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## Serological surveys to estimate cumulative incidence of SARS-CoV-2 infection in adults (Sero-MAss study), Massachusetts, July-August 2020- a mail-based crosssectional study

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# Serological surveys to estimate cumulative incidence of SARS-CoV-2 infection in adults (Sero-MAss study), Massachusetts, July-August 2020: a mail-based cross-sectional study 

Teah Snyder, ${ }^{1 \dagger}$ Johanna Ravenhurst, ${ }^{1 \dagger}$ Estee Y. Cramer, ${ }^{1 \dagger}$ Nicholas G. Reich, ${ }^{1}$ Laura B. Balzer, ${ }^{1}$ Dominique Alfandari, ${ }^{2}$ and Andrew A. Lover ${ }^{1 *}$<br>$\dagger$ Equal contributions.<br>${ }^{1}$ Department of Biostatistics and Epidemiology, School of Public Health and Health Sciences, University of Massachusetts- Amherst, Amherst MA<br>${ }^{2}$ Department of Veterinary and Animal Sciences, College of Natural Sciences, University of Massachusetts- Amherst, Amherst MA<br>* Correspondence to: alover@umass.edu<br>School of Public Health and Health Sciences<br>University of Massachusetts- Amherst<br>715 North Pleasant Street<br>Amherst, MA 01003-9304<br>phone: 413-545-7426


#### Abstract

Objectives: To estimate the seroprevalence of anti-SARS-CoV-2 $\operatorname{IgG}$ and IgM among Massachusetts residents and to better understand asymptomatic SARS-CoV-2 transmission during the summer of 2020.

Design: Mail-based cross-sectional study Setting: Massachusetts, United States Participants: Primary sampling group: sample of undergraduate students at the University of Massachusetts, Amherst $(\mathrm{n}=548)$ and a member of their household $(\mathrm{n}=231)$. Secondary sampling group: sample of graduate students, faculty, librarians and staff ( $\mathrm{n}=214$ ) and one member of their household ( $\mathrm{n}=78$ ). All participants were MA residents without prior COVID-19 diagnosis.

Primary and secondary outcome measures: Prevalence of SARS-CoV-2 seropositivity. Association of seroprevalence with variables including age, gender, race, geographic region, occupation, and symptoms. Results: Approximately 27,000 persons were invited via email to assess eligibility. 1,001 households were mailed dried blood spot sample kits, 762 returned blood samples for analysis. In the primary sample group, 36 (4.6\%) had IgG antibodies detected for an estimated weighed prevalence in this population of $5.3 \%$ ( $95 \%$ CI: 3.5 to 8.0 ). In the secondary sampling group, 10 (3.4\%) had IgG antibodies detected for an estimated adjusted prevalence of $4.0 \%(95 \% \mathrm{CI}: 2.2$ to 7.4). No samples were $\operatorname{IgM}$ positive. No association was found in either group between seropositivity and self-reported work duties or customer-facing hours. In the primary sampling group, self-reported febrile illness since Feb 2020, male sex, and minority race (Black or American Indian/Alaskan Native) were associated with seropositivity. No factors except geographic regions within the state were associated with evidence of prior SARS-CoV-2 infection in the secondary sampling group.

Conclusions: This study fills a critical gap in estimating the levels of sub-clinical and asymptomatic infection. Estimates can be used to calibrate models estimating levels of population immunity over time, and these data are critical for informing public health interventions and policy.


## Strengths and limitations of this study

- Our study collected serological samples in a well-defined rigorous sample frame in a contact-less (mail-based) survey in an early stage of the pandemic in an area of high SARS-CoV-2 burden.
- A range of potentially associated demographic, occupational, and behavioral factors were surveyed to contextualize seropositivity across geographic regions within the state.
- Our study sampled from populations affiliated with a large public university in Massachusetts, and may not be generalizable to the general population


## Introduction

Since emergence in early 2020, the SARS-CoV-2 virus has severely impacted the entire globe. The state of Massachusetts was heavily impacted in the earliest stages of the pandemic, and a "super-spreader" event in the state in April 2020 may have seeded large case clusters throughout the country. ${ }^{1}$ However, the trajectory of the early stages of transmission in this state, as well as across the US remain poorly understood due to changes in case definitions and limited testing of both symptomatic and asymptomatic persons during the summer of 2020. ${ }^{2}$ To assess seroprevalence across the state, a mail-based serosurvey was implemented July-August 2020. At the time of this survey, the Massachusetts Department of Public Health had reported over 109,143 confirmed COVID-19 cases and over 8,081 deaths. ${ }^{3}$ Seroepidemiological studies are a critical tool to explore infection dynamics, especially where asymptomatic or subclinical infections are common, as for SARS-CoV-2. ${ }^{4}$ This study helps to fill a critical gap in estimating the levels of sub-clinical and asymptomatic infection to inform consequent levels of populationimmunity. ${ }^{5}$

Concurrent to this study, a number of seroprevalence studies were conducted on the east coast of the United States; these studies focused on specific populations at high risk and found varying results. A survey in April 2020 in a convenience sample of 200 asymptomatic residents of Chelsea, MA found an estimated of seroprevalence of $31.5 \%(17.5 \% \mathrm{IgM}+/ \mathrm{IgG}+, 9.0 \%$ $\mathrm{IgM}+/ \mathrm{IgG}-$ and $5.0 \% \mathrm{IgM}-/ \mathrm{IgG}+) .{ }^{6}$ This study used a small convenience sample and did not include any randomization. ${ }^{6}$ A study with a larger sample of over 28,000 clinical patient samples in New York City, USA found an IgG seropositivity prevalence of $44 \%$ with over $50 \%$ of participants reporting no symptoms. ${ }^{7}$

Other seroprevalence surveys across the US have found generally low- to moderate prevalence in a diverse set of study populations. A study of 790 university students in Los Angeles, California conducted in April and May of 2020 estimated a prevalence of SARS-CoV-2 IgG antibody of 4.0\% ( $95 \%$ CI: 3.0 to $5.1 \%) .{ }^{8}$ During May - April of 2020, a cross-sectional study in St. Louis found IgG seropositivity to be estimated at $1.71 \%$ ( $95 \% \mathrm{CI} 0.04 \%$ to $3.38 \%$ ) in pediatric patients and $3.11 \%$ ( $95 \%$ CI: $0.92 \%$ to $5.32 \%$ ) in adult patients. In the most comprehensive serosurvey from the spring and summer months of 2020, 16,025 clinical samples were analyzed with $\operatorname{IgG}$
spike protein sero-reactivity ranging from $1.0 \%$ in the San Francisco Bay Area to $6.9 \%$ in New York City. ${ }^{9}$ These disparate results highlight major geographic variability in the trajectory of infections, and reinforce the need for additional seroprevalence studies to more fully contextualize trends in immunity to SARS-CoV-2 targeting specific geographic regions.

Though community seroprevalence studies generally rely on serum samples collected in health facilities, the use of dried blood spot (DBS) samples is a practical and effective alternative. ${ }^{10}$ DBS samples involve a small finger-prick sample self-collected by participants in their own homes. The use of dried blood samples for antibody assays has been validated in other work prior to the current pandemic, ${ }^{10,11}$ and previous studies have evaluated the feasibility, validity, and acceptability of using DBS samples for SARS-CoV-2 antibody testing. ${ }^{12-15}$ This method of sample collection facilitates efficient population-level sampling while minimizing social mixing and concurrent potential exposures.

This study estimated the prevalence of previous infection with SARS-CoV-2 in individuals who had not been diagnosed with COVID-19 and were asymptomatic with representative coverage across the entire state of Massachusetts, USA. Information from this study can provide knowledge regarding the seropositivity of this population and can be used to inform decisionmaking regarding community re-openings during the pandemic.

## Methods

## Study Design and Participants

The study population included undergraduate students, graduate students, staff, and faculty members currently affiliated with the University of Massachusetts, Amherst and their household members. On-campus classes were suspended in mid-March 2020; consequently, undergraduates had exposure to the local epidemiology within their communities from March until sampling in July-August throughout the state (primary sampling group). Conversely, graduate students, faculty, staff, librarians and their family members (secondary sampling group) almost universally reside in close proximity to Amherst, and broadly reflect transmission in the Western part of the state. UMass affiliates were eligible to participate in this study if they were above the age of 18, had been living in Massachusetts for the past eight weeks; had never received a COVID-19
diagnosis from a medical professional; and did not have a fever greater than $100.4^{\circ} \mathrm{F}$ at the time of survey completion. Household members were eligible for inclusion if they met all of these same criteria and were between the ages of 23 and 78 (chosen to expand sampling beyond college-age population groups).

An institutional email list was provided by university administration for recruitment. Initial emails were sent out to UMass affiliates between June 23, 2020 and June 26, 2020 for participant recruitment. The email provided information about the study and links to a screening eligibility survey, informed consent document, initial survey regarding COVID-19 risk factors, and information regarding shipping addresses. If the UMass affiliate had a household member interested in participating, a single household member was invited to complete an eligibility, consent, and initial survey forms. To increase participation rates, two reminder emails were sent to all non-respondents (day three and six after initial solicitation). All survey responses were collected and stored in REDCap. ${ }^{16}$

The survey was closed after a three-week enrollment period, and a subset of participants were selected to receive a test kit. To select a population representative of the broader UMass community across the entire state of Massachusetts, two sampling schemes were applied. The first consisted of all undergraduates and their household members (primary sampling group); the second sampling frame consisted of graduate students, staff, faculty members, librarians, and their household members (secondary sampling group). Within the primary sampling group, selection for biosample collection used probability proportional to population size, using the most recent census data aggregated to state-level emergency response regions due to sparse county-level populations (Figure 1). ${ }^{17}$ For the secondary sampling group, selection for biosampling was via simple random sampling.

The full sample frame selection is shown in Figure 2. Briefly, an email invitation was sent to a total of 27,339 individuals, of which 4,124 completed the screening, informed consent, and initial survey forms. A total of 1,001 individuals were then randomly selected to receive a sampling kit. Participants were mailed all materials to safely collect and return samples,
including lancets, alcohol wipes, gauze, gloves, bandages, a bloodspot collection card, a pre-paid shipping box, and detailed printed instructions cards (including a nurse call line).
Upon mailing out the test kits, participants were also emailed a link containing a video on how to collect the DBS, along with a detailed survey form with demographics, risk factors, and any current symptoms or COVID diagnoses. No participants reported a COVID-19 diagnosis between the initial survey, and sample collection several weeks later. All shipments utilized a Biological Substance Category B (UN3373) shipping box.

## Ethical approvals

This study was approved by the University of Massachusetts-Amherst Human Research Protection Office (Approval \#2062; April 27, 2020).

## Sample Preparation and ELISA Analysis

Upon receipt of boxes, the sample cards (Whatman ${ }^{\circledR}$ Protein Saver 903) were heat-treated (30 minutes at $56^{\circ} \mathrm{C}$ ); a single blood spot per card was punched ( 0.25 -inch diameter) and transferred to an ELISA plate. Plates were coated with $1 \mu \mathrm{~g} / \mathrm{ml}$ of purified RBD diluted in PBS overnight at 4C and blocked with Tris-Buffered Saline with $0.1 \%$ Tween 20 (TBST) containing 5\% non-fat dry milk. DBS were eluted in $500 \mu 1$ of TBST overnight at 4 C and $50 \mu 1$ of each sample was added to the ELISA plate preloaded with $50 \mu 1$ of TBST containing $2 \%$ non-fat dry milk. Samples were then assayed for SARS-CoV-2 antibodies according to published protocols. ${ }^{18,19}$ The RBD protein was produced in-house via transfection of HEK293T cells using polyethylenimine (Plasmid was a generous gift from Pr. F. Kramer mount Sinai School of Medicine). Batches were control for purity by SDS-PAGE followed by Coomassie staining and ELISA using an anti His-tag monoclonal antibody. Optical densities were read at 405 nm , and each 96-well plate contained seven negative controls and one positive control (serum from PCRconfirmed case at $1 / 100$ dilution). Samples were tested against IgG and positive samples were confirmed and tested with anti IgM antibodies. Optical density values were normalized to the mean optical density of negative controls daily.

## Data Analysis

Sample size and power

The study was designed to assess seropositivity within the primary sampling group with sufficient precision to inform policy. With 750 persons, and an assumed $5 \%$ positivity, the $95 \%$ CI for this estimate is $3.6 \%$ to $6.9 \%$. Within the five emergency response subregions, at $5 \%$ seropositivity, the survey is powered for a precision of $2.3 \%$ to $10.2 \%$. The secondary sampling group ( $\mathrm{n}=250$ ) sample size was based on logistic limitations, but was powered to a precision of $2.8 \%$ to $8.8 \%$. All confidence intervals are binomial exact, without adjustments for study design effects or non-response.

## Analysis of serology data

Finite mixture models were used to determine seropositivity cutoffs. These latent-class models estimate "breakpoints" for seropositive and seronegative subpopulations, and have been applied to a range of pathogen serosurvey data, including rubella, pertussis, and parvovirus. ${ }^{20-22}$ From this analysis, all samples with IgG optical density ratio $\geq 2.49$-fold above daily background were considered positive for SARS-CoV-2.

## Adjusted estimates

All reported prevalences and prevalence ratio estimates are adjusted with non-response weights, which were estimated using inverse weighting. Briefly, logistic regression models were used to calculate propensity scores for each individual in the sample using reported gender and race categories. These were transformed to probabilities; a small number of individuals had extremely large weights due to sparse strata; these weights were truncated at $1 / 0.02 .{ }^{23}$ Weights were then used for all prevalence and prevalence ratio estimations using the survey package in R. ${ }^{24}$ The primary sampling group sample was self-weighting due to probability-proportional to population size sampling. Sampling weights were not used in the secondary sampling group as selection used simple random sampling.

## Multivariable analyses for prevalence ratios

Prevalence ratios were estimated to assess factors associated with seropositivity, with separate Poisson models ${ }^{25}$ for both of the two sampling groups, with robust (sandwich) errors to address clustering within households.

Bivariate analyses were performed for each factor separately. All variables with a p-value $<0.20$ based on bivariate association with outcome were further evaluated for inclusion in final models. All final models were adjusted for age (continuous), race, and gender (see Table 1). Due to several very sparse categories, some were combined in final models. Specifically, all race/ethnicity categories and all geographic regions were not included in analysis of the secondary sampling group due to unstable estimates.

Model parsimony was evaluated using AIC/BIC and all tests were two-tailed, with $\alpha=0.05$. R version 4.0.3 and SAS version 9.4 (SAS Institute, Inc, Cary, NC, USA) were used for analysis.

## Patient and public involvement

All members of the university community were invited to participate, and serological testing was proved at no cost to the sampled individuals and their selected household contact.

## Results

A total of 1,001 individuals were enrolled into the study; this included 752 undergraduate students, 90 graduate students, 63 faculty/librarians, and 96 staff members (Figure 2). Seventysix percent of these $(\mathrm{n}=762)$ returned blood samples for analysis; 548 in the primary sampling group, and 214 in the secondary sampling group. Of the 548 participants in the primary sampling group, 230 enrolled a household member. One household member submitted a sample without the sample of the main participant, bringing the total number of undergraduate household members to 231 . Of the 214 participants in the secondary sampling group, 78 enrolled a household member. Two returned samples were excluded from analysis due to unlinkable samples. A total of 1,071 samples were included in the final analyses: 762 main participants and 309 household members (Figure 2).

Demographic characteristics of both sampling groups are presented in Table 1. Race categories do not total to $100 \%$ due to non-response and multiple possible answers. Age, gender, and essential worker status were broadly similar between those invited to participate and those who completed the study (Supplemental Table 1).

Of the total 1,071 samples tested, 46 were positive for SARS-CoV-2 antibodies. Demographic results are stratified by IgG serostatus (Table 2); no samples showed evidence for IgM positivity. Seropositivity was low-to-moderate across the survey groups, with several important exceptions. Variation is apparent by sex, race, and across geographic regions; however, several strata have wide confidence intervals due to small sample sizes.

Of the 779 primary sampling group participants and their household members, 36 were positive for SARS-CoV-2 antibodies. This corresponds to an overall seroprevalence of $5.3 \%(95 \% \mathrm{CI}$ : $3.1-7.5$ ) of the population after adjustment for nonresponse and geographic location. In the secondary sampling group, of the 292 graduate students, staff, librarians, faculty members, and their household members, ten (adjusted 4.0 \%, 95\% CI: 1.6-6.5) had evidence for prior SARS-CoV-2 infection. (Table 2). Results were also further stratified by UMass affiliate vs. household member. Of the 548 undergraduate students in the primary sampling group, 27 were positive for SARS-CoV-2 IgG antibodies (population positivity rate of 5.3\% (95\% CI: $3.1-7.6 \%$ ). Of the 231 household members of undergraduate participants, nine (adjusted $5.2 \%, 95 \% \mathrm{CI}: 1.2-9.2$ ) were positive for SARS-CoV-2 IgG antibodies. In the secondary sampling group, eight University affiliates (adjusted $4.3 \%$, $95 \%$ CI: 1.3-7.3) were positive for SARS-CoV-2 IgG antibodies. Of the household members in the secondary sampling group, 2 were seropositive, with a weighted seroprevalence of ( $3.3 \%, 95 \%$ CI: $0.0-7.8 \%$ ). (Table 2). The overall distributions of measured IgG lognormal optical density ratios by subgroups are broadly similar (Supplemental Figure 1).

After adjustments for age, gender, region, and self-reported febrile illness since February 2020, the factor with the strongest association with seropositivity in the primary sampling group was Black or AIAN Race ( $\mathrm{PR}=4.49,95 \% \mathrm{CI}=1.57,12.9$ ) (Table 3). This indicates that individuals who report being Black or American Indian / Alaskan Native have a prevalence 3.49 times higher than White individuals after adjustment. Additionally, after adjustments, females and those who are gender diverse were at a significantly lower risk of prior SARS-CoV-2 infection $(\mathrm{PR}=0.5 ; 95 \% \mathrm{CI}=0.27,0.92)$ compared to males. Those who reported a febrile illness in February were more likely to be seropositive than those who were not sick ( $\mathrm{PR}=2.42,95 \% \mathrm{CI}=$ $1.24,4.75)$. No significant associations were found across each of the 5 regions in the primary
sampling group, however the prevalence of seropositivity was $48 \%$ higher in Region 1 compared to Region $4(\mathrm{PR}=1.48,95 \% \mathrm{CI}=0.62,3.52)$.

Within the secondary sampling population, after adjustments for age, race, gender, region, household member and self-reported febrile illness since February (Table 4), participants who reported residing in Region 2, 3, 4, or 5 had greater than 4 times higher prevalence of SARS-CoV-2 antibodies as compared to those who resided in Region $1(\mathrm{PR}=4.08,95 \% \mathrm{CI}=1.09$, 15.33). No other factors included in the model were significantly associated with seropositivity.

## Discussion

This mail-based serosurvey of two university-affiliated populations across Massachusetts in July and August 2020 found an estimated seroprevalence of $\sim 5 \%$ of antibodies to SARS-CoV-2. These results indicate that even with extensive morbidity and mortality across the state at the time of sampling, there had been limited exposure to SARS-CoV-2 at a population-level. This estimated seroprevalence is lower than that detected with concurrent community-based studies in other states. An estimated $14.3 \%$ of the United States population had been previously infected with SARS-CoV-2 by November 2020, as estimated in a pooled analysis of multiple seroprevalence surveys. ${ }^{26}$

Our estimates are substantially lower than some models of COVID-19 seroprevalence in Massachusetts. One model estimates a seroprevalence of $16.2 \%$ (no CIs provided) on Jul 27, 2020 (closest modeled date). These estimates are nearly double our measured seroprevalence with inclusion of 110,000 confirmed cases at that date (ca. $1.5 \%)^{27}$. These differences might be caused by a number of reasons, including a non-representative population by age or geographic range, or waning of antibody titers. Without CIs, we are unable to evaluate coverage outside the reported point estimate. However, alternate nowcasting estimates suggest a total statewide attack rate on July 31, 2020 of $6.9 \%$ ( $95 \%$ CrI: $5.5-8.4 \%$ ) in Massachusetts ${ }^{28}$. Our results are closely aligned with these estimates.

Within the surveyed groups, approximately $24 \%$ of the primary sampling group and $18 \%$ of the secondary sampling group reported illness since February. Febrile illness since February was
associated with an increased prevalence of seropositivity in both sampling groups, but after multivariable analysis this association was found in only the primary sampling group. These results reinforce results from other studies: asymptomatic illness is an important contributor to observed force of infection; and important limitations of testing availability at the time of survey.

Differing antibody dynamics have been reported in other studies. A number of studies have found sustained antibody levels for over 3 months, ${ }^{29,30}$ while others have found IgG levels can remain 6 months or more. ${ }^{31-33}$ An additional study has reported rapid waning of routine serological markers in individuals who had lower initial antibody responses. ${ }^{34}$. Only $7.1 \%$ of those with high titers at baseline seroreverted to a level below the threshold for positivity within 60 days, compared to $64.9 \%$ of those with lower titers at baseline. ${ }^{34}$ Evidence for IgM seropositivity was not detected in any of IgG positive samples, which is consistent with results from other surveys studies that included asymptomatic or subclinical populations due to rapidly waning titers. ${ }^{32,35}$ Studies have shown that IgM levels decline more rapidly after infection than IgA and IgG levels, ${ }^{30,36,37}$, and this is especially apparent with asymptomatic and sub-clinical infections. ${ }^{32,35}$

Trends in the patterns of SARS-CoV-2 antibody levels vary greatly depending on the timing of sampling and severity of disease. ${ }^{31,32,35}$ Seroconversion times vary depending on the study, but one study found a median time-to-seroconversion for $\operatorname{IgM}$ of 8 days and median seroconversion for IgG of 10 days. Additionally, the SARS-CoV-2 IgG response generally begin around 10-15 days after symptom onset. ${ }^{2}$ For this reason, repeated serial sampling, repeated serial sampling of convalescent populations should be prioritized to more fully understand the dynamics of immune response.

SARS-CoV-2 seropositivity was associated with minority race status in this survey. While the total number of non-white participants was limited, the large effect size reinforces other studies suggesting that marginalized communities have been and continue to be excessively impacted by the pandemic. Results from the primary sampling group analysis suggest that self-reported Black race is a risk factor for previous SARS-CoV-2 infection, which is consistent with findings from other studies. ${ }^{38-40}$ A number of factors may play a role in this significant association
including the fact that Black individuals are more likely to work in frontline industries or live in areas with a higher population density. ${ }^{41}$ No parallel associations were found in the analysis of the secondary sampling group due to limited sample size in some strata. In the secondary sampling group, the aggregation of Race categories into White race and Non-white race likely obscured meaningful associations between Race and exposure.

Results from the primary sampling group showed higher prevalence of IgG seropositivity among males. After adjusting for age, race, and region, male gender was a significant risk factor for evidence of SARS-CoV-2 infection. Other studies have similarly found that males have higher rates of infection than females for asymptomatic infections ${ }^{4243}$ These findings may reflect differences in care seeking behavior (biased recruitment), true biological differences, or differences in health behaviors such as smoking, alcohol use, and COVID-19 prevention measures. ${ }^{44}$ This association was not observed in the secondary sampling group.

The primary sampling group also showed increased risk of seropositivity with self-reported illness since February 2020; this association was not observed in analyses of the secondary sampling group. This finding may indicate that some of the participants in our study were not strictly asymptomatic and were simply unable to obtain a COVID-19 test due to limited availability during the beginning of the pandemic.

This study was population-based and had broad eligibility criteria but is subject to several limitations. The exclusion of persons with confirmed diagnoses and any current symptoms (due to biosafety concerns) also inherently limited capture of sub-clinical infections. As such, the estimates are likely a lower bound. However, participants who suspected they may have been previously infected with SARS-CoV-2 might be more likely to participate compared to those that were not concerned with prior infection. This is a pervasive issue in community-based studies, where characteristics of those who volunteer to participate in community-based research differ from the general population. ${ }^{45}$

Randomization after a three-week enrollment period helped to address this limitation, as using only the first participants to volunteer could have biased the sample to include those who were most motivated to receive their antibody test results. If participants were more motivated to
receive their results because they thought they suspected a prior exposure to SARS-CoV2, this would have inflated the observed prevalence of seropositivity in the study population.

Another limitation of the study is the self-reported response of the lack of a prior COVID-19 diagnosis and current fever. It is possible that some participants shielded their answer and submitted samples for analysis without meeting the eligibility criteria; this would have inflated our estimation of seroprevalence in asymptomatic groups, Thirdly, the limited number of nonWhite, and gender-diverse participants also limited some analyses. Fourthly, while multiple studies have validated DBS sampling for SARS-CoV-2, ${ }^{46}$ waning antibodies in asymptomatic individuals could be below the limited of detection of the ELISA assay. Finally, generalizability is limited due to the recruitment of a university-affiliated population in a relatively restricted geographic area. ${ }^{8}$

This serosurvey estimates prevalence of prior SARS-CoV-2 infections in a university-affiliated population in Massachusetts. Risk factors for IgG seropositivity included recent self-reported febrile illness, minority race status, and male gender. This study provides estimates of seroprevalence in Massachusetts after a 'first wave' of SARS-CoV-2 infections in the spring of 2020. Repeat seroprevalence studies in this population could provide estimates in changes of seropositivity rates given subsequent waves of SARS-CoV-2 transmission. This study reinforces the critical need for targeted serosurveys in highest-risk and marginalized communities, both in Massachusetts, and nationwide.

## Tables

Table 1. Demographics of study populations, SARS-CoV-2 serology surveys in universityaffiliated populations, Massachusetts Jul-Aug, 2020.

| Characteristic | Primary Sampling Group ( $\mathrm{N}=779$ ) | Secondary Sampling Group ( $\mathrm{N}=292$ ) |
| :---: | :---: | :---: |
| Gender <br> Female <br> Male <br> Gender Diverse <br> Missing | $\begin{gathered} 499(64.1 \%) \\ 274(35.2 \%) \\ 5(0.6 \%) \\ 1(0.1 \%) \\ \hline \end{gathered}$ | $\begin{gathered} 154(52.7 \%) \\ 136(46.6 \%) \\ 2(0.7 \%) \\ 0(0.0 \%) \\ \hline \end{gathered}$ |
| Race <br> AIAN <br> Asian <br> Black <br> Hispanic <br> Multiple <br> White <br> Missing | $17(2.2 \%)$ $78(10.0 \%)$ $12(1.5 \%)$ $36(4.6 \%)$ $37(4.8 \%)$ $545(70.0 \%)$ $54(6.9 \%)$ | $1(0.3 \%)$ $34(11.6 \%)$ $3(1.0 \%)$ $9(3.1 \%)$ $11(3.8 \%)$ $217(74.3 \%)$ $17(5.8 \%)$ |
| Age <br> Mean <br> Median Range | $\begin{gathered} 29.9 \\ 21 \\ 18-75 \end{gathered}$ | $\begin{gathered} 41.6 \\ 39 \\ 21-75 \\ \hline \end{gathered}$ |
| Education <br> HS/GED <br> Some College <br> BA/BS <br> More Than BA/BS <br> Missing | $\begin{gathered} 102(13.1 \%) \\ 483(62.0 \%) \\ 117(15.0 \%) \\ 74(9.5 \%) \\ 3(0.4 \%) \end{gathered}$ | $\begin{gathered} 5(1.7 \%) \\ 24(8.2 \%) \\ 78(26.7 \%) \\ 183(62.7 \%) \\ 2(0.7 \%) \end{gathered}$ |
| Essential Worker | $\begin{gathered} 533 \text { (68.4\%) } \\ 195(25.0 \%) \\ 51(6.6 \%) \\ \hline \end{gathered}$ | $\begin{gathered} 224 \text { (76.7\%) } \\ 51(17.5 \%) \\ 17(5.8 \%) \\ \hline \end{gathered}$ |
| Self-reported attitude about COVID-19 <br> Strongest fear Somewhat fearful Neutral/Missing <br> Somewhat not fearful Not fearful | $\begin{gathered} 135(17.3 \%) \\ 389(49.9 \%) \\ 139(17.8 \%) \\ 86(11.0 \%) \\ 30(3.9 \%) \\ \hline \end{gathered}$ | $\begin{gathered} 72(24.7 \%) \\ 122(41.8 \%) \\ 63(21.6 \%) \\ 23(7.9 \%) \\ 12(4.1 \%) \\ \hline \end{gathered}$ |
| Self-reported febrile illness since February | $\begin{gathered} \text { No: } 534(68.6 \%) \\ \text { Yes: } 188(24.1 \%) \\ \text { Missing: } 57(7.3 \%) \\ \hline \end{gathered}$ | $\begin{gathered} \text { No: } 224(76.7 \%) \\ \text { Yes: } 53(18.2 \%) \\ \text { Missing: } 15(5.1 \%) \\ \hline \end{gathered}$ |
| Self-reported care | No: 112 (59.6\%) | No: 32 (60.4\%) |


| Seeking (if illness since | Yes: $75(39.9 \%)$ | Yes: $21(39.6 \%)$ |
| :--- | :---: | :---: |
| Feb) | Missing: $1(0.5 \%)$ | Missing: $0(0.0 \%)$ |

*Notes: AIAN= American Indian/Alaska Native. The primary sampling group includes UMass undergraduates and household members, and the secondary sampling group includes UMass affiliated faculty, staff, and graduate students and household members.

Table 2. Weighted seropositivity by main demographic variables, SARS-CoV-2 serology surveys in university-affiliated populations, Massachusetts Jul-Aug, 2020.

| Characteristic | Primary Sampling Group | Secondary Sampling Group |
| :---: | :---: | :---: |
| Age in years, median | $\begin{gathered} 21 \\ 95 \% \mathrm{CI}(20-21) \end{gathered}$ | $\begin{gathered} 41 \\ 95 \% \mathrm{CI}(38-44) \end{gathered}$ |
| Prevalence of SARS-CoV-2 Antibodies, by sub-group |  |  |
| Overall Population Prevalence (95\% CI) | 5.3\% (3.5-8.0) | 4.0\% (2.2-7.4) |
| Sex \%, (95\% CI) Female <br>  Male <br> Gender Diverse/No Response  | $\begin{gathered} 4.0 \%(2.4-6.6) \\ 8.7 \%(5.1-15.0) \\ 0.0 \end{gathered}$ | $\begin{gathered} 4.9 \%(2.2-10.7) \\ 3.0 \%(1.1-8.6) \\ 0.0 \end{gathered}$ |
| Race * \% (95\% CI) |  |  |
| Primary: |  |  |
| White | 3.9 (2.6-5.9) | - |
| Multiple Asian | $\begin{aligned} & 6.3(1.7-23.7) \\ & 6.2(2.7-14.5) \end{aligned}$ | - |
| Missing | $1.9(0.3-13.5)$ | - |
| Hispanic | 5.4 (1.4-21.0) | - |
| Black/AIAN | 21.0 (5.8-76.4) | - |
| Secondary: |  |  |
| White | - | $4.2(2.2-7.9)$ |
| Essential worker status |  |  |
| Yes | 4.2 (2.0-8.8) | 7.1 (2.3-21.3) |
| No | 5.8 (3.5-9.7) | 3.6 (1.7-7.6) |
| Missing Response | 3.5 (0.9-14.2) |  |

Table 2 cont. Weighted seropositivity by main demographic variables, SARS-CoV-2 serology surveys in university-affiliated populations, Massachusetts Jul-Aug, 2020.

| Participant type |  |  |
| ---: | :---: | :---: |
| University-affiliate | $0.053(0.035-0.081)$ | $4.3(2.1-8.6)$ |
| Household member | $0.051(0.024-0.112)$ | $3.3(0.8-13.0)$ |
| State Emergency Response |  |  |
| Region (Figure 1) |  | - |
| Region 1 | $7.8(3.9-15.6)$ | - |
| Region 2 | $1.6(0.2-11.3)$ | - |
| Region 3 | $3.2(1.0-10.7)$ | - |
| Region 4 | $5.7(3.0-10.8)$ | $3.1(1.4-6.5)$ |
| Region 5 | $6.2(2.3-16.5)$ | $11.3(4.1-31.3)$ |
|  |  |  |
| Region 1 | - | - |
| Regions 2/3/4/5 |  |  |

*Notes: AIAN= American Indian/Alaska Native. All proportions are adjusted for non-response.

Table 3. Multivariable associations for SARS-CoV-2 seropositivity, Primary Sampling Group, MA USA, Jul-Aug 2020.

| Characteristic | Prevalence Ratio | 95\% CI | p-value |
| :---: | :---: | :---: | :---: |
| Emergency Response Region <br> Region 1 <br> Region 2 <br> Region 3 <br> Region 4 <br> Region 5 | $\begin{gathered} 1.48 \\ 0.34 \\ 0.53 \\ \text { Reference } \\ 1.02 \end{gathered}$ | $\begin{aligned} & 0.62,3.52 \\ & 0.05,2.45 \\ & 0.14,1.96 \\ & 0.35,2.98 \end{aligned}$ | $\begin{aligned} & 0.38 \\ & 0.28 \\ & 0.34 \\ & \\ & 0.97 \end{aligned}$ |
| Age (years) | 1.04 | 0.96, 1.12 | 0.33 |
| Gender <br> Female, Gender diverse, or No response | Reference $0.50$ | 0.27, 0.92 | 0.027 |
| RaceWhite <br>  <br>  <br>  <br> Multiple <br> Asian <br>  <br>  <br> Missing Race <br> Hispanic <br>  | Reference <br> 1.91 <br> 1.66 <br> 0.51 <br> 1.76 <br> 4.49 | $\begin{aligned} & 0.46,7.98 \\ & 0.66,4.16 \\ & 0.07,3.71 \\ & 0.44,7.04 \\ & 1.57,12.9 \end{aligned}$ | $\begin{gathered} 0.38 \\ 0.28 \\ 0.51 \\ 0.42 \\ \mathbf{0 . 0 0 5} \end{gathered}$ |
| Febrile illness since February <br> Missing Response | $\begin{gathered} \text { Reference } \\ 2.42 \\ 0.33 \\ \hline \end{gathered}$ | $\begin{aligned} & 1.24,4.75 \\ & 0.04,2.45 \end{aligned}$ | $\begin{gathered} \mathbf{0 . 0 1 0} \\ 0.28 \end{gathered}$ |
| Other household member | $\begin{gathered} \text { Reference } \\ 0.29 \end{gathered}$ | $0.01,7.69$ | 0.46 |

Table 4. Multivariable associations for SARS-CoV-2 seropositivity, Secondary Sampling Group, Massachusetts, USA, Jul-Aug 2020.

| Characteristic | Prevalence Ratio | 95\% CI | p-value |
| :---: | :---: | :---: | :---: |
| Age (years) | 1.03 | 0.99, 1.07 | 0.17 |
| Gender | Reference $1.35$ | $0.43,4.31$ | $0.61$ |
| Race <br> All Other/Multiple/Missing | $\begin{gathered} \text { Reference } \\ 0.58 \end{gathered}$ | $0.07,4.86$ | $0.62$ |
| Febrile illness since February <br> Missing Response | $\begin{gathered} \text { Reference } \\ 2.56 \\ 2.35 \\ \hline \end{gathered}$ | $\begin{gathered} 0.68,9.67 \\ 0.21,26.73 \\ \hline \end{gathered}$ | $\begin{gathered} - \\ 0.17 \\ 0.49 \end{gathered}$ |
| Emergency Response Region <br> Region 1 <br> Regions 2/3/4/5 | Reference $4.08$ | $1.09,15.33$ | $0.039$ |
| Other household member | Reference $0.70$ | $0.18,2.72$ | $0.61$ |

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Figures

Fig 1. Emergency Response Region sampling frames, for SARS-CoV-2 seropositivity, Massachusetts, USA, Jul-Aug 2020.

Fig 2. Participant inclusion (CONSORT) enrollment, SARS-CoV-2 serosurvey, Massachusetts, USA, Jul-Aug 2020.

## Author contributions

AAL and DA led study conceptualization and design.
LB and NR informed overall aims, sampling design, and statistical analysis.
JR, TS, and EYC implemented all field implementation and associated data collection. DA organized, performed and reported all laboratory-based testing.
JR, TS, EYC, and AAL performed data cleaning and analyses, and wrote the first draft. All authors contributed to revisions of the final manuscript.
All authors read and approved the final manuscript.

## COI statements

None of the authors report any conflicts of interest, financial or non-financial, regarding this study.

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## Data sharing

Fully-identified data will be deposited at: https://osf.io/437tg/.

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Excluded, not eligible ( $n=407$ )

- Not in MA for past 8 weeks ( $\mathrm{n}=355$ )
- Previous COVID-19 diagnosis ( $\mathrm{n}=14$ )


## Supplemental Information

Serological surveys to estimate cumulative incidence of SARS-CoV-2 infection in adults (Sero-MAss study), Massachusetts, July-August 2020: a mail-based cross-sectional study<br>* Correspondence to: alover@umass.edu<br>School of Public Health and Health Sciences<br>University of Massachusetts- Amherst<br>715 North Pleasant Street<br>Amherst, MA 01003-9304<br>phone: 413-545-7426

## Supplemental Information

Supplemental Table 1: Comparison of characteristics of participants selected for randomization and those not selected for randomization, SARS-CoV-2 serosurvey, Massachusetts, USA, JulAug 2020.


Supplemental Figure 1. Distributions of IgG log normal optical densities (OD) ratios, by subgroups, SARS-CoV-2, Massachusetts, USA, Jul-Aug 2020.


| Sero-MAss] | $\begin{gathered} \text { Item } \\ \text { No } \\ \hline \end{gathered}$ | Recommendation | $\begin{gathered} \text { Page } \\ \text { No } \\ \hline \end{gathered}$ |
| :---: | :---: | :---: | :---: |
| Title and abstract | 1 | (a) Indicate the study's design with a commonly used term in the title or the abstract | 1 |
|  |  | (b) Provide in the abstract an informative and balanced summary of what was done and what was found | 2-3 |
| Introduction |  |  |  |
| Background/rationale | 2 | Explain the scientific background and rationale for the investigation being reported | 4-5 |
| Objectives | 3 | State specific objectives, including any prespecified hypotheses | 4 |
| Methods |  |  |  |
| Study design | 4 | Present key elements of study design early in the paper | 5 |
| Setting | 5 | Describe the setting, locations, and relevant dates, including periods of recruitment, exposure, follow-up, and data collection | 5-6 |
| Participants | 6 | (a) Give the eligibility criteria, and the sources and methods of selection of participants | 5-6 |
| Variables | 7 | Clearly define all outcomes, exposures, predictors, potential confounders, and effect modifiers. Give diagnostic criteria, if applicable | 15 <br> (Table <br> 1) |
| Data sources/ measurement | 8* | For each variable of interest, give sources of data and details of methods of assessment (measurement). Describe comparability of assessment methods if there is more than one group | 6-7 |
| Bias | 9 | Describe any efforts to address potential sources of bias | 8 |
| Study size | 10 | Explain how the study size was arrived at | $\begin{aligned} & 7 ; 21 \\ & \text { (Fig } 2) \end{aligned}$ |
| Quantitative variables | 11 | Explain how quantitative variables were handled in the analyses. If applicable, describe which groupings were chosen and why | 8-9 |
| Statistical methods | 12 | (a) Describe all statistical methods, including those used to control for confounding | 7-9 |
|  |  | (b) Describe any methods used to examine subgroups and interactions | 8 |
|  |  | (c) Explain how missing data were addressed | 7-8 |
|  |  | (d) If applicable, describe analytical methods taking account of sampling strategy | 8 |
|  |  | (e) Describe any sensitivity analyses | n/a |
| Results |  |  |  |
| Participants | 13* | (a) Report numbers of individuals at each stage of study-eg numbers potentially eligible, examined for eligibility, confirmed eligible, included in the study, completing follow-up, and analysed | $\begin{aligned} & 7 ; 21 \\ & (\text { Fig 2) } \end{aligned}$ |
|  |  | (b) Give reasons for non-participation at each stage | 9 |
|  |  | (c) Consider use of a flow diagram | Fig. 2 |
| Descriptive data | 14* | (a) Give characteristics of study participants (eg demographic, clinical, social) and information on exposures and potential confounders | Fig 1. |
|  |  | (b) Indicate number of participants with missing data for each variable of interest | Supp. |


| Outcome data | 15* | Report numbers of outcome events or summary measures | 16 <br> (Table <br> 2) |
| :---: | :---: | :---: | :---: |
| Main results | 16 | (a) Give unadjusted estimates and, if applicable, confounder-adjusted estimates and their precision (eg, $95 \%$ confidence interval). Make clear which confounders were adjusted for and why they were included | 9-11 |
|  |  | (b) Report category boundaries when continuous variables were categorized | $\mathrm{n} / \mathrm{a}$ |
|  |  | (c) If relevant, consider translating estimates of relative risk into absolute risk for a meaningful time period | n/a |
| Other analyses | 17 | Report other analyses done-eg analyses of subgroups and interactions, and sensitivity analyses | 10 |
| Discussion |  |  |  |
| Key results | 18 | Summarise key results with reference to study objectives | 11-12 |
| Limitations | 19 | Discuss limitations of the study, taking into account sources of potential bias or imprecision. Discuss both direction and magnitude of any potential bias | 13 |
| Interpretation | 20 | Give a cautious overall interpretation of results considering objectives, limitations, multiplicity of analyses, results from similar studies, and other relevant evidence | 13-14 |
| Generalisability | 21 | Discuss the generalisability (external validity) of the study results | 14 |
| Other information |  |  |  |
| Funding | 22 | Give the source of funding and the role of the funders for the present study and, if applicable, for the original study on which the present article is based | 3 |

*Give information separately for exposed and unexposed groups.

Note: An Explanation and Elaboration article discusses each checklist item and gives methodological background and published examples of transparent reporting. The STROBE checklist is best used in conjunction with this article (freely available on the Web sites of PLoS Medicine at http://www.plosmedicine.org/, Annals of Internal Medicine at $\mathrm{http}: / / \mathrm{www} . a n n a l s . o r g /$, and Epidemiology at http://www.epidem.com/). Information on the STROBE Initiative is available at www.strobe-statement.org.

## BMJ Open

## Serological surveys to estimate cumulative incidence of SARS-CoV-2 infection in adults (Sero-MAss study), Massachusetts, July-August 2020- a mail-based crosssectional study

| Journal: | BMJ Open |
| :---: | :---: |
| Manuscript ID | bmjopen-2021-051157.R1 |
| Article Type: | Original research |
| Date Submitted by the Author: | 23-Jul-2021 |
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| <b>Primary Subject Heading</b>: | Infectious diseases |
| Secondary Subject Heading: | Epidemiology |
| Keywords: | EPIDEMIOLOGY, Public health < INFECTIOUS DISEASES, Epidemiology < INFECTIOUS DISEASES |

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# Serological surveys to estimate cumulative incidence of SARS-CoV-2 infection in adults (Sero-MAss study), Massachusetts, July-August 2020: a mail-based cross-sectional study 

Teah Snyder, ${ }^{1 \dagger}$ Johanna Ravenhurst, ${ }^{1 \dagger}$ Estee Y. Cramer, ${ }^{1 \dagger}$ Nicholas G. Reich, ${ }^{1}$ Laura B. Balzer, ${ }^{1}$ Dominique Alfandari, ${ }^{2}$ and Andrew A. Lover ${ }^{1 *}$<br>$\dagger$ Equal contributions.<br>${ }^{1}$ Department of Biostatistics and Epidemiology, School of Public Health and Health Sciences, University of Massachusetts- Amherst, Amherst MA<br>${ }^{2}$ Department of Veterinary and Animal Sciences, College of Natural Sciences, University of Massachusetts- Amherst, Amherst MA<br>* Correspondence to: alover@umass.edu<br>School of Public Health and Health Sciences<br>University of Massachusetts- Amherst<br>715 North Pleasant Street<br>Amherst, MA 01003-9304<br>phone: 413-545-7426


#### Abstract

Objectives: To estimate the seroprevalence of anti-SARS-CoV-2 $\operatorname{IgG}$ and IgM among Massachusetts residents and to better understand asymptomatic SARS-CoV-2 transmission during the summer of 2020.

Design: Mail-based cross-sectional study Setting: Massachusetts, United States Participants: Primary sampling group: sample of undergraduate students at the University of Massachusetts, Amherst $(\mathrm{n}=548)$ and a member of their household $(\mathrm{n}=231)$.

Secondary sampling group: sample of graduate students, faculty, librarians and staff ( $\mathrm{n}=214$ ) and one member of their household $(\mathrm{n}=78)$. All participants were residents of Massachusetts without prior COVID-19 diagnosis.

Primary and secondary outcome measures: Prevalence of SARS-CoV-2 seropositivity. Association of seroprevalence with variables including age, gender, race, geographic region, occupation, and symptoms.

Results: Approximately 27,000 persons were invited via email to assess eligibility. 1,001 households were mailed dried blood spot sample kits, 762 returned blood samples for analysis. In the primary sample group, 36 (4.6\%) had IgG antibodies detected for an estimated weighted prevalence in this population of $5.3 \%(95 \% \mathrm{CI}: 3.5$ to 8.0$)$. In the secondary sampling group, 10 (3.4\%) had IgG antibodies detected for an estimated adjusted prevalence of $4.0 \%(95 \% \mathrm{CI}: 2.2$ to 7.4). No samples were IgM positive. No association was found in either group between seropositivity and self-reported work duties or customer-facing hours. In the primary sampling group, self-reported febrile illness since Feb 2020, male sex, and minority race (Black or American Indian/Alaskan Native) were associated with seropositivity. No factors except geographic regions within the state were associated with evidence of prior SARS-CoV-2 infection in the secondary sampling group.

Conclusions: This study fills a critical gap in estimating the levels of sub-clinical and asymptomatic infection. Estimates can be used to calibrate models estimating levels of population immunity over time, and these data are critical for informing public health interventions and policy.


## Strengths and limitations of this study

- Our study collected serological samples in a well-defined rigorous sample frame in a contact-less (mail-based) survey in an early stage of the pandemic in an area of high SARS-CoV-2 burden.
- A range of potentially associated demographic, occupational, and behavioral factors were surveyed to contextualize seropositivity across geographic regions within the state.
- Our study sampled from populations affiliated with a large public university in Massachusetts, and may not be generalizable to the general population.


## Introduction

Since emergence in early 2020, the SARS-CoV-2 virus has severely impacted the entire globe. The state of Massachusetts was heavily impacted in the earliest stages of the pandemic, and a "super-spreader" event in the state in April 2020 may have seeded large case clusters throughout the country. ${ }^{1}$ However, the trajectory of the early stages of transmission in the state, as well as across the US remain poorly understood due to changes in case definitions and limited testing of both symptomatic and asymptomatic persons during the summer of 2020. ${ }^{2}$ To assess seroprevalence across the state, a mail-based serosurvey was implemented July-August 2020. At the time of this survey, the Massachusetts Department of Public Health had reported over 109,143 confirmed COVID-19 cases and over 8,081 deaths. ${ }^{3}$ Sero-epidemiological studies are a critical tool to explore infection dynamics, especially where asymptomatic or subclinical infections are common, as for SARS-CoV-2. ${ }^{4}$ This study helps to fill a critical gap in estimating the levels of sub-clinical and asymptomatic infection to inform consequent levels of populationimmunity. ${ }^{5}$

Concurrent to this study, a number of seroprevalence studies were conducted on the east coast of the United States; these studies focused on specific populations at high risk and found varying results. A survey in April 2020 in a convenience sample of 200 asymptomatic residents of Chelsea, MA found an estimated of seroprevalence of $31.5 \%(17.5 \% \mathrm{IgM}+/ \mathrm{IgG}+, 9.0 \%$ $\mathrm{IgM}+/ \mathrm{IgG}-$ and $5.0 \% \mathrm{IgM}-/ \mathrm{IgG}+) .{ }^{6}$ This study used a small convenience sample and did not include any randomization. ${ }^{6}$ A study with a larger sample of over 28,000 clinical patient samples in New York City, USA found an IgG seropositivity prevalence of $44 \%$ with over $50 \%$ of participants reporting no symptoms. ${ }^{7}$

Other seroprevalence surveys across the US have found generally low- to moderate prevalence in a diverse set of study populations. A study of 790 university students in Los Angeles, California conducted in April and May of 2020 estimated a prevalence of SARS-CoV-2 IgG antibody of 4.0\% ( $95 \%$ CI: 3.0 to $5.1 \%) .{ }^{8}$ During May - April of 2020, a cross-sectional study in St. Louis found IgG seropositivity to be estimated at $1.71 \%$ ( $95 \% \mathrm{CI} 0.04 \%$ to $3.38 \%$ ) in pediatric patients and $3.11 \%(95 \% \mathrm{CI}: 0.92 \%$ to $5.32 \%)$ in adult patients. In the most comprehensive serosurvey from the spring and summer months of 2020, 16,025 clinical samples were analyzed with IgG
spike protein sero-reactivity ranging from $1.0 \%$ in the San Francisco Bay Area to $6.9 \%$ in New York City. ${ }^{9}$ These disparate results highlight major geographic variability in the trajectory of infections, and reinforce the need for additional seroprevalence studies to more fully contextualize trends in immunity to SARS-CoV-2 targeting specific geographic regions.

Though community seroprevalence studies generally rely on serum samples collected in health facilities, the use of dried blood spot (DBS) samples is a practical and effective alternative. ${ }^{10}$ DBS samples involve a small finger-prick sample self-collected by participants in their own homes. The use of dried blood samples for antibody assays has been validated in other work prior to the current pandemic, ${ }^{10,11}$ and previous studies have evaluated the feasibility, validity, and acceptability of using DBS samples for SARS-CoV-2 antibody testing. ${ }^{12-15}$ This method of sample collection facilitates efficient population-level sampling while minimizing social mixing and concurrent potential exposures.

This study estimated the prevalence of previous infection with SARS-CoV-2 in individuals who had not been diagnosed with COVID-19 and were asymptomatic with representative coverage across the entire state of Massachusetts, USA. Information from this study can provide knowledge regarding the seropositivity of this population and can be used to inform decisionmaking regarding community re-openings during the pandemic.

## Methods

## Study Design and Participants

The study population included undergraduate students, graduate students, staff, and faculty members currently affiliated with the University of Massachusetts, Amherst (UMass)_and their household members. On-campus classes were suspended in mid-March 2020; consequently, undergraduates had exposure to the local epidemiology within their communities from March until sampling in July-August throughout the state (primary sampling group). Conversely, graduate students, faculty, staff, librarians and their family members (secondary sampling group) almost universally reside in close proximity to Amherst, and broadly reflect transmission in the Western part of the state. UMass affiliates were eligible to participate in this study if they were above the age of 18, had been living in Massachusetts for the past eight weeks; had never
received a COVID-19 diagnosis from a medical professional; and did not have a fever greater than $100.4^{\circ} \mathrm{F}$ at the time of survey completion. Household members were eligible for inclusion if they met all of these same criteria and were between the ages of 23 and 78 (chosen to expand sampling beyond college-age population groups). Both UMass affiliates and their household members had to complete online consent forms in order to participate in the study. Upon meeting eligibility criteria, participants were directed to a consent form which they reviewed prior to providing their first and last name, the date, and an electronic signature.

An institutional email list was provided by university administration for recruitment. Initial emails were sent out to UMass affiliates between June 23, 2020 and June 26, 2020 for participant recruitment. The email provided information about the study and links to a screening eligibility survey, informed consent document, initial survey regarding COVID-19 risk factors, and information regarding shipping addresses. If the UMass affiliate had a household member interested in participating, a single household member was invited to complete the eligibility, consent, and initial survey forms. The household member was invited to participate prior to analyzing samples from the main participant. To increase participation rates, two reminder emails were sent to all non-respondents (day three and six after initial solicitation). All survey responses were collected and stored in REDCap. ${ }^{16}$

The survey was closed after a three-week enrollment period, and a subset of participants were selected to receive a test kit. To select a population representative of the broader UMass community across the entire state of Massachusetts, two sampling schemes were applied. The first consisted of all undergraduates and their household members (primary sampling group); the second sampling frame consisted of graduate students, staff, faculty members, librarians, and their household members (secondary sampling group). Within the primary sampling group, selection for biosample collection used probability proportional to population size, using the most recent census data aggregated to state-level emergency response regions due to sparse county-level populations (Figure 1). ${ }^{17}$ For the secondary sampling group, selection for biosampling was via simple random sampling.

The full sample frame selection is shown in Figure 2. Briefly, an email invitation was sent to a total of 27,339 individuals, of which 4,124 completed the screening, informed consent, and initial survey forms. A total of 1,001 individuals were then randomly selected to receive a sampling kit. Participants were mailed all materials to safely collect and return samples, including lancets, alcohol wipes, gauze, gloves, bandages, a bloodspot collection card, a pre-paid shipping box, and detailed printed instruction cards (including a nurse call line).

Upon mailing out the test kits, participants were also emailed a link containing a video on how to collect the DBS, along with a detailed survey form with demographics, risk factors, and any current symptoms or COVID diagnoses. No participants reported a COVID-19 diagnosis between the initial survey and sample collection several weeks later. All shipments utilized a Biological Substance Category B (UN3373) shipping box.

## Ethical approvals

This study was approved by the University of Massachusetts-Amherst Human Research Protection Office (Approval \#2062; April 27, 2020).

## Sample Preparation and ELISA Analysis

Upon receipt of boxes, the sample cards (Whatman ${ }^{\circledR}$ Protein Saver 903) were heat-treated (30 minutes at $56^{\circ} \mathrm{C}$ ); a single blood spot per card was punched ( 0.25 -inch diameter) and transferred to an ELISA plate. Plates were coated with $1 \mu \mathrm{~g} / \mathrm{ml}$ of purified RBD diluted in PBS overnight at $4^{\circ} \mathrm{C}$ and blocked with tris-buffered saline with $0.1 \%$ Tween 20 (TBST) containing $5 \%$ non-fat dry milk. DBS were eluted in $500 \mu 1$ of TBST overnight at $4^{\circ} \mathrm{C}$ and $50 \mu 1$ of each sample was added to the ELISA plate preloaded with $50 \mu \mathrm{l}$ of TBST containing $2 \%$ non-fat dry milk. Samples were then assayed for SARS-CoV-2 antibodies according to published protocols. ${ }^{18,19}$ The RBD protein was produced in-house via transfection of HEK293T cells using polyethylenimine (plasmid was a generous gift from Pr. F. Kramer Mount Sinai School of Medicine). Batches were controlled for purity by SDS-PAGE followed by Coomassie staining and ELISA using an anti His-tag monoclonal antibody. Optical densities were read at 405 nm , and each 96-well plate contained seven negative controls and one positive control (serum from PCR-confirmed case at $1 / 100$ dilution). Samples were tested against IgG and positive samples
were confirmed and tested with anti-IgM antibodies. Optical density values were normalized to the mean optical density of negative controls daily.

## Data Analysis

Sample size and power
The study was designed to assess seropositivity within the primary sampling group with sufficient precision to inform policy. With 750 persons, and an assumed 5\% positivity, the $95 \%$ CI for this estimate is $3.6 \%$ to $6.9 \%$. Within the five emergency response subregions, at $5 \%$ seropositivity, the survey is powered for a precision of $2.3 \%$ to $10.2 \%$. The secondary sampling group ( $\mathrm{n}=250$ ) sample size was based on logistic limitations, but was powered to a precision of $2.8 \%$ to $8.8 \%$. All confidence intervals are binomial exact, without adjustments for study design effects or non-response.

## Analysis of serology data

Finite mixture models were used to determine seropositivity cutoffs. These latent-class models estimate breakpoints for seropositive and seronegative subpopulations, and have been applied to a range of pathogen serosurvey data, including rubella, pertussis, and parvovirus. ${ }^{20-22}$ From this analysis, all samples with an IgG optical density ratio $\geq 2.49$-fold above daily background were considered positive for SARS-CoV-2 (Distributions shown in Supplemental Figure 1; and sensitivity analysis with alternative cutpoints can be found in Supplemental Table 2).

## Adjusted estimates

All reported prevalences and prevalence ratio estimates are adjusted with non-response weights, which were estimated using inverse weighting. Briefly, logistic regression models were used to calculate propensity scores for each individual in the sample using reported gender and race categories. These were transformed to probabilities; a small number of individuals had extremely large weights due to sparse strata; these weights were truncated at $1 / 0.02 .{ }^{23}$ Weights were then applied to all prevalence and prevalence ratio estimations using the survey package in R. ${ }^{24}$ The primary sampling group sample was self-weighting due to probability proportional to population size sampling. Sampling weights were not used in the secondary sampling group as selection used simple random sampling.

## Multivariable analyses for prevalence ratios

Prevalence ratios were estimated to assess factors associated with seropositivity, with indivikdual Poisson models ${ }^{25}$ for both of the two sampling groups, with robust (sandwich) errors to address clustering within households.

Bivariate analyses were performed for each factor separately. All variables with a p-value $<0.20$ based on bivariate association with outcome were further evaluated for inclusion in final models. All final models were adjusted for age (continuous), race, and gender (see Table 1). Due to several very sparse categories, some were combined in the final models. Specifically, all race/ethnicity categories and all geographic regions were not included in analysis of the secondary sampling group due to unstable estimates.

Model parsimony was evaluated using AIC/BIC and all tests were two-tailed, with $\alpha=0.05$. R version 4.0.3 and SAS version 9.4 (SAS Institute, Inc, Cary, NC, USA) were used for analysis.

## Patient and public involvement

All members of the university community were invited to participate, and serological testing was proved at no cost to either the sampled individuals or to their selected household contact.

## Results

A total of 1,001 individuals were enrolled into the study; this included 752 undergraduate students, 90 graduate students, 63 faculty/librarians, and 96 staff members (Figure 2). Seventysix percent of these $(\mathrm{n}=762)$ returned blood samples for analysis; 548 in the primary sampling group, and 214 in the secondary sampling group. Of the 548 participants in the primary sampling group, 230 enrolled a household member. One household member submitted a sample without the sample of the main participant, bringing the total number of undergraduate household members to 231 . Of the 214 participants in the secondary sampling group, 78 enrolled a household member. Two returned samples were excluded from analysis due to unlinkable samples. A total of 1,071 samples were included in the final analyses: 762 main participants and 309 household members (Figure 2).

Demographic characteristics of both sampling groups are presented in Table 1. Race categories do not total to $100 \%$ due to non-response and multiple possible answers. Age, gender, and essential worker status were broadly similar between those invited to participate and those who completed the study (Supplemental Table 1).

Of the total 1,071 samples tested, 46 were positive for SARS-CoV-2 antibodies. Demographic results are stratified by IgG serostatus (Table 2); no samples showed evidence for IgM positivity. Seropositivity was low-to-moderate across the survey groups, with several important exceptions. Variation is apparent by sex, race, and across geographic regions; however, several strata have wide confidence intervals due to small sample sizes.

Of the 779 primary sampling group participants and their household members, 36 were positive for SARS-CoV-2 antibodies. This corresponds to an overall seroprevalence of $5.3 \%$ ( $95 \% \mathrm{CI}$ : $3.1-7.5$ ) of the population after adjustment for nonresponse and geographic location. In the secondary sampling group, of the 292 graduate students, staff, librarians, faculty members, and their household members, ten (adjusted $4.0 \%, 95 \%$ CI: 1.6-6.5) had evidence for prior SARS-CoV-2 infection. (Table 2). Results were also further stratified by UMass affiliate vs. household member. Of the 548 undergraduate students in the primary sampling group, 27 were positive for SARS-CoV-2 IgG antibodies (population positivity rate of $5.3 \%$ ( $95 \% \mathrm{CI}: 3.1-7.6 \%$ ). Of the 231 household members of undergraduate participants, nine (adjusted 5.2\%, 95\% CI: 1.2-9.2) were positive for SARS-CoV-2 IgG antibodies. In the secondary sampling group, eight University affiliates (adjusted $4.3 \%$, $95 \%$ CI: 1.3-7.3) were positive for SARS-CoV-2 IgG antibodies. Of the household members in the secondary sampling group, 2 were seropositive, with a weighted seroprevalence of ( $3.3 \%, 95 \%$ CI: $0.0-7.8 \%$ ). (Table 2). The overall distributions of measured IgG lognormal optical density ratios by subgroups are broadly similar (Supplemental Figure 1).

After adjustments for age, gender, region, and self-reported febrile illness since February 2020, the strongest association with seropositivity in the primary sampling group was Black or AIAN Race ( $\mathrm{PR}=4.49,95 \% \mathrm{CI}=1.57,12.9$ ) (Table 3). This indicates that individuals who reported being Black or American Indian/Alaskan Native had a prevalence 3.49 times higher than White
individuals after adjustment. Additionally, after adjustments, females and those who are gender diverse were at a significantly lower risk of prior SARS-CoV-2 infection $(\mathrm{PR}=0.5 ; 95 \% \mathrm{CI}=$ $0.27,0.92$ ) compared to males. Those who reported a febrile illness in February were more likely to be seropositive than those who did not report any febrile illness in this time period (PR $=2.42,95 \% \mathrm{CI}=1.24,4.75)$. No significant associations were found across each of the 5 geographic regions in the primary sampling group, however the prevalence of seropositivity was $48 \%$ higher in Region 1 compared to Region $4(\mathrm{PR}=1.48,95 \% \mathrm{CI}=0.62,3.52)$.

Within the secondary sampling population, after adjustments for age, race, gender, region, household member and self-reported febrile illness since February (Table 4), participants who reported residing in either Region 2, 3, 4, or 5 had greater than 4 times higher prevalence of SARS-CoV-2 antibodies as compared to those who resided in Region $1(\mathrm{PR}=4.08,95 \% \mathrm{CI}=$ $1.09,15.33)$. No other factors included in the model were significantly associated with seropositivity.

## Discussion

This mail-based serosurvey of two university-affiliated populations across Massachusetts in July and August 2020 found an estimated seroprevalence of $\sim 5 \%$ of antibodies to SARS-CoV-2. These results indicate that even with extensive morbidity and mortality across the state at the time of sampling, there had been limited exposure to SARS-CoV-2 at a population-level. This estimated seroprevalence is lower than that detected with concurrent community-based studies in other states. An estimated $14.3 \%$ of the United States population had been previously infected with SARS-CoV-2 by November 2020, as estimated in a pooled analysis of multiple seroprevalence surveys. ${ }^{26}$

Our estimates are substantially lower than some models of COVID-19 seroprevalence in Massachusetts. One model estimates a seroprevalence of $16.2 \%$ (no CIs provided) on Jul 27, 2020 (the closest modeled date to these surveys). These estimates are nearly double our measured seroprevalence with inclusion of 110,000 confirmed cases at that date (ca. 1.5\%). ${ }^{27}$ These differences may be due to several factors, including a non-representative population by age or geographic range, or waning of antibody titers. Without CIs, we are unable to evaluate
coverage outside the reported point estimate. However, alternate nowcasting estimates suggest a total statewide attack rate on July 31, 2020 of $6.9 \%$ ( $95 \%$ CrI: $5.5-8.4 \%$ ) in Massachusetts, ${ }^{28}$ which are closely aligned with our estimates.

The primary sampling group also showed increased risk of seropositivity with self-reported illness since February 2020; this association was not observed in analyses of the secondary sampling group in multivariable analysis. Within the surveyed groups, approximately $24 \%$ of the primary sampling group and $18 \%$ of the secondary sampling group reported illness since February. This finding may indicate that some participants in our study may not have been strictly asymptomatic, and were simply unable to obtain a COVID-19 test due to limited availability during the beginning of the pandemic. This finding reinforces results suggesting asymptomatic illness is an important contributor to the observed pandemic trajectories.

Differing antibody dynamics have been reported in other studies. A number of studies have found sustained antibody levels for over 3 months, ${ }^{29,30}$ while others studies suggest IgG levels can remain 6 months or more. ${ }^{31-33}$ An additional study has reported rapid waning of routine serological markers in individuals who had lower initial antibody responses. ${ }^{34}$. Only $7.1 \%$ of those with high titers at baseline seroreverted to a level below the threshold for positivity within 60 days, compared to $64.9 \%$ of those with lower titers at baseline. ${ }^{34}$ Evidence for IgM seropositivity was not detected in any of IgG positive samples, which is consistent with results from other surveys studies that included asymptomatic or subclinical populations due to rapidly waning titers. ${ }^{32,35}$ Studies have shown that IgM levels decline more rapidly after infection than IgA and IgG levels, ${ }^{30,36,37}$ and this is especially apparent with asymptomatic and sub-clinical infections. ${ }^{32,35}$

Trends in SARS-CoV-2 antibody levels are complex, and vary greatly depending on the measured antibody, timing of sampling, and severity of disease. ${ }^{31,32,35}$ Seroconversion times vary depending on the study, but one study found a median time-to-seroconversion for $\operatorname{IgM}$ of 8 days and median seroconversion for IgG of 10 days. Additionally, the SARS-CoV-2 IgG response generally begin around 10-15 days after symptom onset. ${ }^{2}$ For this reason, repeated serial
sampling of convalescent populations should be prioritized to more fully understand the dynamics of immune response.

SARS-CoV-2 seropositivity was associated with minority race status in this survey. While the total number of non-white participants was limited, the large effect size reinforces other studies suggesting that marginalized communities have been and continue to be excessively impacted by the pandemic. Results from the primary sampling group analysis suggest that self-reported Black race is a risk factor for previous SARS-CoV-2 infection, which is consistent with findings from other studies. ${ }^{38-40}$ A number of factors may contribute to this significant association including data suggesting that Black individuals are more likely to work in frontline industries or live in areas with a higher population density in many settings. ${ }^{41}$ No parallel associations were found in the analysis of the secondary sampling group due to limited sample size in some strata. In the secondary sampling group, the aggregation of Race categories into White race and Nonwhite race likely obscured meaningful associations between Race and seropositivity.

Results from the primary sampling group showed increased prevalence of IgG seropositivity among males. After adjusting for age, race, and region, male gender remained a statistically significant risk factor for evidence of prior SARS-CoV-2 infection. Other studies have similarly found that males have higher rates of infection than females for asymptomatic infections ${ }^{4243}$ These findings may reflect differences in care seeking behavior (recruitment biases), true biological differences, or differences in health behaviors such as smoking, alcohol use, and COVID-19 prevention measures. ${ }^{44}$ This association was not observed in the secondary sampling group.

This study was population-based and had broad eligibility criteria but is subject to several limitations. The exclusion of persons with confirmed diagnoses and any current symptoms (due to biosafety concerns) also inherently limited capture of sub-clinical infections. As such, the estimates are likely a lower bound. However, participants who suspected they may have been previously infected with SARS-CoV-2 might be more likely to participate compared to those that were not concerned with prior infection. This is a pervasive issue in community-based studies, where characteristics of those who volunteer to participate in community-based research differ
from the general population. ${ }^{45}$ Randomization after a three-week enrollment period helped to address this limitation, as using only the first participants to volunteer could have biased the sample to include those who were most motivated to receive their antibody test results. If participants were more motivated to receive their results because they thought they suspected a prior exposure to SARS-CoV2, this would have inflated the observed prevalence of seropositivity in the study population.

Another limitation of the study is the self-reported response of the lack of a prior COVID-19 diagnosis and current fever. It is possible that some participants shielded their answer and submitted samples for analysis without meeting the eligibility criteria; this would have inflated our estimation of seroprevalence in asymptomatic groups, The limited number of non-white, and gender-diverse participants also limited some analyses and restricted out ability to assess any differences in prevalence across all racial groups. Next, while multiple studies have validated DBS sampling for SARS-CoV-2, ${ }^{46}$ waning antibodies in asymptomatic individuals could be below the limited of detection of the ELISA assay. Additionally, the cutoff was determined in this study to be 2.49 for seropositivity. When using a single cut-point for a continuous variable, there is the possibility of outcome misclassification, however we attempted to reduce misclassification through the use of a finite mixture model. Finally, generalizability is limited due to the recruitment of a university-affiliated population in a relatively restricted geographic area. ${ }^{8}$ The population in our study primarily included young adults, those of working age, and their household members. Individuals in older age groups or who reside in institutional settings would not have been recruited for our study. This population may not be representative of the broader US population and may be healthier, include few non-Whites, has higher education levels, and may have differing socio-political attitudes with consequent health impacts. Additionally, samples were collected during the summer months of 2020 in Massachusetts during which the state was in phase 3 of the reopening plan. The state government implemented a strict "lockdown" in March 2020 and progression to each reopening phase required a reduction in COVID-19 cases and hospitalizations. Massachusetts also had a mandatory mask order in all public spaces beginning May 1, 2020. Other states in New England followed similar timelines, however the implementation timelines and effectiveness differed widely across the United States.

It is probable that the government policy measures on a state-wide level influenced seroprevalence, with stricter guidelines resulting in lower viral exposure.

In conclusion, this serosurvey estimates prevalence of prior SARS-CoV-2 infections in a university-affiliated population in Massachusetts, with adjusted prevalences of $5.3 \%$ in the primary sampling group and $4.0 \%$ within the secondary sampling group. Risk factors for IgG seropositivity included self-reported recent febrile illness, minority race status, and male gender. This study reinforces the critical need for targeted serosurveys in highest-risk and marginalized communities, both in Massachusetts, and nationwide.

This study provides important estimates of seroprevalence in Massachusetts after the "first wave" of SARS-CoV-2 infections in the spring of 2020. These are critical to benchmark modeling studies, and to more comprehensively understand the dynamics of population-level sero-status throughout the continuing pandemic, especially as vaccines become widely available.

Tables

Table 1. Demographics of study populations, SARS-CoV-2 serology surveys in universityaffiliated populations, Massachusetts Jul-Aug, 2020.

| Characteristic | Primary Sampling Group $(\mathrm{N}=779)$ | Secondary Sampling Group $(\mathrm{N}=292)$ |
| :---: | :---: | :---: |
| Gender |  |  |
| Female | 499 (64.1\%) | 154 (52.7\%) |
| Male | 274 (35.2\%) | 136 (46.6\%) |
| Gender Diverse | 5 (0.6\%) | 2 (0.7\%) |
| Missing | 1 (0.1\%) | 0 (0.0\%) |
| Race |  |  |
| AIAN | 17 (2.2\%) | 1 (0.3\%) |
| Asian | 78 (10.0\%) | 34 (11.6\%) |
| Black | 12 (1.5\%) | 3 (1.0\%) |
| Hispanic | 36 (4.6\%) | 9 (3.1\%) |
| Multiple | 37 (4.8\%) | 11 (3.8\%) |
| White | 545 (70.0\%) | 217 (74.3\%) |
| Missing | 54 (6.9\%) | 17 (5.8\%) |
| Age |  |  |
| Mean | 29.9 | 41.6 |
| Median | 21 | 39 |
| Range | 18-75 | 21-75 |
| Education |  |  |
| HS/GED | 102 (13.1\%) | 5 (1.7\%) |
| Some College | 483 (62.0\%) | 24 (8.2\%) |
| BA/BS | 117 (15.0\%) | 78 (26.7\%) |
| More Than BA/BS | 74 (9.5\%) | 183 (62.7\%) |
| Missing | 3 (0.4\%) | 2 (0.7\%) |
| Essential Worker |  |  |
| No | 533 (68.4\%) | 224 (76.7\%) |
| Yes | 195 (25.0\%) | - 51 (17.5\%) |
| Missing | 51 (6.6\%) | 17 (5.8\%) |
| Self-reported attitude about COVID-19 |  |  |
| Strongest fear | 135 (17.3\%) | 72 (24.7\%) |
| Somewhat fearful | 389 (49.9\%) | 122 (41.8\%) |
| Neutral/Missing | 139 (17.8\%) | 63 (21.6\%) |
| Somewhat not fearful | 86 (11.0\%) | 23 (7.9\%) |
| Not fearful | 30 (3.9\%) | 12 (4.1\%) |
| Self-reported febrile illness since | No: 534 (68.6\%) | No: 224 (76.7\%) |
| February | Yes: 188 (24.1\%) | Yes: 53 (18.2\%) |
|  | Missing: 57 (7.3\%) | Missing: 15 (5.1\%) |


| Self-reported care seeking | No: $112(59.6 \%)$ | No: $32(60.4 \%)$ |
| :--- | :---: | :---: |
| (if reporting illness since Feb) | Yes: $75(39.9 \%)$ | Yes: $21(39.6 \%)$ |
|  | Missing: $1(0.5 \%)$ | Missing: $0(0.0 \%)$ |

*Notes: AIAN= American Indian/Alaska Native. The primary sampling group includes UMass undergraduates and household members, and the secondary sampling group includes UMass affiliated faculty, staff, and graduate students and household members.

Table 2. Weighted seropositivity by main demographic variables, SARS-CoV-2 serology surveys in university-affiliated populations, Massachusetts Jul-Aug, 2020.

| Characteristic | Primary Sampling Group | Secondary Sampling Group |
| :---: | :---: | :---: |
| Age in years, median | $\begin{gathered} 21 \\ 95 \% \mathrm{CI}(20-21) \end{gathered}$ | $\begin{gathered} 41 \\ 95 \% \mathrm{CI}(38-44) \end{gathered}$ |
| Prevalence of SARS-CoV-2 antibodies, by sub-group |  |  |
| Overall Population Prevalence ( $95 \% \mathrm{CI}$ ) | 5.3\% (3.5-8.0) | 4.0\% (2.2-7.4) |
| Sex \%, (95\% CI) <br> Female <br> Male <br> Gender Diverse/No Response | $\begin{gathered} 4.0 \%(2.4-6.6) \\ 8.7 \%(5.1-15.0) \\ 0.0 \end{gathered}$ | $\begin{gathered} 4.9 \%(2.2-10.7) \\ 3.0 \%(1.1-8.6) \\ 0.0 \end{gathered}$ |
| Race * \% (95\% CI) |  |  |
| Primary: |  |  |
| White | 3.9 (2.6-5.9) | - |
| Multiple | 6.3 (1.7-23.7) | - |
| Asian | $6.2(2.7-14.5)$ | - |
| Missing | $1.9(0.3-13.5)$ | - |
| Hispanic | 5.4 (1.4-21.0) | - |
| Black/AIAN | 21.0 (5.8-76.4) | - |
| Secondary: |  |  |
| White | - | $4.2(2.2-7.9)$ |
| Essential worker status |  |  |
| Yes | 4.2 (2.0-8.8) | 7.1 (2.3-21.3) |
| No | 5.8 (3.5-9.7) | 3.6 (1.7-7.6) |
| Missing Response | 3.5 (0.9-14.2) | 0 |

Table 2 cont. Weighted seropositivity by main demographic variables, SARS-CoV-2 serology surveys in university-affiliated populations, Massachusetts Jul-Aug, 2020.

| Participant type |  |  |
| ---: | :---: | :---: |
| University-affiliate | $5.3(3.5-8.1)$ | $4.3(2.1-8.6)$ |
| Household member | $5.1(2.4-11.2)$ | $3.3(0.8-13.0)$ |
| State Emergency Response |  |  |
| Region (see Figure 1) |  | - |
| Region 1 | $7.8(3.9-15.6)$ | - |
| Region 2 | $1.6(0.2-11.3)$ | - |
| Region 3 | $3.2(1.0-10.7)$ | - |
| Region 4 | $5.7(3.0-10.8)$ | $3.1(1.4-6.5)$ |
| Region 5 | $6.2(2.3-16.5)$ | $11.3(4.1-31.3)$ |
|  |  |  |
| Region 1 | - | - |

*Notes: AIAN= American Indian/Alaska Native. All prevalences are adjusted for non-response.

Table 3. Multivariable associations for SARS-CoV-2 seropositivity, Primary Sampling Group, Massachusetts USA, Jul-Aug 2020.

| Characteristic | Prevalence Ratio | 95\% CI | p-value |
| :---: | :---: | :---: | :---: |
| Emergency Response Region |  |  |  |
| Region 1 | 1.48 | 0.62, 3.52 | 0.38 |
| Region 2 | 0.34 | 0.05, 2.45 | 0.28 |
| Region 3 | 0.53 | 0.14, 1.96 | 0.34 |
| Region 4 | Reference | - | - |
| Region 5 | 1.02 | 0.35, 2.98 | 0.97 |
| Age (years) | 1.04 | 0.96, 1.12 | 0.33 |
| Gender |  |  |  |
| Male | Reference | - | - |
| Female, Gender diverse, or No response | 0.50 | 0.27, 0.92 | 0.027 |
| Race |  |  |  |
| White | Reference | - | - |
| Multiple | 1.91 | 0.46, 7.98 | 0.38 |
| Asian | 1.66 | 0.66, 4.16 | 0.28 |
| Missing Race | 0.51 | 0.07, 3.71 | 0.51 |
| Hispanic | 1.76 | 0.44, 7.04 | 0.42 |
| Black or AIAN | 4.49 | 1.57, 12.9 | 0.005 |
| Febrile illness since February |  |  |  |
| No | Reference | $\square{ }^{-}$ | - |
| Yes | 2.42 | 1.24, 4.75 | 0.010 |
| Missing Response | 0.33 | 0.04, 2.45 | 0.28 |
| Other household member |  |  |  |
| no | Reference |  | - |
| yes | 0.29 | 0.01, 7.69 | 0.46 |

Table 4. Multivariable associations for SARS-CoV-2 seropositivity, Secondary Sampling Group, Massachusetts, USA, Jul-Aug 2020.

| Characteristic | Prevalence Ratio | 95\% CI | p-value |
| :---: | :---: | :---: | :---: |
| Age (years) | 1.03 | 0.99, 1.07 | 0.17 |
| Gender <br> Female or Gender Diverse | Reference $1.35$ | $0.43,4.31$ | $0.61$ |
| Race <br> White <br> All Other/Multiple/Missing | $\begin{gathered} \text { Reference } \\ 0.58 \end{gathered}$ | $0.07,4.86$ | $0.62$ |
| Febrile illness since February | $\begin{gathered} \text { Reference } \\ 2.56 \\ 2.35 \end{gathered}$ | $\begin{gathered} 0.68,9.67 \\ 0.21,26.73 \\ \hline \end{gathered}$ | $\begin{gathered} - \\ 0.17 \\ 0.49 \\ \hline \end{gathered}$ |
| Emergency Response Region <br> Region 1 <br> Regions 2/3/4/5 | Reference $4.08$ | $1.09,15.33$ | $0.039$ |
| Other household member | Reference 0.70 | $0.18,2.72$ | $0.61$ |

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## Figures

Fig 1. Study sampling frames for SARS-CoV-2 seropositivity surveys, Massachusetts, USA, JulAug 2020. (Note: State-level Emergency Response Regions shown in orange; anonymized participant locations shown as red markers).

Fig 2. Participant enrollment diagram, (CONSORT), SARS-CoV-2 serosurvey, Massachusetts, USA, Jul-Aug 2020.

## Author contributions

AAL and DA led study conceptualization and design.
LB and NR informed overall aims, sampling design, and statistical analysis.
JR, TS, and EYC implemented all field implementation and associated data collection.
DA organized, performed and reported all laboratory-based testing.
JR, TS, EYC, and AAL performed data cleaning and analyses, and wrote the first draft. All authors contributed to revisions of the final manuscript. All authors read and approved the final manuscript.

## COI statements

None of the authors report any conflicts of interest, financial or non-financial, regarding this study.

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## Data sharing

Fully-identified data will be deposited at: https://osf.io/437tg/.

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Fig 1. Study sampling frames for SARS-CoV-2 seropositivity surveys, Massachusetts, USA, Jul-Aug 2020. (Note: State-level Emergency Response Regions shown in orange; anonymized participant locations shown as red markers).

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# Supplemental Information 

Serological surveys to estimate cumulative incidence of SARS-CoV-2
infection in adults (Sero-MAss study), Massachusetts, July-August
2020: a mail-based cross-sectional study

* Correspondence to: alover@umass.edu

School of Public Health and Health Sciences
University of Massachusetts- Amherst
715 North Pleasant Street
Amherst, MA 01003-9304
phone: 413-545-7426

## Supplemental Information

Supplemental Table 1: Comparison of characteristics of participants selected for randomization and those not selected for randomization, SARS-CoV-2 serosurvey, Massachusetts, USA, JulAug 2020.

| n | Primary Sampling Group (undergraduates) <br> Not <br> Randomized Randomized <br> 869 <br> 752 |  | Secondary sampling group (employees) <br> Not Randomized Randomized 2253 <br> 249 |  |
| :---: | :---: | :---: | :---: | :---: |
| Age (mean (SD)) | 20.64 (3.09) | 20.18 (1.79) | 42.05 (13.53) | 40.69 (13.88) |
| Gender (\%) <br> Unknown <br> Female <br> Gender diverse <br> Male | $\begin{gathered} 0(0.0) \\ 525(60.4) \\ 12(1.4) \\ 332(38.2) \end{gathered}$ | $\begin{gathered} 0(0.0) \\ 453(60.2) \\ 8(1.1) \\ 291(38.7) \end{gathered}$ | $\begin{gathered} 1(0.0) \\ 1262(56.0) \\ 49(2.2) \\ 941(41.8) \end{gathered}$ | $\begin{gathered} 0(0.0) \\ 147(59.0) \\ 3(1.2) \\ 99(39.8) \end{gathered}$ |
| Self-reported febrile illness since February (\%) <br> Unknown <br> No <br> Not sure <br> Yes | $\begin{gathered} 1(0.1) \\ 568(65.4) \\ 75(8.6) \\ 225(25.9) \\ \hline \end{gathered}$ | $\begin{gathered} 2(0.3) \\ 499(66.4) \\ 57(7.6) \\ 194(25.8) \end{gathered}$ | $\begin{gathered} 6(0.3) \\ 1658(73.6) \\ 154(6.8) \\ 435(19.3) \\ \hline \end{gathered}$ | $\begin{gathered} 0(0.0) \\ 189(75.9) \\ 16(6.4) \\ 44(17.7) \end{gathered}$ |
| Education (\%) <br> Unknown <br> BA/BS <br> High school / GED <br> More than BA/BS <br> Prefer not to answer <br> Some college | $\begin{gathered} 4(0.5) \\ 15(1.7) \\ 90(10.4) \\ 2(0.2) \\ 1(0.1) \\ 757(87.1) \end{gathered}$ | $\begin{gathered} 3(0.4) \\ 14(1.9) \\ 108(14.4) \\ 0(0.0) \\ 0(0.0) \\ 627(83.4) \end{gathered}$ | $\begin{gathered} 6(0.3) \\ 509(22.6) \\ 46(2.0) \\ 1528(67.8) \\ 13(0.6) \\ 151(6.7) \\ \hline \end{gathered}$ | $\begin{gathered} 1(0.4) \\ 64(25.7) \\ 2(0.8) \\ 166(66.7) \\ 1(0.4) \\ 15(6.0) \end{gathered}$ |
| Response to, "I am afraid of COVID-19" (\%) <br> Unknown <br> Neither agree nor disagree <br> Somewhat agree <br> Somewhat disagree <br> Strongly agree <br> Strongly disagree | $\begin{gathered} 1(0.1) \\ 199(22.9) \\ 390(44.9) \\ 115(13.2) \\ 121(13.9) \\ 43(4.9) \\ \hline \end{gathered}$ | $\begin{gathered} 2(0.3) \\ 141(18.8) \\ 374(49.7) \\ 99(13.2) \\ 104(13.8) \\ 32(4.3) \\ \hline \end{gathered}$ | $\begin{gathered} 2(0.1) \\ 399(17.7) \\ 1051(46.6) \\ 153(6.8) \\ 573(25.4) \\ 75(3.3) \end{gathered}$ | $\begin{gathered} 0(0.0) \\ 51(20.5) \\ 100(40.2) \\ 22(8.8) \\ 66(26.5) \\ 10(4.0) \\ \hline \end{gathered}$ |

Supplemental Table 2: Comparison of estimated seropositivity by sampling groups using alternative cutpoints for seropositivity, SARS-CoV-2 serosurvey, Massachusetts, USA, Jul-Aug 2020.

| Sampling Group | Cutoff Value | Unadjusted Prevalence (\%) | Adjusted Prevalence (\%) (95\% confidence interval) |
| :---: | :---: | :---: | :---: |
| Primary sampling group (undergraduates) |  |  |  |
|  | 2.49 | 4.6 | 5.3 (3.5-8.0) |
|  | 2.0 | 10.9 | 11.6 (9.1-15.0) |
|  | 2.1 | 8.7 | $9.2(6.9-12.3)$ |
|  | 2.2 | 7.1 | 7.6 (5.5-10.6) |
|  | 2.3 | 5.9 | 6.5 (4.5-9.3) |
|  | 2.4 | 5.0 | 5.6 (3.8-8.4) |
|  | 2.5 | 4.5 | 5.1 (3.3-7.9) |
|  | 2.6 | 4.0 | 4.5 (2.8-7.1) |
|  | 2.7 | 3.7 | 4.3 (2.6-6.9) |
|  | 2.8 | 3.3 | $3.9(2.3-6.5)$ |
|  | 2.9 | 2.8 | 3.5 (2.0-6.2) |
|  | 3.0 | 2.4 | 3.3 (1.8-5.9) |
| Secondary sampling group <br> (employees) |  |  |  |
|  | 2.49 | 3.4 | 4.0 (2.2-7.4) |
|  |  | - |  |
|  | 2.0 | 7.5 | 8.2 (5.3-12.7) |
|  | 2.1 | 5.5 | 6.0 (3.7-9.8) |
|  | 2.2 | 4.8 | 5.4 (3.2-9.2) |
|  | 2.3 | 4.8 | 5.4 (3.2-9.2) |
|  | 2.4 | 3.8 | 4.5 (2.5-8.0) |
|  | 2.5 | 3.4 | 4.0 (2.2-7.4) |
|  | 2.6 | 2.4 | 3.0 (1.5-6.2) |
|  | 2.7 | 2.1 | 2.6 (1.2-5.7) |
|  | 2.8 | 2.1 | 2.6 (1.2-5.7) |
|  | 2.9 | 1.4 | 1.7 (0.7-4.6) |
|  | 3.0 | 1.0 | 1.3 (0.4-4.0) |

Supplemental Figure 1. Distributions of IgG lognormal optical density (OD) ratios, by subgroups, SARS-CoV-2, Massachusetts, USA, Jul-Aug 2020.


| Sero-MAss] | Item No | Recommendation | $\begin{gathered} \text { Page } \\ \text { No } \\ \hline \end{gathered}$ |
| :---: | :---: | :---: | :---: |
| Title and abstract | 1 | (a) Indicate the study's design with a commonly used term in the title or the abstract | 1 |
|  |  | (b) Provide in the abstract an informative and balanced summary of what was done and what was found | 2-3 |
| Introduction |  |  |  |
| Background/rationale | 2 | Explain the scientific background and rationale for the investigation being reported | 4-5 |
| Objectives | 3 | State specific objectives, including any prespecified hypotheses | 4 |
| Methods |  |  |  |
| Study design | 4 | Present key elements of study design early in the paper | 5 |
| Setting | 5 | Describe the setting, locations, and relevant dates, including periods of recruitment, exposure, follow-up, and data collection | 5-6 |
| Participants | 6 | (a) Give the eligibility criteria, and the sources and methods of selection of participants | 5-6 |
| Variables | 7 | Clearly define all outcomes, exposures, predictors, potential confounders, and effect modifiers. Give diagnostic criteria, if applicable | 15 <br> (Table <br> 1) |
| Data sources/ measurement | 8* | For each variable of interest, give sources of data and details of methods of assessment (measurement). Describe comparability of assessment methods if there is more than one group | 6-7 |
| Bias | 9 | Describe any efforts to address potential sources of bias | 8 |
| Study size | 10 | Explain how the study size was arrived at | $\begin{aligned} & 7 ; 21 \\ & \text { (Fig } 2) \end{aligned}$ |
| Quantitative variables | 11 | Explain how quantitative variables were handled in the analyses. If applicable, describe which groupings were chosen and why | 8-9 |
| Statistical methods | 12 | (a) Describe all statistical methods, including those used to control for confounding | 7-9 |
|  |  | (b) Describe any methods used to examine subgroups and interactions | 8 |
|  |  | (c) Explain how missing data were addressed | 7-8 |
|  |  | (d) If applicable, describe analytical methods taking account of sampling strategy | 8 |
|  |  | (e) Describe any sensitivity analyses | n/a |
| Results |  |  |  |
| Participants | 13* | (a) Report numbers of individuals at each stage of study-eg numbers potentially eligible, examined for eligibility, confirmed eligible, included in the study, completing follow-up, and analysed | $\begin{aligned} & 7 ; 21 \\ & (\text { Fig 2) } \end{aligned}$ |
|  |  | (b) Give reasons for non-participation at each stage | 9 |
|  |  | (c) Consider use of a flow diagram | Fig. 2 |
| Descriptive data | 14* | (a) Give characteristics of study participants (eg demographic, clinical, social) and information on exposures and potential confounders | Fig 1. |
|  |  | (b) Indicate number of participants with missing data for each variable of interest | Supp. |


| Outcome data | 15* | Report numbers of outcome events or summary measures | 16 <br> (Table <br> 2) |
| :---: | :---: | :---: | :---: |
| Main results | 16 | (a) Give unadjusted estimates and, if applicable, confounder-adjusted estimates and their precision (eg, $95 \%$ confidence interval). Make clear which confounders were adjusted for and why they were included | 9-11 |
|  |  | (b) Report category boundaries when continuous variables were categorized | $\mathrm{n} / \mathrm{a}$ |
|  |  | (c) If relevant, consider translating estimates of relative risk into absolute risk for a meaningful time period | n/a |
| Other analyses | 17 | Report other analyses done-eg analyses of subgroups and interactions, and sensitivity analyses | 10 |
| Discussion |  |  |  |
| Key results | 18 | Summarise key results with reference to study objectives | 11-12 |
| Limitations | 19 | Discuss limitations of the study, taking into account sources of potential bias or imprecision. Discuss both direction and magnitude of any potential bias | 13 |
| Interpretation | 20 | Give a cautious overall interpretation of results considering objectives, limitations, multiplicity of analyses, results from similar studies, and other relevant evidence | 13-14 |
| Generalisability | 21 | Discuss the generalisability (external validity) of the study results | 14 |
| Other information |  |  |  |
| Funding | 22 | Give the source of funding and the role of the funders for the present study and, if applicable, for the original study on which the present article is based | 3 |

*Give information separately for exposed and unexposed groups.

Note: An Explanation and Elaboration article discusses each checklist item and gives methodological background and published examples of transparent reporting. The STROBE checklist is best used in conjunction with this article (freely available on the Web sites of PLoS Medicine at http://www.plosmedicine.org/, Annals of Internal Medicine at $\mathrm{http}: / / \mathrm{www} . a n n a l s . o r g /$, and Epidemiology at http://www.epidem.com/). Information on the STROBE Initiative is available at www.strobe-statement.org.

