	Midlands	-0.0185	0.1385	-0.13	0.894	-0.2901	0.2530			
	Scot, Wal, NI	0.0551	0.1714	0.32	0.748	- 0.2809	0.3910			
	I know how to	use the inte	rnet to ar	nswer ques	tions abou	t my health	L .			
q5_1		Coeff	SE	Z score	p value	95% Conf interval	idence			
Gender	Male	Reference								
	Female	-0.1000	0.0867	-1.15	0.249	-0.2700	0.0700			
	18-39	Reference								
Age Group	40-59	0.0396	0.1096	0.36	0.718	-0.1752	0.2544			
	60+	0.3839	0.1132	3.39	0.001	0.1621	0.6057			
Ethnicity	White	Reference								
	BAME	0.0752	0.1307	0.58	0.565	-0.1809	0.3313			
Social Group	ABC1	Reference								
-	C2DE	0.2718	0.0954	2.85	0.004	0.0848	0.4589			
	Low	0.4670	0.1281	3.65	0.000	0.2160	0.7180			
Education	Medium	0.2097	0.0995	2.11	0.035	0.0147	0.4047			
	High	Reference								
	South	Reference	0.4453		0.00-	0.0	0.055			
Region	North	0.1365	0.1131	1.21	0.228	-0.0852	0.3582			
	Midlands	-0.0413	0.1108	-0.37	0.709	-0.2584	0.1757			
	Scot, Wal, NI	-0.0484	0.1355	-0.36	0.721	- 0.3141	0.2172			

	I know how to	use apps to	answer q	uestions al	bout my he	alth				
						95% Conf	ìdence			
q5_2		Coeff	SE	Z score	<i>p</i> value	interval				
Gender	Male	Reference								
	Female	0.0413	0.0825	0.50	0.616	-0.1203	0.2030			
	18-39	Reference		1	1					
Age group	40-59	0.0393	0.1041	0.38	0.706	-0.1648	0.2434			
	60+	0.7316	0.1089	6.72	0.000	0.5181	0.9450			
Ethnicity	White	Reference								
	BAME	-0.2210	0.1227	-1.80	0.072	-0.4616	0.0195			
Social group	ABC1	Reference								
	C2DE	0.1257	0.0908	1.38	0.166	- 0.0523	0.3037			
	Low	-0.1335	0.1218	-1.10	0.273	-0.3723	0.1052			
Education	Medium	-0.1332	0.0951	-1.40	0.161	-0.3197	0.0532			
	High	Reference								
	South	Reference								
Region	North	0.0389	0.1079	0.36	0.719	-0.1726	0.2504			
nugion	Midlands	-0.0303	0.1053	-0.29	0.773	-0.2367	0.1760			
	Scot, Wal, NI	0.0197	0.1301	0.15	0.880	-0.2354	0.2748			
	I know how to	use social n	nedia to a	nswer que	stions abo	ut my healt	h			
q5_3		Coeff	SE	Z score	<i>p</i> value	95% Conf interval	ìdence			
Gender	Male	Reference								
	Female	-0.0831	0.0819	-1.01	0.311	-0.2436	0.0775			
	18-39	Reference								
Age Group	40-59	-0.1349	0.1030	-1.31	0.190	-0.3367	0.0669			
	60+	0.5348	0.1075	4.98	0.000	0.3241	0.7455			
Fthnicity	White	Reference								
Etimetty	BAME	-0.2885	-0.5275	-0.0495						
Social Group	ABC1	Reference								
Social Group	C2DE	0.0310	0.0900	0.34	0.731	-0.1454	0.2074			
	Low	-0.2758	0.1212	-2.28	0.023	-0.5134	-0.0382			
Education	Medium	-0.2245	0.0942	-2.38	0.017	-0.4090	-0.0399			
	High	Reference								
	South	Reference								
Region	North	- 0.0997	0.1072	- 0.93	0.352	-0.3098	0.1103			
Region	Midlands	-0.1976	0.1049	-1.88	0.060	-0.4032	0.0080			
	Scot, Wal, NI	-0.2346	0.1282	-1.83	0.067	-0.4860	0.0167			
	I know where h	elpful heal	th resour	ces are ava	ilable on tl	ne internet				
q6_1		Coeff	SE	Z score	<i>p</i> value	95% Conf interval	ìdence			
Gender	Male	Reference								
	Female	-0.2090	0.0860	-2.43	0.015	-0.3776	-0.0403			
	18-39	Reference								
Age group	40-59	0.3339	0.1091	3.06	0.002	0.1201	0.5477			
	60+	0.7773	0.1132	6.87	0.000	0.5554	0.9992			
Ethnicity	White	Reference								
Luniony	BAME	0.0746	0.1282	0.58	0.560	-0.1765	0.3258			
Social group	ABC1	Reference				-				
Social group	C2DE	0.2989	0.0950	3.15	0.002	0.1128	0.4850			
	Low	0.5056	0.1270	3.98	0.000	0.2566	0.7545			
Education	Medium	0.2461	0.0988	2.49	0.013	0.0525	0.4397			
	High	Reference								
	South	Reference								
Region	North	0.1140	0.1125	1.01	0.311	-0.1065	0.3345			
ingion	Midlands	0.1105	0.1096	1.01	0.313	-0.1043	0.3254			
	Scot, Wal, NI	-0.0039	0.1347	-0.03	0.977	-0.2679	0.2601			

	I know where h	nelpful heal	th resour	ces are avai	ilable on a	pps	
						95% Conf	ìdence
q6_2		Coeff	SE	Z score	<i>p</i> value	interval	
Gender	Male	Reference					
	Female	0.0335	0.0818	0.41	0.682	-0.1269	0.1939
	18-39	Reference			1		
Age group	40-59	0.1526	0.1040	1.47	0.142	-0.0511	0.3564
	60+	0.8143	0.1089	7.47	0.000	0.6007	1.0278
Ethnicity	White	Reference					
	BAME	-0.2085	0.1220	-1.71	0.087	-0.4477	0.0306
Social group	ABC1	Reference					
	C2DE	0.0864	0.0906	0.95	0.340	- 0.0911	0.2639
	Low	- 0.2939	0.1210	-2.43	0.015	-0.5310	-0.0567
Education	Medium	-0.1733	0.0950	-1.82	0.068	-0.3595	0.0129
	High	Reference					
	South	Reference					
Region	North	0.0173	0.1071	0.16	0.872	-0.1926	0.2273
nugion	Midlands	-0.0213	0.1050	-0.20	0.839	-0.2271	0.1845
	Scot, Wal, NI	0.0459	0.1278	0.36	0.720	-0.2046	0.2964
	I know where h	nelpful heal	th resour	ces are avai	ilable on s	ocial media	L
q6_3		Coeff	SE	Z score	<i>p</i> value	95% Conf interval	ìdence
Gender	Male	Reference			ù		
Gender	Female	-0.1278	0.0819	-1.56	0.119	-0.2883	0.0327
	18-39	Reference		-		-	
Age group	40-59	-0.0388	0.1035	-0.37	0.708	-0.2416	0.1640
	60 +	0.6177	0.1078	5.73	0.000	0.4064	0.8289
Ethnicity	White	Reference					
BAME -0.2338 0.1205 -1.94 0.052 -0.4					-0.4700	0.0023	
Social group	ABC1	Reference					
Social group	C2DE	-0.0543	0.0904	-0.60	0.548	-0.2315	0.1229
	Low	-0.4383	0.1210	- 3.62	0.000	-0.6755	-0.2011
Education	Medium	-0.3138	0.0947	- 3.31	0.001	-0.4993	-0.1282
	High	Reference					
	South	Reference					
Pagion	North	-0.1324	0.1072	-1.24	0.217	-0.3425	0.0777
Region	Midlands	-0.1459	0.1051	- 1.39	0.165	-0.3519	0.0600
	Scot, Wal, NI	-0.1331	0.1271	-1.05	0.295	-0.3822	0.1159
	How often wou	ıld you use	online or	app-based	health res	ources?	
q7		Coeff	SE	Z score	<i>p</i> value	95% Conf interval	ìdence
Gender	Male	Reference					
	Female	-0.1314	0.0843	-1.56	0.119	- 0.2966	0.0339
	18-39	Reference					
Age group	40-59	0.3121	0.1050	2.97	0.003	0.1064	0.5178
	60 +	1.0752	0.1111	9.67	0.000	0.8573	1.2930
Fthnicity	White	Reference					
Etimetty	BAME	-0.2583	0.1249	-2.07	0.039	-0.5032	-0.0134
Social group	ABC1	Reference					
Social group	C2DE	0.4089	0.0932	4.39	0.000	0.2262	0.5915
	Low	0.5550	0.1251	4.43	0.000	0.3097	0.8002
Education	Medium	0.2614	0.0959	2.72	0.006	0.0733	0.4494
	High	Reference					
	South	Reference					
Pagion	North	0.0325	0.1112	0.29	0.770	-0.1855	0.2505
Region	Midlands	0.1058	0.1080	0.98	0.327	-0.1059	0.3176
	Scot, Wal, NI	-0.0170	0.1304	-0.13	0.897	-0.2725	0.2386

	How often, if a sources to acce	all, would you use the following digital ss COVID-19 updates? NHS website							
a8 1		Coeff	SE	Z score	t value	95% Confi interval	dence		
40_1	Male	Reference	0L	Listoit	<i>p</i> vulue	inter vur			
Gender	Female	-0.4827	0.1039	-4.65	0.000	-0.6864	-0.2791		
	18-39	Reference							
Age group	40-59	0.1219	0.1203	1.01	0.311	-0.1138	0.3576		
001	60+	0.1575	0.1354	1.16	0.245	-0.1079	0.4230		
	White	Reference							
Ethnicity	BAME	0.0280	0.1414	0.20	0.843	-0.2492	0.3051		
	ABC1	Reference							
Social group	C2DE	0.1640	0.1169	1.40	0.161	-0.0651	0.3932		
	Low	0.3073	0.1602	1.92	0.055	-0.0067	0.6213		
Education	Medium	0.1465	0.1128	1.30	0.194	-0.0745	0.3675		
	High	Reference							
	South	Reference							
Decien	North	-0.2649	0.1339	- 1.98	0.048	-0.5274	-0.0025		
Region	Midlands	-0.0843	0.1320	-0.64	0.523	-0.3429	0.1743		
	Scot, Wal, NI	-0.1974	0.1584	-1.25	0.212	-0.5078	0.1129		
	How often, if a	t all, would	you use t	he followir	ng digital	Other, nor	1-NHS,		
	sources to acce	ss COVID	19 update	s:		healthcare	websites		
q8_2		Coeff	SE	Z score	p value	95% Confi interval	dence		
<u> </u>	Male	Reference			-				
Gender	Female	-0.1185	0.1040	-1.14	0.255	-0.3224	0.0854		
	18-39	Reference							
Age group	40-59	-0.3477	0.1238	-2.81	0.005	-0.5903	-0.1050		
	60+	0.0882	0.1368	0.65	0.519	-0.1798	0.3563		
Ethnicity	White	Reference							
Etimetty	BAME	-0.6298	0.1497	-4.21	0.000	-0.9232	-0.3364		
Social group	ABC1	Reference							
Social group	C2DE	0.0081	0.1175	0.07	0.945	-0.2222	0.2385		
	Low	0.2563	0.1584	1.62	0.106	-0.0542	0.5668		
Education	Medium	-0.0403	0.1158	-0.35	0.727	-0.2672	0.1865		
	High	Reference							
	South	Reference							
Region	North	0.0566	0.1364	0.41	0.678	-0.2108	0.3240		
8	Midlands	0.0759	0.1337	0.57	0.570	-0.1861	0.3378		
	Scot, Wal, NI	0.1476	0.1591	0.93	0.353	-0.1642	0.4594		
	How often, if a sources to acce	t all, would ss COVID-:	you use tl 19 update	he followir s?	ng digital	Tabloid ne websites	ews		
			-			95% Confi	dence		
q8_3		Coeff	SE	Z score	<i>p</i> value	interval			
Gender	Male	Reference							
	Female	0.0512	0.1082	0.47	0.636	-0.1609	0.2633		
	18-39	Reference							
Age group	40-59	-0.1717	0.1300	-1.32	0.187	-0.4266	0.0832		
	60+	-0.3291	0.1431	-2.30	0.021	-0.6094	-0.0487		
Ethnicity	White	Reference	0.4.400		0.000	0.0004			
	BAME	-0.6280	0.1490	-4.21	0.000	-0.9201	-0.3359		
Social group	ABCI	Keterence	0 1212	1.04	0.200	0.2622	0.1121		
	C2DE	-0.1256	0.1212	- 1.04	0.300	-0.3632	0.1121		
Education	LOW	- 0.6695	0.1052	- 4.05	0.000	-0.9933	-0.3458		
Education	Wealum	-0.62/1	0.1209	- 5.19	0.000	-0.8641	-0.3900		
	11igii South	Reierence							
	North	_0.0371	0 1417	-0.26	0 794	-0.3147	0.2406		
Region	Midlande	0.0452	0 1397	0.20	0.747	-0.2287	0 3190		
	Scot, Wal, NI	0.3084	0.1678	1.84	0.066	- 0.0204	0.6372		

	How often, if a sources to acce	t all, would you use the following digital Broadsheet news ss COVID-19 updates? websites								
q8_4		Coeff	SE	Z score	p value	95% Confi interval	idence			
Condon	Male	Reference								
Gender	Female	0.1206	0.1028	1.17	0.241	-0.0810	0.3222			
	18-39	Reference								
Age group	40-59	-0.0291	0.1214	-0.24	0.810	-0.2670	0.2087			
	60+	0.0158	0.1361	0.12	0.907	-0.2510	0.2826			
Fthnicity	White	Reference								
Lumenty	BAME	-0.4834	0.1453	- 3.33	0.001	-0.7682	-0.1986			
Social group	ABC1	Reference								
social group	C2DE	0.3527	0.1173	3.01	0.003	0.1228	0.5827			
	Low	1.0389	0.1606	6.47	0.000	0.7242	1.3537			
Education	Medium	0.6026	0.1148	5.25	0.000	0.3776	0.8275			
	High	Reference								
	South	Reference								
Region	North	0.2428	0.1353	1.80	0.073	-0.0223	0.5080			
8	Midlands	0.4897	0.1329	3.68	0.000	0.2292	0.7502			
	Scot, Wal, NI	0.2299	0.1578	1.46	0.145	-0.0794	0.5393			
	How often, if a sources to acce	t all, would ss COVID-	you use t 19 update	he followir s?	ng digital	BBC news	website			
q8_5		Coeff	SE	Z score	p value	95% Confi interval	idence			
0.1	Male	Reference			-					
Gender	Female	0.0979	0.1023	0.96	0.339	-0.1026	0.2984			
	18-39	Reference								
Age group	40-59	0.0816	0.1198	0.68	0.496	-0.1532	0.3164			
	60 +	0.0223	0.1346	0.17	0.868	-0.2414	0.2861			
P (1 · · ·	White	Reference								
Ethnicity	BAME	-0.3406	0.1427	-2.39	0.017	-0.6203	-0.0609			
C	ABC1	Reference								
Social group	C2DE	0.4004	0.1177	3.40	0.001	0.1696	0.6312			
	Low	0.3876	0.1578	2.46	0.014	0.0782	0.6970			
Education	Medium	0.1006	0.1122	0.90	0.370	-0.1193	0.3206			
	High	Reference								
	South	Reference								
Region	North	0.0329	0.1335	0.25	0.805	-0.2288	0.2946			
Region	Midlands	-0.0798	0.1312	-0.61	0.543	-0.3369	0.1772			
	Scot, Wal, NI	0.0257	0.1567	0.16	0.870	-0.2814	0.3328			
	How often, if a sources to acce	t all, would ss COVID-	you use t 19 update	he followir s?	ng digital	Social me	dia			
q8_6		Coeff	SE	Z score	p value	95% Confi interval	idence			
. –	Male	Reference			-					
Gender	Female	-0.2297	0.1032	-2.23	0.026	-0.4320	-0.0274			
	18-39	Reference					L			
Age group	40-59	0.3140	0.1209	2.60	0.009	0.0770	0.5510			
	60+	0.8784	0.1381	6.36	0.000	0.6078	1.1490			
-1	White	Reference					<u> </u>			
Ethnicity	BAME	-0.5864	0.1430	-4.10	0.000	- 0.8666	-0.3062			
	ABC1	Reference								
Social group	C2DE	-0.1531	0.1162	- 1.32	0.187	-0.3809	0.0746			
	Low	-0.3400	0.1588	-2.14	0.032	-0.6513	-0.0287			
Education	Medium	-0.3084	0.1140	-2.70	0.007	-0.5319	-0.0849			
	High	Reference				1				
	South	Reference								
D :	North	0.0555	0.1357	0.41	0.683	-0.2105	0.3214			
Region	Midlands	0.0073	0.1327	0.05	0.956	-0.2528	0.2674			
	Scot, Wal, NI	-0.0878	0.1557	-0.56	0.573	-0.3930	0.2175			

	How often, if a traditional sou	t all, would you use the following rces to access COVID-19 updates? Television									
q9_1		Coeff	SE	Z score	p value	95% Confi interval	dence				
• -	Male	Reference			*						
Gender	Female	0.0519	0.0832	0.62	0.533	-0.1113	0.2150				
	18-39	Reference									
Age group	40-59	-0.6109	0.1064	-5.74	0.000	-0.8195	-0.4022				
	60+	-1.0029	0.1108	-9.06	0.000	- 1.2200	-0.7859				
P -1 - 1 -	White	Reference									
Ethnicity	BAME	-0.1000	0.1262	-0.79	0.428	-0.3473	0.1473				
01	ABC1	Reference									
Social group	C2DE	-0.0471	0.0915	-0.51	0.607	-0.2263	0.1322				
	Low	-0.3215	0.1229	- 2.62	0.009	-0.5624	-0.0806				
Education	Medium	-0.1522	0.0954	- 1.59	0.111	-0.3392	0.0349				
	High	Reference									
	South	Reference									
Region	North	-0.3504	0.1086	- 3.23	0.001	-0.5632	-0.1376				
Region	Midlands	-0.3028	0.1067	-2.84	0.005	-0.5120	- 0.0936				
	Scot, Wal, NI	0.0274	0.1311	0.21	0.834	-0.2295	0.2843				
	How often, if a traditional sou	t all, would rces to acce	you use t ss COVII	he followir)-19 updat	ng æs?	Radio					
q9_2		Coeff	SE	Z score	<i>p</i> value	95% Confi interval	dence				
Gender	Male	Reference									
	Female	0.1449	0.0830	1.74	0.081	-0.0179	0.3076				
	18-39	Reference									
Age group	40-59	-0.6250	0.1057	- 5.91	0.000	-0.8322	-0.4178				
	60+	-0.5573	0.1102	-5.06	0.000	-0.7733	-0.3413				
Ethnicity	White	Reference			[
,	BAME	0.1600 0.1257 1.27 0.203 -0.0863 0.4064									
Social group	ABC1	Reference									
0 1	C2DE	0.0891	0.0914	0.97	0.330	-0.0901	0.2683				
	Low	0.4706	0.1235	3.81	0.000	0.2286	0.7126				
Education	Medium	0.0448	0.0954	0.47	0.638	-0.1422	0.2319				
	High	Reference									
	South	Reference	0.4000	0.00	0.400		0.40/4				
Region	North	-0.0876	0.1090	- 0.80	0.422	-0.3014	0.1261				
	Midlands	-0.2649	0.1068	- 2.48	0.013	-0.4/43	-0.0555				
	Scot, Wal, NI	-0.0/44	0.1292	-0.58	0.565	-0.32/5	0.1/88				
	How often, if a traditional sou	t all, would rces to acce	you use t ss COVII	he followir)-19 updat	ng æs?	newspape	rs/				
-0.2		Care	CT.	7		95% Conf	dence				
49_5	Mala	Deference	3E	Z score	<i>p</i> value	Interval					
Gender	Female	0 1458	0.0970	1.50	0.133	-0.0443	0.3359				
	18_39	Reference	0.0770	1.50	0.155	0.0115	0.5555				
Age group	40-59	-0.2488	0 1304	-191	0.056	-0.5045	0.0068				
1180 810 up	60+	-0.8712	0.1303	-6.69	0.000	- 1.1265	-0.6158				
	White	Reference									
Ethnicity	BAME	-0.6434	0.1458	-4.41	0.000	-0.9292	-0.3576				
	ABC1	Reference									
Social group	C2DE	-0.0643	0.1057	- 0.61	0.543	-0.2714	0.1428				
	Low	-0.8453	0.1411	- 5.99	0.000	-1.1219	-0.5686				
Education	Medium	-0.5594	0.1165	-4.80	0.000	-0.7877	-0.3310				
	High	Reference			-						
	South	Reference									
_ .	North	-0.1049	0.1274	-0.82	0.411	-0.3547	0.1449				
Region	Midlands	-0.1911	0.1228	- 1.56	0.120	-0.4317	0.0495				
	Scot, Wal, NI	0.0156	0.1564	0.10	0.921	-0.2910	0.3222				

	How often, if a traditional sou	t all, would you use the following Print broadsheet newspapers							
q9_4		Coeff	SE	Z score	p value	95% Confi interval	idence		
Condon	Male	Reference							
Gender	Female	0.2679	0.0899	2.98	0.003	0.0918	0.4441		
	18-39	Reference							
Age group	40-59	-0.2993	0.1164	-2.57	0.010	-0.5274	-0.0712		
	60 +	-0.7542	0.1199	-6.29	0.000	-0.9891	-0.5193		
Ethnicity	White	Reference							
Linnerty	BAME	-0.3560	0.1335	-2.67	0.008	-0.6177	-0.0943		
Social group	ABC1	Reference							
Social group	C2DE	0.2966	0.1011	2.93	0.003	0.0985	0.4948		
	Low	0.6394	0.1362	4.69	0.000	0.3724	0.9064		
Education	Medium	0.3303	0.1029	3.21	0.001	0.1287	0.5320		
	High	Reference							
	South	Reference							
Region	North	0.0704	0.1184	0.59	0.552	-0.1616	0.3024		
Region	Midlands	0.1005	0.1156	0.87	0.385	-0.1261	0.3270		
	Scot, Wal, NI	0.0446	0.1420	0.31	0.753	-0.2336	0.3229		
	Which, if eithe information or	r, of the foll COVID-1	lowing so 9?	urces do yo	ou prefer to	o use to gat	her		
q10		Coeff	SE	Z score	p value	95% Confi interval	idence		
Conton	Male	Reference							
Gender	Female	-0.0327	0.1051	-0.31	0.756	-0.2386	0.1733		
	18-39	Reference			1				
Age group	40-59	0.7266	0.1273	5.71	0.000	0.4770	0.9762		
	60+	1.6095	0.1407	11.44	0.000	1.3338	1.8853		
Tel 1 1	White	Reference							
Ethnicity	BAME	0.0685	0.1538	0.45	0.656	-0.2329	0.3699		
01	ABC1	Reference			1				
Social group	C2DE	0.2629	0.1169	2.25	0.025	0.0338	0.4919		
	Low	0.5237	0.1572	3.33	0.001	0.2156	0.8318		
Education	Medium	0.2198	0.1177	1.87	0.062	-0.0109	0.4506		
	High	Reference							
	South	Reference							
р.:	North	0.4455	0.1383	3.22	0.001	0.1744	0.7166		
Region	Midlands	0.0746	0.1345	0.55	0.579	-0.1890	0.3382		
	Scot, Wal, NI	-0.1791	0.1635	-1.10	0.273	-0.4996	0.1414		
	To what extent information yo	, if at all, do ou receive fr	o you trus com:	t the COV	ID-19	NHS Web	site		
		a a	07	-		95% Conf	idence		
q11_1	Mala	Соеп	SE	Z score	<i>p</i> value	interval			
Gender	Famala	0.0250	0 1020	0.24	0 000	0.2266	0.1766		
	18 20	Peference	0.1029	-0.24	0.808	-0.2200	0.1700		
A ge group	40.59	0.1723	0.1290	1.34	0.182	-0.0805	0.4251		
Age group	40-39 60+	0.1725	0.1250	1.94	0.132	0.0020	0.5319		
	White	Deference	0.1332	1.97	0.040	0.0020	0.5519		
Ethnicity	BAME	0 2010	0.1478	1.08	0.048	0.0023	0.5815		
	ABC1	Deference	0.1478	1.90	0.040	0.0023	0.3813		
Social group	CODE	0.4568	0.1101	4.15	0.000	0.2409	0.6727		
	Low	0.4300	0.1101	4.13	0.000	0.2409	0.6727		
Education	Madium	0.3499	0.1480	2.30	0.018	0.0398	0.0400		
Education	Lich	0.15/2 Doferro	0.1188	1.32	0.186	-0.0/5/	0.3901		
	South	Reierence							
	North	A OF 12	0 1 2 2 4	0.41	0.695	0.2157	0.2072		
Region	Midlanda	-0.0542	0.1354	- 0.41	0.085	-0.3150	0.20/3		
	Scot. Wal NI	-0 2754	0.1670	-1.65	0.099	-0.6027	0.0519		
	,, , , , , , , , , , , , , , , , ,	5.2754	0.10/0	1.05		0.002/			

	To what extent information yo	if at all, do you trust the COVID-19 Other, non-NHS, healthcare websites								
q11_2		Coeff	SE	Z score	p value	95% Confi interval	idence			
<u>c</u> 1	Male	Reference								
Gender	Female	-0.0565	0.0930	-0.61	0.543	-0.2388	0.1257			
	18-39	Reference								
Age group	40-59	-0.1654	0.1153	-1.44	0.151	-0.3914	0.0605			
	60+	0.3121	0.1218	2.56	0.010	0.0734	0.5507			
Ethnicity	White	Reference								
Etimenty	BAME	-0.2725	0.1391	- 1.96	0.050	-0.5451	0.0002			
Social group	ABC1	Reference								
Social group	C2DE	0.2236	0.1042	2.15	0.032	0.0194	0.4277			
	Low	0.2266	0.1380	1.64	0.101	-0.0440	0.4971			
Education	Medium	0.0475	0.1054	0.45	0.652	-0.1592	0.2541			
	High	Reference								
	South	Reference								
Region	North	0.1049	0.1203	0.87	0.383	-0.1309	0.3406			
Region	Midlands	-0.2370	0.1203	- 1.97	0.049	-0.4729	-0.0011			
	Scot, Wal, NI	-0.0277	0.1470	-0.19	0.850	-0.3159	0.2604			
	To what extent information yo	, if at all, do ou receive fr	o you trus com:	t the COV	ID-19	Tabloid ne websites	ews			
a11 3		Coeff	SE .	7 score	t value	95% Confi	idence			
q11_5	Male	Reference	3L	2 score	<i>p</i> value	inter var				
Gender	Female	-0.0557	0.0941	-0.59	0 554	-0.2401	0.1287			
	18-39	Reference	0.0941	0.59	0.554	0.2401	0.1207			
Age group	40-59	-0.0147	0 1183	-0.12	0.901	-0.2465	0.2172			
	60+	-0.5060	0.1234	-4.10	0.000	-0.7479	-0.2641			
	White	Reference								
Ethnicity	BAME	-0.3464	0 1392	- 2 49	0.013	-0.6193	-0.0735			
	ABC1	Reference	0.1392	2.19	0.015	0.0175	0.0755			
Social group	C2DE	- 0.2544	0.1037	- 2.45	0.014	-0.4577	-0.0512			
	Low	- 0.6805	0.1388	- 4.90	0.000	-0.9525	-0.4085			
Education	Medium	-0.6597	0.1080	-6.11	0.000	-0.8713	-0.4480			
	High	Reference								
	South	Reference								
	North	0.0197	0.1225	0.16	0.872	-0.2203	0.2597			
Region	Midlands	-0.0512	0.1199	-0.43	0.670	-0.2862	0.1838			
	Scot, Wal, NI	0.2600	0.1512	1.72	0.085	-0.0362	0.5563			
	To what extent	, if at all, do	o you trus	t the COV	ID-19	Broadshee	et news			
	information yo	ou receive fr	om:			websites				
q11 4		Coeff	SE	Z score	p value	95% Confi interval	idence			
	Male	Reference			-					
Gender	Female	0.0475	0.0928	0.51	0.609	-0.1344	0.2293			
	18-39	Reference								
Age group	40-59	-0.0116	0.1140	-0.10	0.919	-0.2350	0.2119			
	60+	-0.0329	0.1201	-0.27	0.784	-0.2683	0.2025			
	White	Reference								
Ethnicity	BAME	-0.2663	0.1376	- 1.94	0.053	-0.5360	0.0034			
	ABC1	Reference								
Social group	C2DE	0.3693	0.1025	3.60	0.000	0.1684	0.5702			
	Low	1.0575	0.1409	7.50	0.000	0.7812	1.3337			
Education	Medium	0.6312	0.1049	6.02	0.000	0.4255	0.8369			
	High	Reference								
	South	Reference								
Destau	North	0.3204	0.1217	2.63	0.008	0.0818	0.5590			
Region	Midlands	0.2463	0.1182	2.08	0.037	0.0146	0.4780			
	Scot, Wal, NI	0.2141	0.1441	1.49	0.137	-0.0684	0.4966			

	To what extent information yo	, if at all, do you trust the COVID-19 ou receive from: BBC news website									
		c c	07	-		95% Conf	idence				
q11_5	241	Coeff	SE	Z score	<i>p</i> value	interval					
Gender	Male	Reference	0.0000	0.17	0.964	0.1590	0.1902				
	18 30	0.0152 Reference	0.0888	0.17	0.864	-0.1589	0.1892				
Age group	40-59	0.0859	0 1099	0.78	0.434	-0.1295	0 3014				
nge group	60+	-0.0140	0.1150	-0.12	0.903	-0.2394	0.2115				
	White	Reference	011100	0.112	01500	0.2031	0.2110				
Ethnicity	BAME	-0.2507	0.1317	- 1.90	0.057	-0.5089	0.0075				
	ABC1	Reference									
Social group	C2DE	0.4022	0.0985	4.08	0.000	0.2090	0.5953				
	Low	0.4702	0.1319	3.56	0.000	0.2116	0.7287				
Education	Medium	0.3655	0.1006	3.63	0.000	0.1683	0.5628				
	High	Reference									
	South	Reference									
	North	-0.1055	0.1165	-0.91	0.365	-0.3340	0.1229				
Region	Midlands	-0.0035	0.1128	-0.03	0.975	-0.2246	0.2176				
	Scot, Wal, NI	-0.0110	0.1380	-0.08	0.937	-0.2814	0.2594				
	To what extent	, if at all, do	o you trus	t the COV	ID-19						
	information yo	ou receive fr	om:			Social me	dia				
a11 6		Coeff	SE	Z score	t value	95% Confi interval	idence				
4	Male	Reference	02	2.00010	P ·uiue						
Gender	Female	-0.3378	0.0927	- 3.64	0.000	- 0.5195	-0.1561				
	18-39	Reference									
Age group	40-59	0.0686	0.1123	0.61	0.541	-0.1515	0.2887				
001	60+	0.4424	0.1205	3.67	0.000	0.2062	0.6786				
	White	Reference					I				
Ethnicity	BAME	-0.6974	0.1357	-5.14	0.000	-0.9634	-0.4314				
	ABC1	Reference					I				
Social group	C2DE	0.0461	0.1020	0.45	0.651	-0.1538	0.2460				
	Low	-0.6170	0.1393	-4.43	0.000	-0.8900	-0.3440				
Education	Medium	-0.2970	0.1042	- 2.85	0.004	-0.5011	-0.0928				
	High	Reference			1						
	South	Reference									
Destan	North	0.2275	0.1199	1.90	0.058	-0.0075	0.4625				
Region	Midlands	0.1044	0.1203	0.87	0.385	-0.1313	0.3402				
	Scot, Wal, NI	-0.1350	0.1395	-0.97	0.333	-0.4084	0.1385				
	How confident	are you in	telling ap	art reliable	COVID-1	9 informat	ion online				
	or through app	os?			[0.50 0 0					
q12		Coeff	SE	Z score	p value	95% Confi interval	Idence				
- -	Male	Reference			-						
Gender	Female	0.1048	0.0962	1.09	0.276	-0.0838	0.2934				
	18-39	Reference									
Age group	40-59	-0.0945	0.1168	-0.81	0.419	-0.3233	0.1344				
	60+	0.2637	0.1263	2.09	0.037	0.0161	0.5113				
Feb at sites	White	Reference			1						
Ethnicity	BAME	0.1883	0.1378	1.37	0.172	-0.0818	0.4584				
Control amount	ABC1	Reference									
social group	C2DE	0.3390	0.1058	3.21	0.001	0.1317	0.5463				
	Low	0.4029	0.1425	2.83	0.005	0.1237	0.6821				
Education	Medium	0.1202	0.1083	1.11	0.267	-0.0920	0.3325				
	High	Reference									
	South	Reference									
Region	North	-0.0299	0.1247	-0.24	0.810	-0.2743	0.2145				
	Midlands	-0.1126	0.1233	-0.91	0.361	-0.3543	0.1290				
	Scot, Wal, NI	-0.2432	0.1505	-1.62	0.106	-0.5382	0.0517				

	If you saw info following woul	rmation on COVID-19, which of the d contribute towards your trust in it? That it comes from the Government								
q13_1		Coeff	Std. Err	Z score	p value	95% Confi interval	idence			
0.1	Male	Reference								
Gender	Female	0.0044	0.0934	0.05	0.962	-0.1786	0.1875			
	18-39	Reference								
Age group	40-59	-0.2408	0.1182	-2.04	0.042	-0.4725	-0.0091			
	60+	-0.1385	0.1213	-1.14	0.253	-0.3763	0.0992			
Ethnicity	White	Reference								
	BAME	-0.0633	0.1391	-0.46	0.649	-0.3359	0.2093			
Social group	ABC1	Reference								
ooonin group	C2DE	-0.1099	0.1030	- 1.07	0.286	-0.3118	0.0919			
	Low	0.1641	0.1372	1.20	0.232	-0.1049	0.4330			
Education	Medium	0.0774	0.1075	0.72	0.472	-0.1334	0.2882			
	High	Reference								
	South	Reference								
Region	North	-0.0307	0.1223	-0.25	0.802	-0.2704	0.2090			
	Midlands	0.0175	0.1188	0.15	0.883	-0.2154	0.2505			
	Scot, Wal, NI	-0.1789	0.1474	-1.21	0.225	-0.4678	0.1101			
	If you saw info following woul	rmation on d contribut	COVID-	19, which o your trus	of the t in it?	That it con scientists/ institution	nes from scientific 1s			
q13_2		Coeff	SE	Z score	p value	95% Confi interval	idence			
	Male	Reference			-					
Gender	Female	0.1400	0.1056	1.33	0.185	-0.0670	0.3470			
	18-39	Reference								
Age group	40-59	-0.1364	0.1368	-1.00	0.319	-0.4044	0.1317			
	60+	-0.1413	0.1405	-1.01	0.314	-0.4167	0.1340			
Ethnisita	White	Reference								
Ethnicity	BAME	-0.5234	-0.8290	-0.2178						
Social group	ABC1	Reference								
social group	C2DE	-0.3815	0.1116	- 3.42	0.001	-0.6002	-0.1628			
	Low	- 1.0539	0.1535	- 6.87	0.000	-1.3547	-0.7530			
Education	Medium	-0.6077	0.1295	- 4.69	0.000	-0.8615	-0.3540			
	High	Reference								
	South	Reference								
Region	North	-0.2404	0.1403	- 1.71	0.087	-0.5153	0.0346			
Region	Midlands	-0.3917	0.1342	-2.92	0.004	-0.6548	-0.1286			
	Scot, Wal, NI	-0.1932	0.1655	-1.17	0.243	-0.5177	0.1312			
	If you saw info following woul	rmation on d contribut	COVID-	19, which o your trus	of the t in it?	The source from	e it comes			
q13_6		Coeff	SE	Z score	p value	95% Confi interval	idence			
Gender	Male	Reference								
Genuei	Female	-0.1464	0.0990	-1.48	0.139	-0.3404	0.0476			
	18-39	Reference								
Age group	40-59	-0.4019	0.1209	-3.33	0.001	-0.6388	-0.1650			
	60+	-0.7401	0.1291	-5.73	0.000	-0.9930	-0.4871			
Ethnicity	White	Reference								
	BAME	-0.0372	0.1423	-0.26	0.794	-0.3162	0.2417			
Social group	ABC1	Reference								
8 1	C2DE	-0.2930	0.1112	-2.64	0.008	-0.5109	-0.0752			
	Low	-0.8875	0.1501	- 5.91	0.000	-1.1817	-0.5933			
Education	Medium	-0.6311	0.1099	-5.74	0.000	-0.8464	-0.4157			
	High	Reference								
	South	Reference				_				
Region	North	-0.2000	0.1300	- 1.54	0.124	-0.4549	0.0548			
	Midlands	-0.2475	0.12/1	- 1.95	0.052	-0.4966	0.0017			
	Scot, Wal, NI	-0.0121	0.1534	-0.08	0.937	-0.3127	0.2885			

	How often do you double check online or app-based health information that you receive?								
q14		Coeff	SE	Z score	p value	95% Confidence interval			
Carla	Male	Reference							
Gender	Female	-0.1068	0.0964	-1.11	0.268	-0.2958	0.0822		
	18-39	Reference							
Age group	40-59	-0.1628	0.1159	-1.41	0.160	- 0.3899	0.0642		
	60+	0.1477	0.1289	1.15	0.252	-0.1050	0.4004		
Tel 1 1	White	Reference			1				
Ethnicity	BAME	-0.1274	0.1343	-0.95	0.343	-0.3907	0.1358		
01	ABC1	Reference			1				
Social group	C2DE	0.3778	0.1097	3.44	0.001	0.1627	0.5928		
	Low	0.8407	0.1511	5.56	0.000	0.5445	1.1369		
Education	Medium	0.1489	0.1076	1.38	0.166	-0.0619	0.3597		
	High	Reference							
	South	Reference							
	North	0.0520	0.1258	0.41	0.679	-0.1945	0.2986		
Region	Midlands	0.0600	0.1242	0.48	0.629	-0.1834	0.3035		
	Scot, Wal, NI	0.0104	0.1517	0.07	0.945	-0.2870	0.3078		
How likely are you to engage with digital resources if they were directly li									
	is are controlling	g are pull				95% Conf	idence		
q15		Coeff	SE	Z score	<i>p</i> value	interval			
Cander	Male	Reference							
Gender	Female	-0.1072	0.0821	-1.31	0.192	-0.2681	0.0537		
	18-39	Reference							
Age group	40-59	0.0973	0.1012	0.96	0.336	-0.1010	0.2956		
	60+	0.3608	0.1071	3.37	0.001	0.1509	0.5707		
Feb	White	Reference							
Ethnicity	BAME	0.2321	0.1198	1.94	0.053	-0.0027	0.4669		
01	ABC1	Reference							
Social group	C2DE	0.4160	0.0903	4.61	0.000	0.2390	0.5930		
	Low	0.1579	0.1199	1.32	0.188	-0.0771	0.3928		
Education	Medium	0.0178	0.0937	0.19	0.849	-0.1658	0.2014		
	High	Reference							
	South	Reference							
D	North	0.2887	0.1082	2.67	0.008	0.0768	0.5007		
Region	Midlands	0.0410	0.1037	0.40	0.693	-0.1622	0.2442		
	Scot, Wal, NI	0.2440	0.1275	1.91	0.056	- 0.0059	0.4939		
	How comfortal personal data v	able are you in sharing the following with a Government COVID-19 contact							
	tracing app?					NHS num	ber		
q16_1		Coeff	SE	Z score	p value	95% Confidence interval			
0.1	Male	Reference							
Gender	Female	0.2449	0.0822	2.98	0.003	0.0838	0.4061		
	18-39	Reference							
Age group	40-59	0.1674	0.1031	1.62	0.104	-0.0346	0.3695		
	60 +	-0.1668	0.1063	-1.57	0.116	-0.3751	0.0414		
- 1	White	Reference							
Ethnicity	BAME	0.1770	0.1211	1.46	0.144	-0.0604	0.4145		
	ABC1	Reference							
Social group	C2DE	0.1285	0.0895	1.44	0.151	-0.0469	0.3040		
	Low	0.1292	0.1196	1.08	0.280	-0.1052	0.3637		
Education	Medium	-0.0076	0.0943	-0.08	0.935	-0.1924	0.1771		
	High	Reference			l				
	South	Reference							
	North	0.0569	0.1075	0.53	0.597	-0.1538	0.2677		
Region	Midlands	-0.0437	0.1037	-0.42	0.673	-0.2469	0.1595		
	Scot, Wal, NI	0.0877	0.1294	0.68	0.498	-0.1659	0.3413		

	How comfortal personal data w tracing app?	Age							
q16_2		Coeff	SE	Z score	p value	95% Confidence interval			
Condon	Male	Reference							
Gender	Female	0.1095	0.0844	1.30	0.195	- 0.0560	0.2750		
	18-39	Reference							
Age group	40-59	0.1224	0.1054	1.16	0.245	-0.0841	0.3289		
	60+	-0.1541	0.1100	-1.40	0.161	- 0.3696	0.0615		
Ethnicity	White	Reference							
Ethnicity	BAME	0.2155 0.1220 1.77 0.077 -0.0236 0.4546							
C	ABC1	Reference							
Social group	C2DE	0.1332	0.0922	1.44	0.149	-0.0476	0.3139		
Education	Low	0.0580	0.1234	0.47	0.638	-0.1838	0.2997		
	Medium	-0.0732	0.0969	-0.76	0.450	-0.2631	0.1167		
	High	Reference							
	South	Reference							
Destau	North	0.0548	0.1104	0.50	0.620	-0.1616	0.2711		
Region	Midlands	0.0899	0.1066	0.84	0.399	-0.1191	0.2989		
	Scot, Wal, NI	0.0658	0.1329	0.50	0.621	-0.1946	0.3262		
	How comfortal personal data v tracing app?	ble are you in sharing the following with a Government COVID-19 contact Location							
q16_3		Coeff	SE	Z score	p value	95% Conf interval	idence		
Candar	Male	Reference							
Gender	Female	0.1284	0.0827	1.55	0.120	-0.0336	0.2904		
	18-39	Reference							
Age Group	40-59	-0.2813	0.1037	-2.71	0.007	-0.4845	-0.0781		
	60+	-0.6367	0.1079	- 5.90	0.000	-0.8481	-0.4252		
Ethnisita	White	Reference							
Ethnicity	BAME	0.2632	0.1204	2.19	0.029	0.0272	0.4992		
Sectol Course	ABC1	Reference							
Social Group	C2DE	0.2122	0.0901	2.36	0.018	0.0357	0.3888		
	Low	0.0716	0.1205	0.59	0.553	-0.1646	0.3077		
Education	Medium	0.0403	0.0950	0.42	0.671	-0.1459	0.2264		
	High	Reference							
	South	Reference							
Desion	North	-0.0373	0.1083	-0.34	0.731	-0.2495	0.1750		
Region	Midlands	-0.0062	0.1041	-0.06	0.952	-0.2103	0.1979		
	Scot, Wal, NI	0.1129	0.1289	0.88	0.381	-0.1397	0.3654		
	How comfortable are you in sharing the following personal data with a Government COVID-19 contact tracing app? Medical history								
q16_4		Coeff	SE	Z score	<i>p</i> value	95% Confidence interval			
Gender	Male	Reference							
Gender	Female	0.0879	0.0819	1.07	0.283	-0.0726	0.2484		
	18-39	Reference							
Age group	40-59	0.1331	0.1031	1.29	0.197	- 0.0689	0.3352		
	60+	-0.1849	0.1059	-1.75	0.081	- 0.3925	0.0226		
Tel i si ter	White	Reference							
Ennicity	BAME	0.0222	0.1206	0.18	0.854	-0.2141	0.2586		
0	ABC1	Reference					•		
Social group	C2DE	0.0707	0.0900	0.79	0.432	-0.1057	0.2471		
	Low	-0.1372	0.1210	-1.13	0.257	-0.3744	0.1000		
Education	Medium	-0.1391	0.0939	-1.48	0.138	-0.3230	0.0449		
	High	Reference							
Continued									

	How comfortal personal data v tracing app?	Iow comfortable are you in sharing the following bersonal data with a Government COVID-19 contact racing app?					Medical history		
q16_4		Coeff	SE	Z score	p value	95% Confidence interval			
	South	Reference			-				
	North	0.0393	0.1074	0.37	0.715	-0.1713	0.2498		
Region	Midlands	-0.0922	0.1033	-0.89	0.372	-0.2947	0.1103		
	Scot, Wal, NI	0.0864	0.1285	0.67	0.502	-0.1656	0.3383		
	How comfortal personal data v tracing app?	able are you in sharing the following with an industry-led COVID-19 contact NHS number					ber		
q17_1		Coeff	SE	Z score	<i>p</i> value	95% Conf interval	idence		
Conton	Male	Reference							
Gender	Female	0.1600	0.0865	1.85	0.064	- 0.0095	0.3296		
Age group	18-39	Reference							
	40-59	0.0164	0.1087	0.15	0.880	-0.1966	0.2294		
	60 +	0.1777	0.1128	1.58	0.115	-0.0434	0.3989		
	White	Reference							
Ethnicity	BAME	-0.1019	0.1296	-0.79	0.432	-0.3560	0.1521		
	ABC1	Reference							
Social group	C2DE	-0.0358	0.0951	-0.38	0.707	-0.2222	0.1507		
	Low	0.0090	0.1275	0.07	0.944	-0.2410	0.2589		
Education	Medium	- 0.0089	0.0995	- 0.09	0.929	-0.2038	0.1860		
	High	Reference							
	South Reference								
	North	0.1277	0 1 1 4 7	1 1 1	0.266	-0.0971	0.3526		
Region	Midlands	-0.1230	0.1087	-1.14	0.254	-0.3370	0.0320		
	Sect Mal NI	-0.1239	0.1067	- 1.14	0.234	-0.3370	0.0091		
	Josepherete	0.1038	0.1304	0.70	0.447	-0.1034	0.3710		
	personal data v tracing app?	ble are you in sharing the following with an industry-led COVID-19 contact Age							
q17_2		Coeff	SE	Z score	<i>p</i> value	95% Conf interval	idence		
Gender	Male	Reference			[
	Female	0.1224	0.0825	1.48	0.138	-0.0393	0.2840		
	18-39	Reference							
Age group	40-59	0.2048	0.1035	1.98	0.048	0.0019	0.4077		
	60+	0.6356	0.1070	5.94	0.000	0.4258	0.8453		
Ethnicity	White	Reference							
Luniony	BAME	0.1611	0.1244	1.29	0.195	-0.0827	0.4048		
Social group	ABC1	Reference							
social group	C2DE	-0.0308	0.0897	-0.34	0.732	- 0.2066	0.1451		
	Low	0.1245	0.1205	1.03	0.301	-0.1116	0.3606		
Education	Medium	0.1097	0.0940	1.17	0.243	-0.0746	0.2940		
	High	Reference							
	South	Reference							
Destau	North	0.2006	0.1080	1.86	0.063	-0.0110	0.4122		
Region	Midlands	0.0565	0.1038	0.54	0.586	-0.1470	0.2599		
	Scot, Wal, NI	-0.0382	0.1299	-0.29	0.769	-0.2928	0.2165		
	How comfortal personal data v tracing app?	table are you in sharing the following a with an industry-led COVID-19 contact							
q17_3		Coeff	SE	Z score	p value	95% Conf interval	idence		
	Male	Reference			-				
Gender	Female	0.2100	0.0828	2.54	0.011	0.0477	0.3723		
	18-39	Reference							
Age group	40-59	-0.0181	0 1039	-0.17	0.861	-0.2217	0.1854		
9. 9. 0ur	60+	0 3386	0 1075	315	0.002	0.1280	0 5492		
		5.0000	0.1075		0.002				

broadsheet newspaper sources and the BBC are not as trusted as information sources by those from low social grades and low educational attainment groups (p < 0.01).

Two distinct clusters of responses for this domain of questions (Question 11) were identified. Panel C of Fig. 2 shows the responses of each cluster to each of the constituent questions on which clustering is performed. Clusters were characterised post-hoc based on their responses as either 'mistrustful of non-NHS information' (37.5%) or 'Trusting of NHS, broadsheets and BBC' (62.5%).

Scientific endorsement of information from figures, such as Professor Chris Whitty, is seen as the most important contributor towards trust (70% trust rating). Despite this high rating, in comparison to their reference groups, respondents from BAME backgrounds, lower social grades, low educational attainment groups and those who reside in the Midlands are less likely to trust information that has scientific endorsement. Moreover, the government trust rating was only 40%, with no one demographic either more or less inclined to trust government sourced information in comparison to the reference group. Lastly, those with a high education attainment (213/634) are twice as likely to double check information that they encounter through digital resources than those of a low education attainment (80/535) (Question 14).

Contact tracing. 832/2040 (41%) are unlikely to engage with a digital contact tracing programme, even in the event that compliance was directly linked to easing of quarantine measures. In comparison to their respective reference groups, those above the age of 60 (p < 0.01), those from Northern regions (p < 0.01) and those of the lowest social grade are significantly less likely to engage in the contact tracing programme (p < 0.05) (Question 15).

With respect to industry led contact tracing apps, respondents are uncomfortable with sharing their NHS number (1524/2040 (75%)), medical history (1538/2040 (75%)) and location (1199/2040 (59%)). Those aged above 60 are significantly more uncomfortable in sharing data related to age, location and medical history when using industry led apps, in comparison to their reference counterparts (p < 0.01) (Question 17). In comparison, with respect to government led contact tracing apps, there is less discomfort at sharing NHS number (795/2040 (39%)), medical history (935/2040 (46%)) and location (772/3040 (38%)) (Question 16). With government led contact tracing apps, those of a BAME background and lower social grades are less comfortable in sharing their location than their reference counterparts (p < 0.05), whereas those over the 40+ are more likely to share their location (p < 0.01).

Two distinct clusters of responses for this domain of questions (Questions 15, 16 and 17) were identified. Panel D of Fig. 2 shows the responses of each cluster to each of the constituent questions on which clustering is performed. Clusters were characterised post-hoc based on their responses as either 'comfortable with apps' (59.3%) or 'uncomfortable with apps' (40.7%).

A Brant test was performed to test the proportional odds assumption with respect to each of the ordinal logistic regression models (Appendix 2). We note that the proportional odds assumption was valid except in Questions 2 and 12–17. No single covariate was consistently responsible for violation of the proportional odds assumption across these models. This is likely secondary to the large sample size as well as the high number of explanatory variables included in the models²⁵.

Discussion

This study finds that the UK population exhibits (1) diverse preferences for accessing public health information, (2) mixed self-rated ability to use digital health resources and (3) variable levels of engagement with digital public health approaches, resulting in incomplete digital inclusivity during the COVID-19 pandemic. This study has shown there is a consistent pattern of older people, those of lower social grades and those of lower educational attainment levels displaying greater vulnerability to digital exclusion through poorer access to devices, diminished ability to navigate digital resources pertaining to public health efforts, and reduced inclination to interact with them. In contrast, reported attitudes and behaviours amongst BAME groups are more complex, and do not uniformly align with risk for digital exclusion. With respect to the barriers to digital inclusion, the findings somewhat corroborate the high levels of internet and device availability in the UK as previously described⁹. However, our results also reveal disparities with respect the ability to use and engagement with digital solutions. These findings are particularly marked with regards to digital public health messaging, disease surveillance and contact tracing.

As this was an online survey, we did not expressly ask about internet connectivity, which would have been requisite for respondents. Early 2020 national data⁸ shows that 96% of the UK have internet access and whilst the remaining 4% have not been represented in this work, given they have no access, they would also not be able to engage with digital public health strategies, being the most digitally excluded. Our findings are, therefore, likely to be conservative estimate of the extent of digital exclusion amongst the UK population. Laptop, personal computer or phone access were relatively high across participants of all demographic groups and more frequently used than other device types. Whilst the pandemic has interrupted the publication of the full range of annual ONS data on this topic, these figures appear consistent with other sources²⁶.

National data shows that internet connection in households with an adult aged over 65 years has increased to 80% this year and was predominantly used by the elderly for maintaining social interaction and online shopping prior to the pandemic⁸. Although our data show a continued trend in older, low social grade and lower educational attainment subpopulations using the internet for social interaction, this did not translate to many of these participants accessing digital COVID-19-related public health messaging or contact tracing apps. This discrepancy may be explained by the combination of lower self-reported ability to find and use such information, as well as concerns that participants raised about the reliability of online health information. Although these groups prefer television or print media for COVID-19 updates, and have a degree of mistrust of online

	How comforta personal data tracing app?	ortable are you in sharing the following lata with an industry-led COVID-19 contact p?					Location	
q17_3		Coeff	SE	Z score	p value	95% Confidence interval		
Ethnicity	White	Reference						
Lunneity	BAME	0.2095	0.1244	1.68	0.092	-0.0343	0.4534	
Social group	ABC1	Reference						
	C2DE	0.0834	0.0900	0.93	0.354	- 0.0930	0.2597	
Education	Low	0.1403	0.1213	1.16	0.248	- 0.0975	0.3782	
	Medium	0.0161	0.0949	0.17	0.865	-0.1699	0.2021	
	High	Reference						
	South	Reference						
Desian	North	0.1173	0.1092	1.07	0.283	-0.0967	0.3313	
Region	Midlands	-0.0254	0.1048	-0.24	0.809	-0.2309	0.1801	
	Scot, Wal, NI	-0.0473	0.1294	-0.37	0.715	-0.3008	0.2063	
	How comforta personal data tracing app?	ble are you in sharing the following with an industry-led COVID-19 contact Medical history						
q17_4		Coeff SE Z score p value 95% Confidence interval						
Condon	Male	Reference						
Gender	Female	-0.0029	0.0872	-0.03	0.973	-0.1738	0.1680	
	18-39	Reference						
Age group	40-59	0.0872	0.1096	0.80	0.426	-0.1276	0.3020	
	60+	0.2972	0.1136	2.62	0.009	0.0746	0.5198	
Ethnicity	White	Reference						
		Incluterence						
	BAME	-0.1052	0.1310	-0.80	0.422	-0.3619	0.1515	
Social group	BAME ABC1	-0.1052 Reference	0.1310	-0.80	0.422	-0.3619	0.1515	
Social group	BAME ABC1 C2DE	-0.1052 Reference -0.0718	0.1310	-0.80	0.422	-0.3619	0.1515	
Social group	BAME ABC1 C2DE Low	-0.1052 Reference -0.0718 -0.1227	0.1310 0.0954 0.1291	- 0.80 - 0.75 - 0.95	0.422 0.452 0.342	-0.3619 -0.2587 -0.3757	0.1515 0.1152 0.1304	
Social group Education	BAME ABC1 C2DE Low Medium	-0.1052 Reference -0.0718 -0.1227 -0.1177	0.1310 0.0954 0.1291 0.1003	- 0.80 - 0.75 - 0.95 - 1.17	0.422 0.452 0.342 0.240	-0.3619 -0.2587 -0.3757 -0.3142	0.1515 0.1152 0.1304 0.0788	
Social group Education	BAME ABC1 C2DE Low Medium High	-0.1052 Reference -0.0718 -0.1227 -0.1177 Reference	0.1310 0.0954 0.1291 0.1003	-0.80 -0.75 -0.95 -1.17	0.422 0.452 0.342 0.240	-0.3619 -0.2587 -0.3757 -0.3142	0.1515 0.1152 0.1304 0.0788	
Social group Education	BAME ABC1 C2DE Low Medium High South	-0.1052 Reference -0.0718 -0.1227 -0.1177 Reference Reference	0.1310 0.0954 0.1291 0.1003	-0.80 -0.75 -0.95 -1.17	0.422 0.452 0.342 0.240	-0.3619 -0.2587 -0.3757 -0.3142	0.1515 0.1152 0.1304 0.0788	
Social group Education	BAME ABC1 C2DE Low Medium High South North	-0.1052 Reference -0.0718 -0.1227 -0.1177 Reference Reference 0.0615	0.1310 0.0954 0.1291 0.1003 0.1153	-0.80 -0.75 -0.95 -1.17 0.53	0.422 0.452 0.342 0.240	-0.3619 -0.2587 -0.3757 -0.3142 -0.1645	0.1515 0.1152 0.1304 0.0788 0.2874	
Social group Education Region	BAME BAME ABC1 C2DE Low Medium High South North Midlands	- 0.1052 Reference - 0.0718 - 0.1227 - 0.1177 Reference Reference 0.0615 - 0.1427	0.1310 0.0954 0.1291 0.1003 0.1153 0.1095	-0.80 -0.75 -0.95 -1.17 0.53 -1.30	0.422 0.452 0.342 0.240 0.594 0.192	-0.3619 -0.2587 -0.3757 -0.3142 -0.1645 -0.3573	0.1515 0.1152 0.1304 0.0788 0.2874 0.0719	



resources, including government endorsed media, they continue to use digital devices for social media. Yet, familiarity with, and frequent use of, such platforms in combination with knowledge gaps in identifying reliable information leave people open to the spread of health misinformation²⁷. Notable COVID-19-specific examples of misinformation have led to the destruction of 5G network towers²⁸, case reports of ingested disinfectant²⁹ and poor compliance with face masks³⁰.

The study also reveals factors contributing to scant use of apps for COVID-19 disease surveillance or contact tracing. In the first instance, the elderly, those of lower social grades and of lower educational attainment had less smartphone access³¹, however, sentiments of trust and privacy played a greater role. Amongst the total study population, 41% report being unlikely to engage with such an app, citing reduced trust and concerns sharing health data with non-NHS private partners, such as Apple and Google. These trends were more pronounced still amongst older and those of lower social grades. This is interesting in view of the less secure centralised data storage option preferred by the UK government versus the decentralized but more secure alternative used by the tech giants³². This counterfactual highlights potential knowledge gaps but also the role of privacy and trust in encouraging digital inclusion³³. Furthermore, these barriers to engagement undermine the efficacy of a contact tracing app which requires up to an estimated 60% uptake³⁴, particularly in the absence of an operational test and trace system, as was the case in the UK at the time of the study being conducted³⁵.

The picture of digital exclusion gleaned from this study is far more mixed for the BAME cohort. This is perhaps as BAME is an umbrella that encompasses much heterogeneity in cultural background, income level and education, all of which could have a greater effect on digital inclusion. As such, studying the attitudes and views of BAME people as a single group is unlikely to be an adequate approach³⁶ and focus should be placed on engaging with those without English as a first language, who are recognised as being at risk from the digital divide⁹.





Although this is a UK-based study, the digital divide is by no means a UK-specific phenomenon. The United Nations Sustainable Development Goal 9.c of providing "universal and affordable access to the Internet in least developed countries by 2020" has not been met³⁷. Despite modestly improving internet access rates globally, low digital literacy skills remain a barrier to meaningful participation in a digital society. It is therefore unsurprising that similarly themed studies conducted in countries as varied as Ghana³⁸ and the Netherlands³⁹ suggest that groups vulnerable to digital exclusion have struggled to locate and engage with COVID-19 information disseminated via digital media. This divide is also seen in public-facing clinical digital health interventions during the pandemic, namely tele-medicine services^{40,41}.

Despite increasingly high levels of internet connection and device availability and the pandemic accelerating digital technology adoption, we report a gradient among older, lower social grades and lower education attainment demographic groups interacting with digital public health approaches. The inability to promptly access and understand online information and services prevents individuals from taking protective steps against COVID-19. These same groups are also at higher risk from COVID-19, so the observed digital divide effectively compounds health risks. This suggests that digital inequality potentiates vulnerability to the pandemic, thereby further increasing health inequalities. This is in keeping with previous descriptions of digital inclusion as a wider determinant of health^{42,43}.

Recommendations

Failing to consider how digital interventions can exacerbate health inequalities could be disastrous. Instead, previous national commitments to alleviate digital exclusion⁴⁴ should be reaffirmed. The clustering of responses reveals a lack of consensus across key issues of acquisition and consumption of digital healthcare data, implying

that there is unlikely to be a 'one-size-fits-all' digital strategy to provide equitable coverage across all regions and populations. As such, a multifaceted response, targeting the barriers to digital inclusion is essential.

Access. Though we found relatively high levels of connectivity within our cohort, attention should be given to emerging groups who struggle with slow connection speeds or expensive internet service provision that impede education or employment. We did not study children's experiences but governmental programmes to provide either new or refurbished⁴⁵ laptops and internet connection to children⁴⁶ provides multigenerational support to engage in digital health services⁴⁷.

Skills. Closer collaboration between the technology sector, non-governmental organisations and governmental stakeholders can produce solutions that are scalable and robust. For example, in the USA, Microsoft have provided funding and infrastructural support to provide both devices and access to digital skills training to the Public Library Association⁴⁸. Integration of digital skills assessments within routine services, such as GP services, can also help identify individuals who are at risk of the digital divide and would require support.

Engagement. Greater direct communication between digital service providers and communities can assuage mistrust. The NHS Widening Digital Participation Programme⁴⁹ trains 'digital champions' who are trusted community members and able to provide support to less confident members of the community group⁵⁰. Similarly contact-tracing app developers can and have increased trust and uptake through public information campaigns to improve understanding and transparency in lay terms⁵¹.

Whilst many of these strategies are primarily framed at bridging the digital divide during the COVID-19 pandemic, there is evidence to suggest that laying the groundwork for greater digital inclusion will pay dividends in the post-COVID-19 era in improving health and social equality. However, whilst these strategies are being introduced, it is essential that non-digital options, such as telephone services and staffed public access points, must remain available for those who are unable to engage with digital services.

Limitations. The sampling methodology employed by YouGov is both a strength and limitation of the study. The non-probabilistic method employed allowed for the prompt and cost-effective delivery of a prespecified sample size from segments of the population, who are traditionally difficult to engage in qualitative research. This method, however, precludes nonresponse bias calculations, and harbours a higher degree of bias than probabilistic sampling. Additionally, this cross-sectional survey provides a snapshot of people's preferences, rather than how sentiments evolve over time. Public trust in entities, such as government, varies over the course of a crisis, and could provide some explanation for the low government net trust rating (40%)⁵². The study data did not include comorbidities of respondents therefore exploration of this group, who are potentially vulnerable to COVID-19, could not be performed. Furthermore, the YouGov survey is also unlikely to have accessed proportionate numbers of marginalised people such as migrant workers, the homeless and sex-workers who are at risk of COVID-19, and have poor access to healthcare and digital interventions^{42,53,54}. In addition, as noted, those without internet access will also not have been able to participate in the study.

Conclusion

This study demonstrates an ongoing digital divide in the UK population with older, groups of lower social grade and educational attainment reporting less preparedness for COVID-19 digital health strategies. It highlights how a 'digital first' model of disseminating critical health information, disease surveillance and digital contact tracing have significant potential to marginalise population groups who are concurrently vulnerable to both digital exclusion and poor health outcomes secondary to SARS-CoV-2.

Given the importance of maintaining low transmission rates across all regions and population groups, there is an urgent need for key decision makers to consider further investment in multifaceted strategies to mitigate this possibility. Solutions should be targeted towards the principal drivers of digital exclusion; (1) access, (2) skills and (3) engagement. Through the empowerment of end-users, public health strategies will have a greater chance of containing disease spread and limiting the deepening of inequalities in health outcomes and the digital divide.

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Author contributions

V.S., A.A., S.R.M., H.A. and A.D. planned the study. V.S., J.C., A.A. and H.A. created the survey questions. S.Y. and J.C. conducted the statistical analysis. V.S., J.C. and S.Y. all contributed to the writing of the manuscript. S.R.M., H.A. and A.D. undertook the senior review of the work upon completion of the manuscript preparation.

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Competing interests

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Additional information

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