Background: The progression and geographical distribution of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) infection in the United Kingdom (UK) and elsewhere is unknown because typically only symptomatic individuals are diagnosed. We performed a serological study of blood donors in Scotland in the spring of 2020 to detect neutralising antibodies to SARS-CoV-2 as a marker of past infection and epidemic progression. Aim: Our objective was to determine if sera from blood bank donors can be used to track the emergence and progression of the SARS-CoV-2 epidemic. Methods: A pseudotyped SARS-CoV-2 virus microneutralisation assay was used to detect neutralising antibodies to SARS-CoV-2. The study comprised samples from 3,500 blood donors collected in Scotland between 17 March and 18 May 2020. Controls were collected from 100 donors in Scotland during 2019. Results: All samples collected on 17 March 2020 were negative in the pseudotyped SARS-CoV-2 virus microneutralisation assay. Neutralising antibodies were detected in six of 500 donors from 23 to 26 March. The number of samples containing neutralising antibodies did not significantly rise after 5–6 April until the end of the study on 18 May. We found that infections were concentrated in certain postcodes, indicating that outbreaks of infection were extremely localised. In contrast, other areas remained comparatively untouched by the epidemic. Conclusion: Although blood donors are not representative of the overall population, we demonstrated that serosurveys of blood banks can serve as a useful tool for tracking the emergence and progression of an epidemic such as the SARS-CoV-2 outbreak.

Introduction

Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) emerged in late 2019 in Hubei province, China, as a cause of respiratory disease occasionally leading to coronavirus disease (COVID-19) [1,2]. Older age, male sex, smoking and comorbidities such as...
In this study, we used blood donors as a means of estimating population exposure from the start of the pandemic in March through to mid-May when PCR-detected cases in the United Kingdom (UK) had plateaued [13,14]. The detection frequency of neutralising antibodies in blood donors and a discussion of its applicability for estimating population level exposure are presented.

**Methods**

**Samples**

We analysed six batches of 500 plasma samples collected on 17 March, 21–23 March, 5–6 April, 18–20 April, 2–4 May and 16–18 May from Scotland. Each batch was sampled from a range of health boards across Scotland, with the coverage varying between batches. An additional 500 samples from the Greater Glasgow region, collected between 2 and 4 May were also analysed. This yielded a total of 3,500 post-pandemic blood donor samples. Of these samples, 53.4% were from female donors. The median age of donors was 47 years (IQR: 34–56); children under 16 years are not permitted to donate blood. As negative controls, we tested in parallel 100 blood donor samples from the Scottish National Blood Transfusion Service (SNBTS) anonymous archive collected between September 2018 and December 2019 (IRAS project number 18005), before the first reports of the spread of SARS-CoV-2 in China [1,2]. Seventeen control samples from contract-traced individuals who were PCR-confirmed as SARS-CoV-2 infected were used as positive controls in the study. All the individuals from whom the positive control sera samples were taken had asymptomatic SARS-CoV-2 infections and were recruited through the International Severe Acute Respiratory and Emerging Infection Consortium (ISARIC) World Health Organization Clinical Characterisation Protocol UK (CCP-UK) at the time point of discharge plus 28 days. Samples were heat-inactivated before serological testing by incubation at 56°C for 30 min.

**SARS-CoV-2 pseudotype microneutralisation assay**

A lentivirus-based SARS-CoV-2 pseudovirus particle was constructed displaying the full spike protein on the surface of the pseudotyped virus using a synthetic codon-optimised SARS-CoV-2 expression construct (NCBI reference sequence: YP_009724390.1). Virus infectivity was determined by titration on HEK 293T ACE2-plasmid-transfected cells as previously described [15]. Neutralisation titres are expressed as 50% inhibitory concentration (IC$_{50}$) values. During the assay, plates were barcoded and controls were periodically added to the runs. Laboratory staff were blinded regarding the arrangement of positive controls periodically added to the assay plates.

**Titration**

Pre-pandemic samples and samples collected on 17 March and 21–23 March were all titrated to optimise the neutralisation assay. After this point, samples were initially screened for neutralisation using the highest 1:20 dilution. Dilutions of 1:20 were performed in triplicate along with virus only, no virus and positive control wells. Samples that produced a mean RLU two standard deviations below the mean RLU of all the samples on the plate were then titrated out to obtain IC$_{50}$ values.

**Enzyme-linked immunosorbent assay**

Antibodies to the trimeric spike protein were detected by ELISA. MAXISORP immunoplates (442404; NUNC; Merck, Darmstadt, Germany) were coated with StrepMAB-Classic (2–1507–001; IBA Life Sciences, Göttingen, Germany). Plates were blocked with 2% skimmed milk in phosphate buffered saline (PBS) for 1 h and then incubated with 0.125 μg of soluble SARS-CoV-2 trimeric spike protein or 2% skimmed milk in PBS. After 1 h, plasma was added at 1:50 dilution, followed by alkaline phosphatase (AP)-conjugated anti-human IgG (A9544; Merck, Darmstadt, Germany) at 1:1,000 dilution. This yielded a total of 3,617 samples.

**Figure 1**

Selection criteria for classifying a sample as SARS-CoV-2-neutralising, Scotland, March–May 2020 (n = 3,617)

<table>
<thead>
<tr>
<th>ELISA: enzyme-linked immunosorbent assay; IC$_{50}$: 50% inhibitory concentration; SARS-CoV-2: severe acute respiratory syndrome coronavirus 2.</th>
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<tbody>
<tr>
<td>Samples are required to have an estimated IC$_{50}$ and a standard error at least as small as the worst-neutralising positive control. This penalises samples with poorly defined inhibitory concentrations. Pre-pandemic samples are shown in yellow, positive controls in red and blood donor samples in blue. ELISA results are annotated as + or − on the points.</td>
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Cardiac disease, hypertension and diabetes have been identified as risk factors for severe infections [3,4]. Symptomatic individuals typically exhibit fever, cough and shortness of breath 2–14 days after infection [5]. However, an unknown proportion of individuals experience no symptoms [6–8]. Antibody responses in both symptomatic and asymptomatic individuals are detectable in the blood 14–28 days after infection [9,10]. Subsequently, antibody levels drop and can become undetectable by some antibody assays in the early convalescent phase [9,11,12].

In this study, we used blood donors as a means of estimating population exposure from the start of the pandemic in March through to mid-May when PCR-detected cases in the United Kingdom (UK) had plateaued [13,14]. The detection frequency of neutralising antibodies in blood donors and a discussion of its applicability for estimating population level exposure are presented.

**Figure 1**

Selection criteria for classifying a sample as SARS-CoV-2-neutralising, Scotland, March–May 2020 (n = 3,617)
1:10,000 dilution or AP-conjugated anti-human IgM (A9794; Merck, Darmstadt, Germany) at 1:5,000 dilution. The reaction was developed by the addition of p-nitrophenyl phosphate (PNPP, Merck, Darmstadt, Germany) substrate and stopped with NaOH. The absorbance was measured at 405 nm after 1 h. Further information is provided in Adams et al. [16].

**Figure 2**

SARS-CoV-2 antibody prevalence estimates for each health board through time using the model outlined in the methods, blood donors, Scotland, March–May 2020 (n = 3,500)

*Estimating the 50% inhibitory concentration*

The RLU for each well were standardised against technical positive (cells and virus without serum) and negative (cells without serum or virus) controls on each plate to determine a percentage neutralisation value. We calculated an average neutralisation across the two sample replicates on each plate (for each dilution). Dilution curves were fit to each sample, with the percentage neutralisation modelled as a logistic function of the dilution factor. This yielded an IC₅₀ value.
Dilution curves were estimated using nonlinear least squares in R version 3.6.3 [17]. An error-weighted mean of the IC₅₀ value was calculated for samples that were repeated on more than one plate. We classified positive samples as having an IC₅₀ value greater than the largest negative control (1:69) with a standard error less than or equal to the least neutralising positive control.

Determining test sensitivity and specificity

Test sensitivity (probability of neutralisation in a given positive serum) and specificity (probability of a negative result given no exposure) was estimated using 17 (RT-PCR-confirmed) positive controls and 100 pre-pandemic blood donor samples as negative controls. The highest IC₅₀ observed for a negative control was used as a threshold to determine positive samples (giving 100% specificity; 95% credible interval (CI): 98.10–100; n = 117). Of the 17 positive controls, 16 samples neutralised with high confidence, giving an estimated sensitivity of 94.11% (95% CI: 79.17–99.98).

Accounting for sensitivity and specificity in sample prevalence estimates

Uncertainty in test sensitivity and specificity can be propagated to sample prevalence estimates using a simple hierarchical Bayesian model [18]. The number of positive tests in the positive (n⁺ = 16) and negative (n⁻ = 0) control groups was modelled as a binomial distribution:

\[ n⁺ \sim Binom(\pi⁺, N⁺) \]
\[ n⁻ \sim Binom(1 - \pi⁻, N⁻) \]

where the sensitivity is given by \( \pi⁺ \) and the specificity by \( \pi⁻ \) (\( N⁺ = 17 \) and \( N⁻ = 100 \) are the number of positive and negative controls, respectively). An estimate of the true proportion of positive sera for samples from a given week and health board (\( p_{w,h} \)) comprises neutralising sera that were missed (\( 1 - \pi⁺ \)) and those incorrectly identified as neutralising samples (from \( 1 - \pi⁻ \)). The observed number of positive samples for the week \( w \) and health board \( h \) (\( n_{w,h} \)) was modelled as a binomial distribution accounting for test performance:

\[ n_{w,h} \sim Binom(p_{w,h} \pi⁺ + [1 - p_{w,h}][1 - \pi⁻], N_{w,h}) \]

with \( N_{w,h} \) the number of samples from each health board in each week. Using this method, the uncertainty in test specificity and sensitivity is propagated to the estimate of the seroprevalence; this results in broader credible intervals that better reflect the inherent uncertainty in test parameters.

Modelling sample prevalence

In estimating seroprevalence, we assumed that neutralising antibodies did not wane in the blood donor population during the survey period and accrued to an equilibrium [12]. Making this assumption, we can fit the logistic function to the time series of sample seroprevalence:

\[ P_{w,h} = \frac{\theta_h}{(1 + \exp(\phi \theta_h \eta_\theta))} \]
\[ \theta_h \sim N(\mu_\theta, \sigma_\theta^2) \]
\[ \phi \sim N(\mu_\phi, \sigma_\phi^2) \]
\[ \eta_\theta \sim N(\mu_\eta, \sigma_\eta^2) \]

Here, \( \theta_h \) is the equilibrium seroprevalence, \( p_{w,h} \) is the rate with which the seroprevalence approaches this maximum and \( \tau_h \) is the midpoint of the logistic curve for each health board. Parameters were modelled using hierarchical distributions across health boards (the maximum as a beta to bound it between 0 and 1, the
Figure 4
Health boards showing estimated endpoint seroprevalences for SARS-CoV-2 in blood donors, Scotland, March–May 2020 (n = 3,500)

White health boards were not sufficiently sampled to generate estimates. The colour scale denotes the confidence and seroprevalence in each health board. Labels give estimated blood donor seroprevalence for the health board with associated 95% highest density intervals.
rate and the midpoint as a normal distribution). Priors are given in the Supplementary material. The model was fit in R version 3.6.3 using the Bayesian inference package JAGS version 4.3.0 [19]. Models were run across six chains until convergence (potential scale reduction factor less than 1.02 and effective sample size > 10,000).

Ethical statement
Ethical approval was obtained for the SNBTS anonymous archive - IRAS project number 18005. SNBTS blood donors gave fully informed consent to virological testing, donation was made under the SNBTS Blood Establishment Authorisation and the study was approved by the SNBTS Research and Sample Governance Committee.

Results
The estimated IC_{50} values and standard errors for the control and blood donor samples are shown in Figure 1. Of the 3,500 post-pandemic blood donor samples, a total of 111 contained anti-SARS-CoV-2 neutralising antibodies using the IC_{50} and standard error-based thresholds described in the Methods. The results of the neutralisation assay were positively correlated with ELISA optical density (Supplementary Figure S2; Pearson’s correlation coefficient = 0.86; p < 0.001).

No samples from 17 March showed neutralising activity. Blood donor samples obtained from donations during 21–23 March, 5–6 April, 18–20 April, 2–4 May and 16–18 May contained neutralising anti-SARS-CoV-2 antibodies (Figure 2). The number of samples
significant variation in IC₅₀ value was seen based on age or sex (Supplementary Figure S4).

Central Glasgow had comparatively few samples containing neutralising antibodies (7/195; Figure 5). Analysis of the distribution of samples containing neutralising antibodies by postcodes showed that most of these samples located in the Paisley (14/85) and Motherwell (15/197) postcodes of Greater Glasgow region collected between 18 and 20 April. Of these 490 samples, 42 had neutralising antibodies. Analysis of the distribution of samples containing neutralising antibodies by postcodes showed that most of these samples located in the Paisley (14/85) and Motherwell (15/197) postcodes of Greater Glasgow, indicating that outbreaks in the city and its surrounding localities are localised. By comparison, central Glasgow had comparatively few samples containing neutralising antibodies (7/195; Figure 5).

Discussion

Our study strengthens existing evidence that blood donors can be used as a sentinel population to track the emergence and progression of an epidemic.

While the demographics of blood donors differ in several aspects from the general population, most notably because of the exclusion of those at risk for blood-borne viruses (HIV, hepatitis B and C virus) and syphilis, they might be considered a reasonable representation of the adult population in the absence of any obvious predisposing factors for infection. The only other general exclusions were a 4-week and a 4-month donation deferral period in those people who travelled to specified countries at risk for arbovirus and malaria infections, respectively.

However, estimates of seroprevalence are complicated by non-uniform sampling. The blood donations collected and tested in this study were focused in specific postcodes, based on the locations where weekly donations took place. This produces an added level of complexity as our data showed that outbreaks are focused in specific communities even on the scale of a medium-sized city such as Glasgow. This is further confounded by the absence of samples from individuals younger than 18 and older than 75 years.

The outbreaks as a whole in Scotland were concentrated in the major urban centres, Glasgow and Edinburgh, and the Lanarkshire health authority region (Figures 3 and 4). To explore this phenomenon further, we performed a separate analysis of 490 samples from the Greater Glasgow region collected between 18 and 20 April. Of these 490 samples, 42 had neutralising antibodies. Analysis of the distribution of samples containing neutralising antibodies by postcodes showed that most of these samples located in the Paisley (14/85) and Motherwell (15/197) postcodes of Greater Glasgow, indicating that outbreaks in the city and its surrounding localities are localised. By comparison, central Glasgow had comparatively few samples containing neutralising antibodies (7/195; Figure 5).

The results presented in this study are based on a formally not validated assay. However, by using contact-traced asymptomatic individuals who had been PCR-confirmed as infected by SARS-CoV-2 and 100 blood donations obtained before the epidemic, we were able to ascertain the sensitivity of the assay. Furthermore, a second ELISA was used to confirm the analysis. As this assay detected 16 of 17 PCR-confirmed asymptomatic cases, we estimated its sensitivity at 94.11% (95% CI: 79.17–99.98). Other studies have previously shown that the pMN assay correlates well with other laboratory-based and commercial serological assays [20].

Our assay is designed to be specific for SARS-CoV-2. There are four seasonal coronaviruses, HKU1, OC43, NL63 and 229E, which circulate during the winter months [21]. The 100 pre-pandemic samples collected in the winter months of 2019 will have been from donors previously infected with seasonal coronaviruses (but not SARS-CoV-2). By setting the cut-off for our assay above the highest IC₅₀ value observed in the 100 pre-pandemic samples, we ensured the specificity of the assay and can have a high degree of confidence that the antibodies detected in the samples from March to May 2020 were generated by SARS-CoV-2 infection. The utility of using pMN assays and ELISA to track population exposure is dependent on the assumptions (i) that every infected individual seroconverts and (ii) that once seroconverted, the antibodies remain circulating in the blood at detectable levels. A decrease in total antibody and neutralising antibody titres has been noted in samples drawn up to 2 months after the peak neutralising antibody response (ca 3–4 weeks after infection). In some instances, antibody levels become undetectable when tested with a specific assay and analysis methodology [9,12]. This drop in titres may lead to false negatives in the later time points. However, the dates of collection used in this study all fell within 3 months of the diagnosis of the first confirmed case in Scotland on 1 March [22]. For this reason, it is unlikely that this study is hampered by a drop in neutralising antibody levels, although future seroprevalence studies may potentially underestimate the true level of population exposure. In addition, some individuals may not seroconvert, representing a small pool of false negative patients [11].

Conclusion

Samples containing anti-SARS-CoV-2 neutralising antibodies were detected in blood donors who gave blood between 16 and 17 March 2020 in all Scottish health boards. Subsequently, samples containing anti-SARS-CoV-2 neutralising antibodies were detected at every further time point assayed until the end of the study. Consequently, considering the 14–28 day incubation period before seroconversion, it is likely...
that SARS-CoV-2 began circulating in Scotland in late February 2020 and potentially earlier.

ISARIC4C Investigators


Acknowledgements

We would like to acknowledge the help and collaboration of many SNBTS staff for provision and preparation of samples (anonymous archive and recent donation samples). We acknowledge the wider support of ISARIC4C.

Funding: This work was supported by the Georg and Emily Opel Foundation. This work was supported by the Medical Research Council [grant number MC_PC_19059]. National Institute for Health Research Biomedical Research Centre Funding Scheme (to G.R.S.), the Chinese Academy of Medical Sciences (CAMS) Innovation Fund for Medical Science (CIFMS), China (grant number: 2018-1ZM-2-002) and the Schmidt Foundation. PK, P5 and G.R.S. is supported as a Wellcome Trust Senior Investigator (grant 095545/A/11/Z; WT109965/MA). PK is an NIHR Senior Investigator. NG was supported via grant to Philip Goulder (WT104748MA) and a grant to John Frater (Medical Research Council MR/L006588/1). CPT was funded by an ERC research grant ‘UNIFLUVAC’ and two MRC CIC grants (Ref: RB00140). LNL is supported by a CRUK Cancer Immunology Award (C30332/A23521) to PK. The funders played no role in the design, execution or reporting of the study. RP also supported by funds provided under Professor RW Snow’s Wellcome Trust Principal Fellowship (# 211276). ALM is funded by a NIHR Research Capability Funding grant.

MG Semple, P Klierman, and P Simmons are affiliated to the National Institute for Health Research Health Protection Research Unit (NIHR HPRU) in Emerging and Zoonotic Infections at University of Liverpool in partnership with Public Health England (PHE), in collaboration with Liverpool School of Tropical Medicine and the University of Oxford [award number NIHR200907]. The work was also supported by the NIHR Biomedical Research Centre, Oxford. The views expressed are those of the author(s) and not necessarily those of the MRC, NHS, the NIHR, the Department of Health or Public Health England.
Conflict of interests

None declared.

Authors’ contributions

Thompson, C., Chinnakannan, S., Dejnirattisai, W., Edmans, M., Fye, A., Kooball, K., Lee, L., Lim, N., Liu, C., López-Camacho, C., McNaughton, A.L., Mongkolsapaya, J., Odon, V., Sampson, O., Ramamurthy, N., Ratcliff, J., Supasa, P., Wang, B. and Mentzer, A., performed the sample acquisition, laboratory testing and reporting of the pseudovirus and ELISA testing. Imlach C., McNally C., Harvala, H. and Jarvis, L.M. The ISARIC4C investigators established the sample sets, archiving and data provision of the samples used in the study. Grayson, N., Lorenou, J., Penman, B.S., Semple, M.G., Baillie JK, Bolton, J. and Paton, R.S. performed the data analysis and results interpretation. Paton, R.S, Gupta., Thompson, C., Lorenou, J., wrote and interpreted the sero-prevalence model. Turner, M., Thompson, C., Temperton, N., Gupta, S., Klenerman, P., Scretan, G.R. and Simmonds, P conceived and designed the specifics of the study, the data interpretation and drafting of the manuscript. All co-authors contributed to the editing and final drafting of the manuscript and figures.

References


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