



## The burden of active infection and anti-SARS-CoV-2 IgG antibodies in the general population: Results from a statewide sentinel-based population survey in Karnataka, India

Giridhara R. Babu<sup>a,\*</sup>, Rajesh Sundaresan<sup>b</sup>, Siva Athreya<sup>c</sup>, Jawaid Akhtar<sup>d</sup>, Pankaj Kumar Pandey<sup>e</sup>, Parimala S. Maroor<sup>e</sup>, M. Rajagopal Padma<sup>e</sup>, R. Lalitha<sup>f</sup>, Mohammed Shariff<sup>e</sup>, Lalitha Krishnappa<sup>g</sup>, C.N. Manjunath<sup>h</sup>, Mysore Kalappa Sudarshan<sup>e</sup>, Gopalkrishna Gururaj<sup>i</sup>, Timmanahalli Sobagaiah Ranganath<sup>j</sup>, Kumar D.E. Vasanth<sup>e</sup>, Pradeep Banandur<sup>i</sup>, Deepa Ravi<sup>a</sup>, Shilpa Shiju<sup>e</sup>, Eunice Lobo<sup>a</sup>, Asish Satapathy<sup>k</sup>, Lokesh Alahari<sup>k</sup>, Prameela Dinesh<sup>e</sup>, Vinitha Thakar<sup>e</sup>, Anita Desai<sup>i</sup>, Ambica Rangaiah<sup>j</sup>, Ashok Munivenkatappa<sup>l</sup>, Krishna S<sup>m</sup>, Shantala Gowdara Basawarajappa<sup>j</sup>, H.G. Sreedhara<sup>n</sup>, Siddesh KC<sup>o</sup>, Amrutha Kumari B<sup>p</sup>, Nawaz Umar<sup>q</sup>, Mythri BA<sup>r</sup>, Ravi Vasanthapuram<sup>i</sup>

<sup>a</sup> Indian Institute of Public Health – Bengaluru, Public Health Foundation of India, Magadi Rd 1st Cross, Next to Leprosy Hospital, SIHFW Premises, Bengaluru, Karnataka, India

<sup>b</sup> Indian Institute of Science, CV Raman Rd, Bengaluru, Karnataka, India

<sup>c</sup> Indian Statistical Institute – Bangalore Centre, 8th Mile, Mysore Rd, RVCE Post, Bengaluru, Karnataka, India

<sup>d</sup> Department of Health and Family Welfare Services, Government of Karnataka, Vikasa Soudha, Bengaluru, Karnataka, India

<sup>e</sup> Department of Health and Family Welfare Services, Aarogya Soudha, 1st cross, Magadi Road, Bengaluru, Karnataka, India

<sup>f</sup> State Maternal and PPTCT Consultant, UNICEF, Bengaluru, India

<sup>g</sup> MS Ramaiah Medical College, MS Ramaiah Nagar, Mathikere, Bengaluru, Karnataka, India

<sup>h</sup> Sri Jayadeva Institute of Cardiovascular Sciences and Research, Bannerghatta Main Rd, Phase 3, Jayanagara 9th Block, Jayanagar, Bengaluru, Karnataka, India

<sup>i</sup> National Institute of Mental Health and Neurosciences, Hosur Road, Bengaluru, Karnataka, India

<sup>j</sup> Bangalore Medical College and Research Institute, Fort, K.R. Road, Bengaluru, India

<sup>k</sup> Member Technical Advisory Committee on COVID19, Bengaluru, India

<sup>l</sup> National Institute of Virology, Bangalore Unit, Someshwaranagar, 1st Main, Dharmaram College Post, Bengaluru, India

<sup>m</sup> Vijayanagar Institute of Medical Sciences, Ballari, Karnataka, India

<sup>n</sup> Hassan Institute of Medical Sciences, Sri Chamarajendra Hospital Campus, Krishnaraja Pura, Hassan, Karnataka, India

<sup>o</sup> Shimoga Institute of Medical Sciences, Sagar Road, Shimoga, Karnataka, India

<sup>p</sup> Mysore Medical College and Research Institute, Irwin Road, Mysuru, Karnataka, India

<sup>q</sup> Gulbarga Institute of Medical Sciences, Veeresh Nagar, Sedam Road Kalaburagi, Karnataka, India

<sup>r</sup> Karnataka Institute of Medical Sciences, PB Rd, Vidya Nagar, Hubli, Karnataka, India

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### ABSTRACT

**Objective:** To estimate the burden of active infection and anti-SARS-CoV-2 IgG antibodies in Karnataka, India, and to assess variation across geographical regions and risk groups.

**Methods:** A cross-sectional survey of 16,416 people covering three risk groups was conducted between 3–16 September 2020 using the state of Karnataka's infrastructure of 290 healthcare facilities across all 30 districts. Participants were further classified into risk subgroups and sampled using stratified sampling. All participants were subjected to simultaneous detection of SARS-CoV-2 IgG using a commercial ELISA kit, SARS-CoV-2 antigen using a rapid antigen detection test (RAT) and reverse transcription-polymerase chain reaction (RT-PCR) for RNA detection. Maximum-likelihood estimation was used for joint estimation of the adjusted IgG, active and total prevalence (either IgG or active or both), while multinomial regression identified predictors.

\* Corresponding author at: Indian Institute of Public Health – Bengaluru, Public Health Foundation of India, Magadi Rd 1st Cross, Next to Leprosy Hospital, SIHFW Premises, Bengaluru, Karnataka, 560023, India.

E-mail address: [giridhar@iiph.org](mailto:giridhar@iiph.org) (G.R. Babu).

**Results:** The overall adjusted total prevalence of COVID-19 in Karnataka was 27.7% (95% CI 26.1–29.3), IgG 16.8% (15.5–18.1) and active infection fraction 12.6% (11.5–13.8). The case-to-infection ratio was 1:40 and the infection fatality rate was 0.05%. Influenza-like symptoms or contact with a COVID-19-positive patient were good predictors of active infection. RAT kits had higher sensitivity (68%) in symptomatic people compared with 47% in asymptomatic people.

**Conclusion:** This sentinel-based population survey was the first comprehensive survey in India to provide accurate estimates of the COVID-19 burden. The findings provide a reasonable approximation of the population immunity threshold levels. Using existing surveillance platforms coupled with a syndromic approach and sampling framework enabled this model to be replicable.

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## Introduction

The global pandemic of SARS-CoV-2 causing coronavirus disease 2019 (COVID-19) raged across the world within a few months. India has the second-highest burden of COVID-19, with 8.8 million infected and 130,070 deaths as of 16 November 2020 (Ministry of Health and Family Welfare Services, 2020). Currently, India has only case-based reporting as the prime strategy for estimating disease burden and trend through all the epidemic phases. Case-based reporting has the advantages of rationalizing testing, isolating cases, and tracing and quarantining contacts (Hellewell et al., 2020); however, it does not provide an estimate of the true burden of the disease, as it mostly identifies sicker people seeking care or those who have better access to healthcare. Hence, the reported case counts of COVID-19 have grossly underestimated the true prevalence or disease burden of the pandemic. The two rounds of national seroprevalence surveys conducted by the Indian Council of Medical Research (ICMR) indicated that 81–130 infections were missed for every reported case in the initial survey conducted in May 2020 (Murhekar et al., 2020), which improved to missing nearly 26–32 infections per reported case by August 2020. Serological surveys, such as those conducted by the ICMR, can help to understand the burden of past infections. It is also important to detect active infections during a pandemic; however, this is challenging since 45% of the infected people have mild or no symptoms (Chatterjee et al., 2020). Furthermore, the estimation is affected by poor in-person testing due to inaccessibility, stigma and supply-side inadequacies.

Effective public health measures require reliable understanding of the existing burden of disease through epidemiological investigations. Joint estimation of IgG prevalence and active SARS-CoV-2 infections can help to detect, manage and control the disease outbreak. Seroprevalence estimates from around the world show varying numbers ranging from 0.07% in hospital patients to 54.1% in slum inhabitants (Malani et al., 2021; Herzog et al., 2020; Meyerowitz-Katz and Merone, 2020; Korth et al., 2020; Stringhini et al., 2020; Hallal et al., 2020; Noh et al., 2020; Financial Express, 2021; Aarti et al., 2020). The ICMR survey results reported 0.73% prevalence across India (May–June 2020), which increased to 7% by the end of September (Murhekar et al., 2020). The surveys in slums and non-slums of Mumbai (Malani et al., 2021) showed considerable variation: 54.1% (95% CI 52.7–55.6) and 16.1% (95% CI 14.9–17.4) prevalence, respectively. Serosurveys in a healthcare setting of North India showed prevalence increasing from 2.3% in April to 50.6% in July (Siddiqui et al., 2020). However, there are concerns about using only IgG prevalence as a marker of population immunity threshold and for estimation of period prevalence of infection. These include inability to detect IgG antibodies over time, varying sampling methods, the unreliable nature of the predictive value of positive antibody tests with varying sensitivity and specificity of different tests affecting the tests' reliability, and the presence of other types of immune response (La Marca et al., 2020; To et al., 2020;

Zou et al., 2020). Some of these concerns could be mitigated by understanding the burden of active infections concurrently with IgG estimation.

Karnataka has an estimated population of 70.7 million (Statistics DoEa, 2013) spread over 191,791 km<sup>2</sup>. The first confirmed COVID-19 case was reported on 09 March 2020. As of 16 November 2020, there were 861,647 cumulative cases, 27,146 active cases and 11,529 deaths (Ministry of Health and Family Welfare Services, 2020). This study aimed to estimate the combined proportion of the population with past or active SARS-CoV-2 infections in Karnataka and assess the variation across geographical regions and risk groups. The results of what is perhaps the first comprehensive statewide sentinel-based survey in India are presented here. The design and analysis methodologies of this survey can serve as the blueprint for other similar surveys.

## Methods

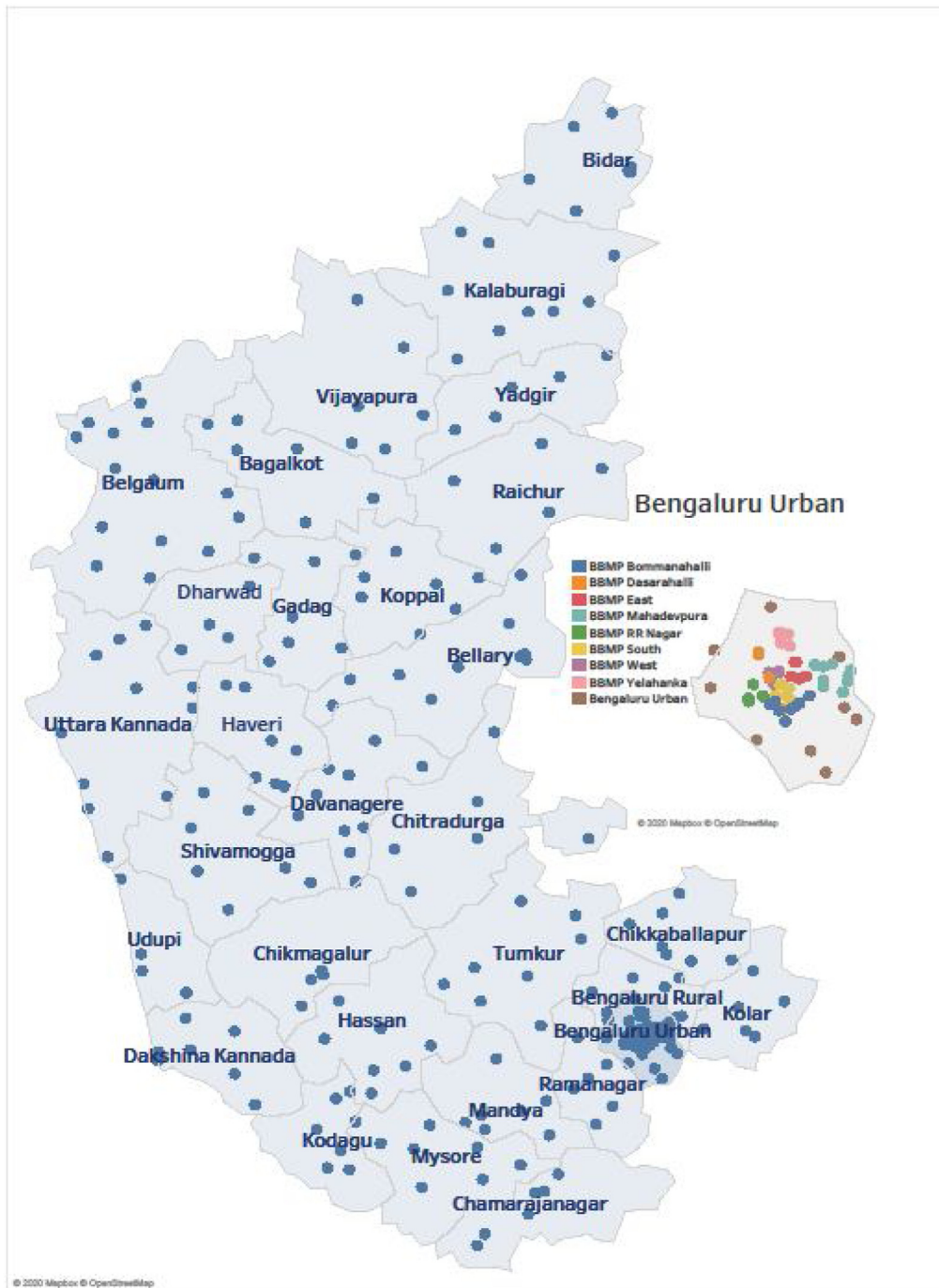
### Study setting, design and sample size

#### Setting

This was the first round of the proposed serial cross-sectional surveys across the districts of Karnataka. This state has 30 administrative districts. The capital district Bengaluru has approximately 13.6 million residents. The study was conducted during 03–16 September 2020.

#### Design

Each district was a unit of the survey, except Bengaluru, which was subdivided into nine units. From the resulting 38 units, geographically representative healthcare facilities (district hospitals or community healthcare centres or primary healthcare centres) with the expertise to conduct the survey were selected for this sentinel-based population survey (Figure 1 and Appendix D). The participants included only adults aged  $\geq 18$  years. The survey excluded those already diagnosed with SARS-CoV-2 infection or those who did not agree to provide informed consent to participate in the survey. The population was stratified into three risk groups based on community exposure and vulnerability to COVID-19: the low-risk group comprised pregnant women presenting for a periodic antenatal check-up, people attending the outpatient departments for common ailments and their attendees; the moderate-risk group comprised people with high contact in the community such as bus conductors and autorickshaw drivers, vendors at vegetable markets, healthcare workers, individuals in containment zones, people in congregate settings (markets, malls, retail stores, bus stops, railway stations), and waste collectors (The Lancet, 2020); the high-risk group comprised the elderly (aged  $\geq 60$  years) and people with comorbid conditions such as chronic liver, lung or renal disease, diabetes, heart disease, hypertension, immunocompromised conditions, and malignancy. The protocol mandated that the elderly be systematically sampled from a population register, and those with comorbidities be systematically sampled from a list maintained by



**Figure 1.** Sites (blue dots) of the survey representing geographical spread across Karnataka. The inset picture shows the sites across Bengaluru (multi-coloured dots).

the facility on those suffering from non-communicable diseases. In each of the other risk subgroups, participants were also systematically sampled.

**Sample size**

A 10% prevalence and design effect of 3 were assumed. A margin of error of 5% was required. Under normal error assumption, a standard calculation for a 95% CI required that the minimum sample size be 432 per unit; this was then equally divided among the risk groups (144 per risk group). In each risk group, the number of samples was further equally divided among the risk subgroups.

This led to a total sample size of 16,416 across the 38 units (Supplementary Table 4/Appendix D).

**Sample collection and laboratory testing**

From participants in the low-risk group, both nasopharyngeal and oropharyngeal swab samples were collected for the RT-PCR test following the ICMR protocol and 4 ml of venous blood for the IgG antibody test. In the moderate-risk and high-risk groups, two swab samples in different media were collected for the antigen and RT-PCR tests and 4 ml of venous blood.

The rapid antigen detection test (RAT) was performed using the Antigen Standard Q COVID-19 Ag detection kit, a rapid chromatographic immunoassay for the qualitative detection of antigens specific to SARS-CoV-2. The RT-PCR test was performed for all low-risk participants and those who tested negative on the RAT through the current ICMR-approved testing network. For antibody testing, the collected venous blood sample was left undisturbed at room temperature for 30 min for clotting, then centrifuged at 3,000 rpm, and the serum was transported to the laboratory by maintaining a cold chain. SARS-CoV-2-specific IgG antibodies were detected using a commercially available, validated and ICMR-approved kit (Covid Kavach Anti SARS-CoV-2 IgG antibody detection ELISA, Zydus Cadila, India) (Sapkal et al., 2020); the test was performed as per the manufacturer's instructions. The results were interpreted as positive or negative for SARS-CoV-2 IgG antibodies based on the cut-off value of optical densities obtained with positive and negative samples provided in the kit.

#### Data collection

After obtaining written informed consent, information on basic demographic details, exposure history to laboratory-confirmed COVID-19 cases, symptoms suggestive of COVID-19 in the preceding month, and clinical history were recorded on a web-based application designed specifically for the study and were linked to the samples using the ICMR Specimen Referral Forms for COVID-19. The category, symptoms, contact, and comorbidity information for participants were gathered using the web-based application. RAT/RT-PCR results were entered into the ICMR test-data portal. The IgG antibody test results were retrieved directly from the labs. A consolidated line-list of all the participants was then created. From this, subsets of participants in risk categories, subcategories, age groups, sex, and geographical units were used to jointly estimate IgG prevalence, active infection fraction, and total burden in the respective categories. Symptoms and comorbidity data, which were part of the consolidated line-list, were used in the regression. Only anonymised data with no personal identifiers were used for the analysis.

#### Ethical considerations

The Institutional Ethics Committee (IEC) of the Indian Institute of Public Health-Bengaluru campus reviewed and approved the study (vide. IIPHBB/TRCIEC/174/2020). The participants were informed of the purpose of the survey, the samples that would be taken and were requested to respond to some screening questions. Those who were aged <18 years, already diagnosed with SARS-CoV-2 infection, unwilling to provide samples for the test, and who did not agree to provide informed consent were excluded. After obtaining informed consent, information on basic demographic details, exposure history, symptoms observed in the previous month, and clinical history were noted. Participants' test results were available and shared with them by the concerned healthcare facility.

#### Statistical analysis

IgG prevalence was defined as the fraction of the sampled population with IgG antibodies. Active infection fraction was defined as the fraction of the sampled population who tested positive on the RT-PCR/RAT test. Total prevalence of COVID-19 was defined as the fraction of the sampled population with either IgG or active infection.

To enable joint estimation of IgG prevalence, active infection fraction and total prevalence, it was first modelled that an

individual be in one of four disease states: having active infection but no IgG antibodies, having IgG antibodies but no evidence of active infection, having both IgG antibodies and active infection, and having neither active infection nor IgG antibodies. The disease state of the individual is, however, hidden and can only be inferred from the RAT, RT-PCR and IgG antibody test outcomes. This leads to a parametric model for the probabilities of test outcomes (observations), given the disease-state probabilities (parameters of the model), after taking the sensitivities and specificities of the tests into account.

To obtain the joint estimates of the parameters in a stratum, maximum likelihood estimation was used, which *ipso facto* provided estimates already adjusted for the sensitivities and specificities of the tests. The joint estimation was an extension of the Rogan-Gladen formula (Rogan and Gladen, 1978). The procedure also accounted for the protocol-induced variation of test types across participants. Confidence intervals were obtained by invoking asymptotic normality of the maximum likelihood estimates, with their covariance matrix being approximated by the inverse of the Fisher information matrix of the parametric model. Further, weighted adjusted estimates for Karnataka were obtained after weighing each district's prevalence estimates by the population fraction in that district. Facility-level weighting was not used because some of the healthcare facilities served the entire district population. Also, the estimates for prevalence across risk subcategories were unweighted. Odds ratios across risk subcategories (with respect to one reference subcategory) were calculated by restricting attention to the relevant subcategories.

To identify the weights on various independent/explanatory variables (symptoms, comorbidities, etc.) for predicting past infection and active infection, multinomial regressions were used to regress the test outcomes on two disjoint sets of independent variables. The procedure could be embedded within the framework of the generalised linear model with multinomial logit functions along with a custom link function that accounted for the test-type variability across participants, and also the tests' sensitivities and specificities. Important explanatory variables were captured using the Wald test.

The details are given in the Supplementary material provided.

#### Results

This was the first statewide sentinel-based population survey conducted in India. It was carried out in 290 healthcare facilities spread across the state of Karnataka. Of the 16,585 people surveyed in the different risk categories, this study presents the results for 15,624 individuals whose RAT plus RT-PCR and COVID Kavach ELISA antibody test results were matched (Appendix C). A total of 16,585 IgG results were provided. The results of 513 were not considered due to missing information and inability to match the participant in the database; 448 entries were further unmapped to the line list because of manual data-entry errors or because data were not retrievable from the ICMR portal. Also, 18 IgG samples were inconclusive (Figure 1 in Supplementary material/Appendix C).

#### IgG prevalence

The overall weighted adjusted seroprevalence of IgG was 16.8% (95% CI 15.5–18.1); this was as of 03 September 2020 and at the state level, obtained after adjusting for the serial sensitivities and specificities of all tests (Table 1).

#### Active infection

It was estimated that 12.6% (95% CI 11.5–13.8) of the seemingly unsuspecting participants in the general population or an



**Table 1**  
Seroprevalence of IgG antibodies against SARS-CoV-2 and active infection in Karnataka.

Category	Type	Samples <sup>a</sup>	% IgG against SARS-CoV-2 <sup>b</sup>	% active infection with COVID-19 <sup>b</sup>	% prevalence of COVID-19 <sup>b</sup>	Odds ratio	
State	Karnataka	Crude	15,939	2,565/15,939 = 16.1%	2,363/14,132 = 16.7%	4,582/15,939 = 28.7%	
		Adjusted	15,624	15.7	12	26.3	
		Weighted adjusted	15,624	16.8 (15.5–18.1)	12.6 (11.5–13.8)	27.7 (26.1–29.3)	
Demography	Sex	Male	8,165	15.8 (14.3–17.4)	15.5 (13.9–17.2)	29.8 (27.7–31.8)	1.51 (1.23–1.88)
		Female	7,445	14.8 (13.2–16.4)	8.4 (7–9.8)	21.9 (19.9–23.8)	1
	Age, years	18–29	5,184	12.5 (10.7–14.3)	7.1 (5.6–8.6)	19 (16.8–21.3)	1
		30–39	3,353	16 (13.6–18.4)	11.2 (9–13.5)	25.7 (22.7–28.7)	1.47 (1.09–1.99)
		40–49	2,447	15.4 (12.6–18.2)	15.8 (12.8–18.8)	29.3 (25.6–33)	1.77 (1.27–2.44)
		50–59	1,792	17.6 (14.2–21)	17.3 (13.7–20.9)	33.3 (28.9–37.7)	2.13 (1.5–3)
		≥60	2,848	18.1 (15.3–20.8)	15.9 (13.2–18.7)	31.6 (28.1–35)	1.97 (1.44–2.67)
	Region	Urban	14,107	15.8 (14.6–17)	12.4 (11.3–13.6)	26.7 (25.2–28.2)	1.54 (1.11–2.23)
		Rural	1,517	10.6 (7.4–13.7)	9 (5.9–12.1)	19.1 (15–23.3)	1
	Risk category	High-risk	5,322	17.9 (15.9–19.9)	15.9 (13.8–17.9)	31.7 (29.1–34.2)	1.78 (1.37–2.31)
Moderate-risk		5,253	14.3 (12.4–16.2)	12.3 (10.4–14.1)	25.4 (23–27.8)	1.3 (1–1.71)	
Low-risk		5,049	13.6 (11.8–15.5)	8.1 (6.5–9.8)	20.7 (18.4–23)	1	
Risk sub-category <sup>d</sup>	High-risk	Elderly	2,445	17.7 (14.7–20.6)	16.8 (13.7–19.8)	32.4 (28.6–36.2)	2.5 (1.7–3.76)
		People with comorbidities	2,455	18.2 (15.2–21.2)	14.7 (11.8–17.6)	30.5 (26.8–34.2)	2.29 (1.55–3.45)
	Moderate-risk	Containment zones	1,138	16.2 (12–20.4)	16.3 (11.9–20.7)	31 (25.5–36.5)	2.34 (1.45–3.81)
		Bus conductors/auto drivers	1,008	16.1 (11.7–20.6)	13.9 (9.5–18.3)	28.9 (23.2–34.6)	2.12 (1.28–3.51)
		Vendors at vegetable markets	1,025	15.4 (11.1–19.8)	13.5 (9.2–17.8)	27.9 (22.3–33.5)	2.02 (1.22–3.34)
		Congregate settings <sup>c</sup>	1,259	13.6 (9.8–17.3)	13.5 (9.6–17.4)	25.8 (20.9–30.8)	1.81 (1.12–2.95)
	Low-risk	Healthcare workers	1,107	11.8 (8–15.6)	4.9 (1.9–7.9)	16 (11.4–20.6)	0.99 (0.54–1.72)
		Outpatient department	2,632	14.8 (12.1–17.5)	13 (10.3–15.6)	26 (22.6–29.5)	1.83 (1.24–2.78)
		Pregnant women	2,555	12.4 (9.8–14.9)	4.1 (2.2–5.9)	16.1 (13.1–19.1)	1
		None	529	18.3 (11.9–24.7)	15.8 (9.3–22.3)	31.3 (23.2–39.4)	1.38 (0.85–2.14)
Pre-existing medical conditions	More than one	1,941	17.1 (13.8–20.4)	17.5 (14–20.9)	32.4 (28.2–36.7)	1.45 (1.1–1.91)	
	One	13,154	14.9 (13.7–16.1)	11.2 (10–12.3)	24.8 (23.3–26.3)	1	
	None	803	15.7 (10.7–20.6)	35.6 (28.7–42.5)	48.9 (41.6–56.2)	3.39 (2.32–5.01)	
Symptoms	More than one	3,423	15.9 (13.5–18.4)	20.6 (17.8–23.4)	34.4 (31.1–37.7)	1.86 (1.47–2.36)	
	One	11,398	15.1 (13.8–16.4)	8 (7–9.1)	22 (20.4–23.5)	1	
	None						

<sup>a</sup> Included only samples that were mapped to individuals.

<sup>b</sup> All estimates were adjusted for sensitivities and specificities of the RAT, RT-PCR and antibody testing kits and procedures. The assumed values were RAT sensitivity 0.5, specificity 0.975; RT-PCR sensitivity 0.95, specificity 0.97; IgG ELISA kit sensitivity 0.921, specificity 0.977 Weighted estimates for Karnataka estimated the prevalence in each unit and then weights according to population.

<sup>c</sup> Markets, malls, retail stores, bus stops, railway stations, waste collectors.

<sup>d</sup> Some individuals recruited in the moderate-risk and low-risk categories were moved to high-risk because of age or comorbidities.

estimated 89,17,539 (95% CI 81,39,023–97,66,828) people had active infection (on 16 September 2020). This was based on the numbers that tested positive on RT-PCR/RAT and after taking into account the IgG outcomes and the serial sensitivities and specificities of all tests (Table 1).

#### Total prevalence of COVID-19

The adjusted total prevalence of COVID-19 at state level was 27.7% (95% CI 26.1–29.3) as of 16 September 2020 (combined IgG and active infection (Table 1)).

#### Stratifications

The seroprevalence of IgG among males and females was similar, but the active infection was higher in males than females (15.5% vs. 8.4%) (Table 1). Thus, the total prevalence was higher in males than in females (29.8% vs. 21.9%). Estimates of both seroprevalence and total prevalence were higher in the elderly population and lower among those aged <30 years. The high-risk population had a higher prevalence (31.7%; 95% CI 29.1–34.2), followed by the moderate-risk population (25.4%; 95% CI 23.0–27.8) and then the low-risk population (20.7%; 95% CI 18.4–23.0) (Table 1).

#### Case-to-infection ratio (CIR)

At the state level, it was estimated that there were 40 infected individuals for every RT-PCR-confirmed case detected as of 16

September 2020 (Table 2). This was estimated by using 484,954 reported number of cases in Karnataka (Ministry of Health and Family Welfare Services, 2020) and the adjusted prevalence of COVID-19 (27.7%) against SARS-CoV-2. The cases-to-infections ratio ranged between 11–112 across units.

#### Infection fatality rate (IFR)

As of 03 September 2020, the IFR due to COVID-19 in Karnataka was estimated as 0.05%, with more than half of the units (21 of 38) above state IFR; the highest was estimated in the Dharwad district (0.21%) (Table 2 and Figure 2).

#### District/unit variations across the state

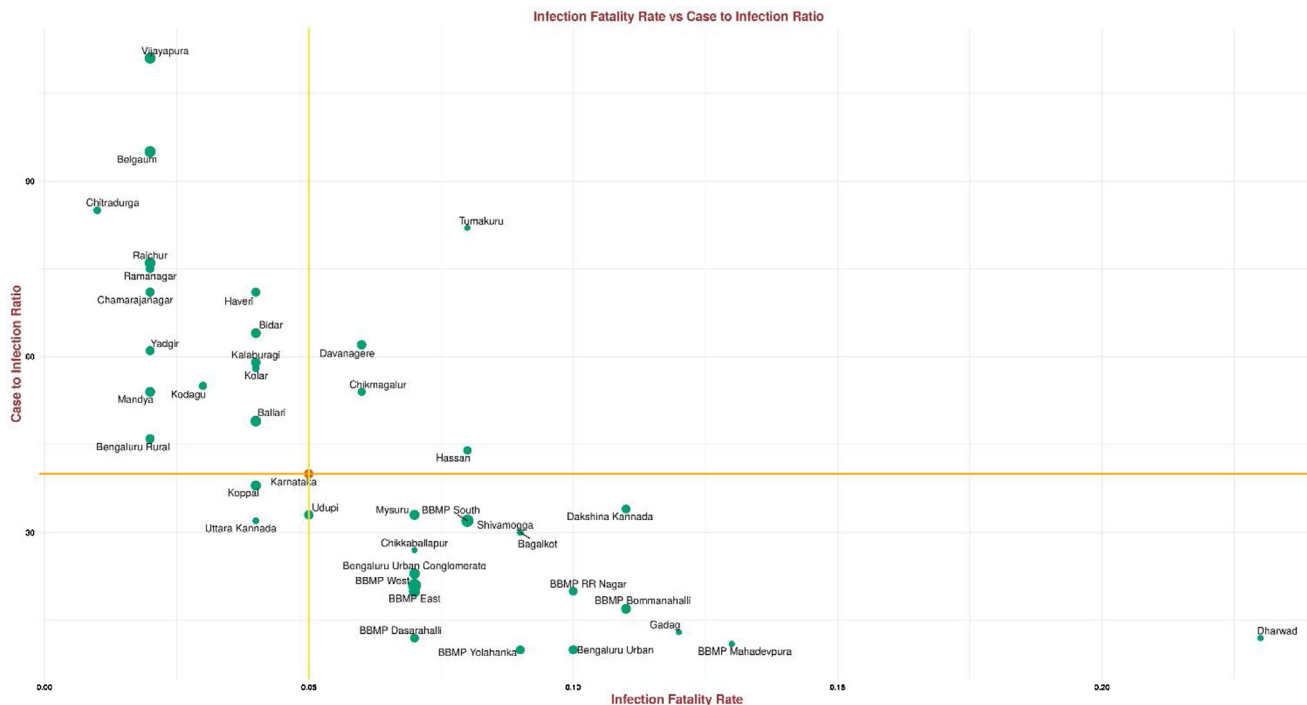
The IgG prevalence was highest in Vijayapura district (24.1%) and lowest in Bagalkot district (4.4%). The state capital Bengaluru had an IgG prevalence of 22.4% (95% CI 19.6–25.3). The active infection fraction was highest in Ballari (34.3%) and lowest in Bidar (0.7%). Bengaluru's active infection fraction was an estimated 9.2% (95% CI 7.1–11.2). The overall COVID-19 prevalence was lowest in Dharwad district (9.2%) and highest in Ballari district (43.5%) (Table 2). The total COVID-19 prevalence in Bengaluru was estimated to be 30.1% (95% CI 26.9–33.3). Within Bengaluru itself (3,617 samples), it was estimated that BBMP West had the highest IgG against SARS-CoV-2 and total prevalence of COVID-19. In contrast, BBMP Mahadevapura had the least (Supplementary Table 1). Again, BBMP RR Nagar had the highest active infection fraction within Bengaluru and BBMP East

**Table 2**  
Seroprevalence of IgG antibodies against SARS-CoV-2 and active infection in districts of Karnataka state (N = 15,624).

Unit	Samples <sup>a</sup>	% IgG against SARS-CoV2 <sup>b</sup>	% active infection <sup>b</sup>	% prevalence of COVID-19 <sup>b</sup>	CIR	IFR
Karnataka	15,624	16.8 (15.5–18.1)	12.6 (11.5–13.8)	27.7 (26.1–29.3)	1 : 40	0.05%
Ballari	406	22.7 (14.9–30.4)	34.3 (25.2–43.3)	43.5 (34–53)	1 : 49	0.04%
Davanagere	412	16.8 (9.8–23.7)	29.1 (20.2–38)	40.9 (31.3–50.5)	1 : 62	0.06%
Udupi	439	17 (10.3–23.7)	22.6 (15–30.2)	37 (28.2–45.9)	1 : 34	0.05%
Vijayapura	381	24.1 (15.9–32.4)	13.8 (6.6–21.1)	35.6 (25.9–45.3)	1 : 112	0.02%
Raichur	404	23.1 (15.2–30.9)	12.1 (5.4–18.7)	34.3 (25–43.6)	1 : 76	0.02%
Chikmagalur	436	12.3 (6.3–18.4)	20.9 (13.1–28.7)	32.1 (23.1–41.1)	1 : 54	0.06%
Yadgir	422	15.7 (9–22.5)	18.5 (11.2–25.9)	31.9 (23–40.8)	1 : 62	0.02%
Hassan	410	13.6 (7.1–20.1)	21.1 (12.8–29.4)	31 (21.7–40.3)	1 : 44	0.08%
Belgaum	430	23.9 (16.2–31.7)	6.4 (1.4–11.4)	30.3 (21.6–39)	1 : 95	0.02%
Kalaburagi	425	17.4 (10.5–24.4)	14.4 (7.8–21)	30.1 (21.4–38.7)	1 : 60	0.04%
Bengaluru Urban Conglomerate	3617	22.4 (19.6–25.3)	9.2 (7.1–11.2)	30.1 (26.9–33.3)	1 : 23	0.07%
Ramanagar	408	14.2 (7.5–20.8)	16.1 (8.7–23.6)	29.6 (20.4–38.7)	1 : 76	0.02%
Tumakuru	429	7 (1.9–12)	25.1 (16.2–34.1)	29.5 (20–39)	1 : 82	0.08%
Bengaluru Rural	432	15.6 (9–22.2)	16.4 (9–23.8)	29 (20.2–37.9)	1 : 46	0.02%
Haveri	417	15 (8.3–21.7)	14.5 (7.8–21.3)	28.8 (20.1–37.6)	1 : 72	0.04%
Mysuru	402	19.2 (11.9–26.6)	8.4 (2.7–14)	27.6 (18.9–36.3)	1 : 34	0.07%
Dakshina Kannada	430	15 (8.5–21.6)	13.5 (6.9–20)	27.4 (19–35.9)	1 : 34	0.11%
Chitradurga	411	10.3 (4.4–16.3)	16 (8.5–23.4)	26 (17.2–34.9)	1 : 85	0.01%
Mandya	414	18.8 (11.4–26.1)	6.7 (1.3–12.1)	25.5 (16.8–34.2)	1 : 54	0.02%
Koppal	427	20 (12.7–27.2)	2.6 (0–6.2)	22.6 (14.7–30.5)	1 : 38	0.04%
Shivamogga	426	8.1 (2.9–13.4)	13.7 (6.8–20.5)	21.8 (13.5–30)	1 : 31	0.09%
Chamarajanagar	383	16 (8.8–23.1)	6.6 (1.1–12.1)	21.3 (12.9–29.7)	1 : 72	0.02%
Kodagu	412	12.3 (6–18.5)	8.7 (2.8–14.5)	20.7 (12.6–28.9)	1 : 56	0.03%
Bidar	407	18.2 (10.9–25.4)	0.7 (0–3.3)	18.9 (11.2–26.5)	1 : 64	0.04%
Uttara Kannada	419	8.4 (3–13.8)	8.7 (3–14.4)	16.6 (9.1–24.1)	1 : 33	0.04%
Kolar	431	10.3 (4.5–16.1)	6.7 (1.6–11.9)	16.3 (9–23.6)	1 : 59	0.04%
Chikkaballapur	412	6.7 (1.6–11.8)	5.8 (0–11.8)	12.4 (4.8–20)	1 : 28	0.07%
Bagalkot	401	4.4 (0–8.9)	9.7 (3.6–15.8)	12.3 (5.3–19.4)	1 : 31	0.08%
Gadag	341	6.8 (1.3–12.3)	2.7 (0–8.5)	9.5 (1.6–17.4)	1 : 14	0.11%
Dharwad	440	7.6 (2.5–12.6)	2 (0–5.5)	9.2 (3.2–15.2)	1 : 13	0.21%

<sup>a</sup> Included only samples that were mapped to individuals; CIR: case-to-infection ratio; IFR: infection fatality rate.

<sup>b</sup> Adjusted for sensitivities and specificities of RAT, RT-PCR, and antibody testing kits and procedure.



**Figure 2.** Scatter plot of CIR versus IFR. The size of the point indicates the IgG prevalence in the units. The horizontal and the vertical lines intersect at Karnataka's IFR and CIR. Moving clockwise from the upper-left quadrant: a unit with a larger green disk had high IgG antibody prevalence, low IFR and high CIR, such a unit was missing cases and deaths; a unit with a larger green disk in the upper-right quadrant had high IgG antibody prevalence, high IFR and high CIR, such a unit was also likely missing cases but death reporting was better than average; a unit with a larger green disk in the bottom-right quadrant had high IgG antibody prevalence, high IFR and low CIR, such a unit did well in identifying cases and had better-than-average reporting of deaths; a unit with a larger green disk in the bottom left had low IFR and low CIR, such a unit saw a surge in cases but did well in identifying cases and had low fatality rates, perhaps due to good clinical practices that could be studied and replicated elsewhere.

had the lowest (Supplementary Table 1). Districts with high case infection ratios (>40) were Vijayapura, Belgaum, Chitradurga, Tumakuru, Raichur, Ramanagar, Haveri, Chamarajanagar, Bidar, Davanagere, Yadgir, Kalaburagi, Kolar, Kodagu, Mandya, Chikmagalur, Ballari, Bengaluru Rural, and Hassan (Table 2). To summarise: there was differential exposure to the disease across the state.

### Explanatory variables

A generalised linear model-based multinomial regression for symptoms indicated that headache, chest pain, wheezing, rhinorrhoea, cough, sore throat, muscle ache, fatigue, chills, and fever were significant variables that predicted active infection, with fever being the most significant. Diarrhoea, chest-pain, rhinorrhoea, fatigue, and fever predicted the presence of IgG antibodies to some extent, with diarrhoea having the highest weight. A second generalised linear model-based multinomial regression yielded additional variables that predicted active infection. These were attendance at the outpatient department of the healthcare facilities and contact with COVID-19-positive patients. Additional variables that predicted the presence of IgG were professions who had more contact with the public, residence in containment zones and the urbanisation level of the district (Tables 3 and 4).

### RAT versus RT-PCR sensitivity for symptomatic and asymptomatic people

As of September 2020, it was noted that RAT was more sensitive for symptomatic individuals. For participants with symptoms, 543 RAT were positive out of the 798 RT-PCR-confirmed positive participants. This yielded a sensitivity of 68.0% for those with symptoms. In contrast, for participants without symptoms, 348 RAT were positive out of 742 RT-PCR-confirmed positive participants. This yielded a sensitivity of 46.9% for those without symptoms.

## Discussion

The study involved healthcare workers and several state and central agencies across the entire state of Karnataka and was based around 290 healthcare facilities. It addressed the important question of measuring past and active infection of COVID-19 in Karnataka. Further, this was the first study in India, and probably

elsewhere, that jointly estimated the proportion of people who already had the SARS-CoV-2 infection (IgG antibody positive) and who currently had an active infection (RT-PCR/RAT positive). The combined estimates could lead to timely, informed and evidence-based public health responses.

The survey, although sentinel-based, sampled from the general population in the moderate-risk and high-risk groups. The low-risk group sampled from the facilities were restricted to healthcare workers and patients coming for periodic care or visiting outpatient departments. Further, the survey ensured good geographical spread across Karnataka. The design was thus aimed at capturing as representative a population as possible. The sampling frame serves as a reference standard and could be used for population-representative surveillance in the future. A serological test for IgG with high sensitivity (0.921) and specificity (0.977) was used, thereby yielding a better predictive value for a positive test.

An estimation of the IgG prevalence alone would have assessed the state's burden at 16.8% prevalence. In contrast, the dual assessment of viral markers and antibodies gave the IgG prevalence and also the active infection fraction of 12.6% and a total COVID-19 burden of 27.7%. This significantly larger estimate calls for an entirely different response from the state and highlights this survey's benefits. Moreover, 2.1% of the population showed both viral RNA and IgG antibodies. The correlates and implications of the simultaneous presence of viral RNA and IgG antibodies might require further examination in future studies.

The statewide CIR was 40 infections for every case. The number of reported cases up to 03 September 2020 was 361,305; another 123,649 reported cases were added between 03 September and 16 September. Clearly there was a surge during this 2-week period and 34% of reported cases were added. This suggests that the active infection fraction during this period should have been 34% of 16.8% estimated IgG level (i.e., 5.7%); however, the active infection fraction from the survey was as high as 12.6%. The CIR for the period prior to 03 September 2020 was 32 versus a CIR of 72 during the survey period, suggesting that a greater number of cases were missed during the surge in the first half of September 2020. The overall CIR of 40 is to be understood as a weighted average of the estimates for the two periods.

Public health responses and control measures in India have been largely shaped by case-based evidence, namely: data from the line list and contact tracing. For example, in a study from Tamil Nadu and Andhra Pradesh, Laxminarayan et al. showed that the

**Table 3**  
Generalized linear model: prediction of active, IgG and simultaneous IgG and active infection.

Predictor	$\beta_1$	$\hat{\sigma}_1$	Active p-value <sup>@</sup>	$\beta_2$	$\hat{\sigma}_2$	IgG p-value <sup>@</sup>	$\beta_3$	$\hat{\sigma}_3$	Active and IgG p-value <sup>@</sup>	$\beta_L$	$\hat{\sigma}_L$	Logistic p-value <sup>@</sup>
Intercept	-2.4	0.044	***	-1.7	0.031	***	-4.1	0.12	***	-1.1	0.021	***
Diarrhoea	0.54	0.48	.	1	0.32	***	1.4	0.85	.	0.79	0.25	**
Abdominal pain	-3	3.8	.	-0.05	0.25	.	-0.11	0.85	.	-0.23	0.18	.
Vomiting	0.56	0.4	.	-0.023	0.39	.	-6.2	24	.	0.18	0.24	.
Headache	0.56	0.16	***	-0.032	0.18	.	0.34	0.46	.	0.23	0.1	*
Other respiratory symptoms	0.37	0.52	.	0.33	0.39	.	-6.4	33	.	0.27	0.28	.
Chest pain	0.68	0.23	**	0.55	0.21	**	-0.51	1.3	.	0.46	0.14	**
Wheezing	1	0.46	*	-0.089	0.58	.	1.3	0.86	.	0.37	0.31	.
Shortness of breath	0.62	0.49	.	0.12	0.59	.	0.37	1.5	.	0.48	0.34	.
Runny nose	0.95	0.27	***	0.55	0.28	*	-7.4	32	.	0.56	0.19	**
Cough	0.84	0.086	***	0.13	0.09	.	0.3	0.28	.	0.41	0.054	***
Sore throat	0.71	0.37	*	-0.078	0.46	.	1	0.8	.	0.32	0.25	.
Muscle ache	0.67	0.18	***	0.033	0.2	.	0.25	0.57	.	0.25	0.12	*
Fatigue	0.68	0.25	**	0.42	0.23	*	-2.4	7.8	.	0.42	0.16	**
Chills	0.77	0.21	***	-0.35	0.3	.	0.47	0.61	.	0.19	0.15	.
Fever	1.5	0.085	***	0.21	0.11	*	1.4	0.23	***	0.8	0.058	***

@ \*\*\* indicates  $p < 0.001$ ; \*\* indicates  $p < 0.01$ ; \* indicates  $p < 0.05$ ; . indicates  $p < 0.1$ .

**Table 4**  
Generalized linear model: prediction of active, IgG and simultaneous IgG and active infection.

Predictor	$\beta_1$	$\hat{\sigma}_1$	Active p-value <sup>@</sup>	$\beta_2$	$\hat{\sigma}_2$	IgG p-value <sup>@</sup>	$\beta_3$	$\hat{\sigma}_3$	Active/IgG p-value <sup>@</sup>	$\beta_L$	$\hat{\sigma}_L$	Logistic p-value <sup>@</sup>
Intercept	-3.3	0.19	***	-2.6	0.13	***	-6.6	0.76	***	-1.8	0.081	***
Chronic liver disease	-0.59	1		0.38	0.66		0.54	1.5		-0.012	0.47	
Chronic renal disease	-7.3	82		-0.03	0.56		-5.6	22		-0.53	0.47	
Diabetes	0.075	0.12		-0.033	0.11		0.034	0.29		0.013	0.068	
Heart disease	-0.17	0.4		0.42	0.27	.	-0.21	1.1		0.15	0.2	
Hypertension	0.14	0.12		-0.072	0.11		0.13	0.3		0.04	0.071	
Immunocompromised condition	-0.44	0.45		-0.6	0.4	.	-1.3	2.2		-0.46	0.23	*
Malignancy	1	0.85		0.46	0.93		-4.6	32		0.53	0.61	
High-risk	0.59	0.29	*	-0.11	0.33		-1	1.3		0.14	0.19	
Moderate-risk	0.41	0.35		-0.56	0.37	.	-0.22	1.9		-0.075	0.23	
OPD attendee	0.67	0.19	***	0.078	0.11		0.93	0.57	.	0.23	0.075	**
Bus conductor or auto driver	0.28	0.41		0.74	0.4	*	0.6	1.9		0.33	0.24	
In containment zone	0.5	0.39		0.74	0.39	*	0.79	1.9		0.43	0.24	.
Healthcare worker	-1	0.45	*	0.27	0.39		-0.29	1.9		-0.25	0.24	
In congregate setting	0.32	0.4		0.52	0.39	.	0.59	1.9		0.29	0.24	
Comorbidity	0.04	0.33		0.37	0.34		1.8	1.4	.	0.17	0.2	
Elderly	0.35	0.34		0.34	0.34		1.4	1.4		0.21	0.2	
Vegetable vendor	0.33	0.4		0.7	0.39	*	0.37	2		0.33	0.24	
Age 30–39	0.21	0.11	*	0.24	0.084	**	0.95	0.42	*	0.2	0.055	***
Age 40–49	0.53	0.11	***	0.17	0.098	*	1.2	0.44	**	0.3	0.062	***
Age 50–59	0.64	0.13	***	0.34	0.11	***	1	0.48	*	0.42	0.07	***
Age 60+	0.31	0.15	*	0.26	0.13	*	1.5	0.51	**	0.29	0.083	***
Male	0.45	0.078	***	0.052	0.063		0.02	0.2		0.19	0.041	***
Other	-0.86	1.8		-28	640000		-14	2300		-1.6	1	
Urban or rural	0.39	0.12	***	0.36	0.11	***	1.2	0.63	*	0.32	0.065	***
Contact with a positive patient	0.75	0.096	***	0.11	0.11		0.77	0.27	**	0.39	0.063	***
Urbanization	-0.56	0.15	***	0.68	0.1	***	0.24	0.37		0.14	0.073	.

@ \*\*\* indicates  $p < 0.001$ ; \*\* indicates  $p < 0.01$ ; \* indicates  $p < 0.05$ ; . indicates  $p < 0.1$ .

infection probabilities ranged from 4.7–10.7% for low-risk and high-risk contacts (Laxminarayan et al., 2020). Kumar et al. showed that both asymptomatic and symptomatic SARS-CoV-2 cases transmitted the infection, with the symptomatic cases being the main driving force within Karnataka state (Kumar et al., 2021). However, the current seroprevalence-based estimates did not suffer from selection bias, as the sampling frame included the general population and went beyond reported cases and contact tracing. Therefore, the current study complements the understanding by providing better estimates of the actual burden of infection in the community.

Across risk groups, the elderly and those with comorbidities had a higher prevalence of COVID-19 (Hu et al., 2020; CDC COVID-19 Response Team, 2020; Guan et al., 2020), suggesting that they are at higher risk of contracting the infection. Despite the expected lower exposure, higher prevalence in them offers the possibility of infection with lower viral dose or that younger age groups have a protective immune response.

The reported IFR due to COVID-19 of 0.05% was likely an underestimate and a function of how well each district reported death data. Studies worldwide have found that the IFR of COVID-19 ranged from 0.17–4.16% (Rinaldi and Paradisi, 2020; Bendavid et al., 2021; Lewis and Torgerson, 2012). A high estimated IFR of 0.21% was seen in Dharwad. Comparing Dharwad (population 1.8 million, IgG level 7.6%, 113 days to survey start date since 50 cases, urbanisation index = 0.56, deaths = 337) with Shivamogga (population 1.7 million, estimated IgG 8.1%, 106 days to survey start date since 50 cases, urbanisation index = 0.30, deaths = 134), 200 more deaths were seen in Dharwad for roughly similar populations, IgG levels and days since 50 cases. The higher deaths could have been due to reporting differences or issues related to clinical practice or travel from neighbouring units to avail critical or tertiary health care facilities at Dharwad. Further research is required to test these hypotheses.

This regression analysis determined which symptoms accurately predict active and past infections. Among symptoms, diarrhoea, chest pain, rhinorrhoea, fatigue, and fever predict the presence of IgG antibodies. This suggests that COVID-19 may have consequences that last beyond the active infection period. Diarrhoea suggests that the gastrointestinal tract manifestations might stay longer, which may have implications to explore with oral vaccines. Influenza-like-illness symptoms and history of contact with a COVID-19-positive person are strong predictors of active infection.

Recruitment of the low risk participants from the facilities may suggest a bias in the estimate. This was mitigated by systematically sampling only among pregnant women and attendees of outpatient clinics. A design effect of 3 was used to account for any possible bias. The low-risk participants were not administered the RAT due to the test's low sensitivity. The statistical methodology handled such partial test administrations. The number of samples per unit of 432 was based solely on a nominal 10% IgG prevalence assumption and a requirement of a 5% margin of error. The actual 95% CI depends on the true prevalence. The reported 95% CIs were based on the inferred estimates and made use of all the available test data (IgG, RT-PCR and RAT outcomes).

This sentinel-based state-wide comprehensive population survey and the associated comparative analysis gave insight into the state of the pandemic in the different districts of Karnataka and the varying levels of prevalence across the different stratifications based on age, sex and risk. Figure 2 shows important epidemiological metrics such as IFR, CIR and their variation across geographical regions and population strata. This shows which units were likely to be missing cases but had better-than-average death reporting, which units did well in identifying cases and had better-than-average reporting of deaths, or which units had not seen a surge in cases, yet had done well in identifying cases and had low IFR. Testing strategies could be tuned to local variation, based



on each unit's total prevalence and CIR. Thus, establishing district-level sentinel-based surveillance to systematically monitor the trend of infection in the long term, to inform local decision-making at a district level, would facilitate and augment the necessary public health response towards the COVID-19 epidemic in Karnataka. It would also help to identify regions with high severity of the disease, identify at-risk populations and enable evidence-based intervention and resource allocation to effectively manage the pandemic. Repetition of the survey can better inform changes in the extent and speed of transmission and help evaluate the potential impact of containment strategies over time in different parts of the state.

The question of designing facility-level public health responses for COVID-19, rather than district-level, immediately arises. As the districts are the first-level administrative units for COVID-19 management, the survey was designed to be tight for a 5% margin of error and a 95% CI at district level. Facility-level responses can be designed if there is sufficient reliability on the estimates at each facility.

The sentinel-based population survey strategy for monitoring COVID-19 through healthcare facilities has great relevance for global public health. Sentinel surveillance systems enable identification of trends over time and are also easier to manage in terms of planning logistics, using existing human resources and are less cost-intensive. These enable timely implementation of serological studies. Team et al. reported a sentinel survey conducted by the Indian Council of Medical Research in the early stages of the pandemic (ICMR COVID Study Group et al., 2020). Such methodologies could be included by the WHO in its Solidarity II global serological study protocol (World Health Organization, 2020). This study's findings hold significant potential to improve clinical management and guide public health interventions to reduce the burden of COVID-19 in India and other lower and middle-income countries.

## Contributors

The survey was a collaborative effort of the Department of Health and Family Welfare, National Institute of Mental Health and Neuro-Sciences, Indian Institute of Public Health - Bangalore, Indian Institute of Science, Indian Statistical Institute (Bangalore Centre), UNICEF, MS Ramaiah Medical College, Bangalore Medical College, and others. Giridhara R Babu, R Lalitha, Lalitha Krishnappa, Pradeep Banandur, and Kumar DE Vasanth designed the protocol. Mysore Kalappa Sudarshan, CN Manjunath, Gopalkrishna Gururaj, Timmanahalli Sobagaiah Ranganath, Asish Satapathy, Lokesh Alahari, and Anita Desai reviewed and provided feedback on the design, implementation of the survey, analyses, and helped articulate the findings. Jawaid Akhtar and Pankaj Kumar Pandey reviewed the protocol, led the implementation of the survey and identified district-level public health responses. M Rajagopal Padma, Mohammed Shariff and Parimala S. Maroor coordinated the implementation across the state and reviewed the manuscript. Deepa Ravi, Shilpa Shiju, Prameela Dinesh, Vinitha Thakar, and Eunice Lobo developed the detailed protocol and standard operating procedures (protocol manual), implemented the study and reviewed the manuscript. Giridhara R. Babu, Siva Athreya and Rajesh Sundaresan planned and executed the data analysis, arrived at the initial findings and drafted the manuscript. Ambica Rangaiah, Ashok Munivenkappa, Krishna S, Shantala Gowdara Basawarajappa, HG Sreedhara, Siddesh KC, Amrutha Kumari B, Nawaz Umar, Mythri BA developed the lab protocols and provided the test results. Ravi Vasanthapuram designed and developed the protocol and revised the manuscript. All authors reviewed and approved the final manuscript.

## Conflict of interest

We declare no competing interests.

## Data sharing

The data are accessible to researchers upon formal request for data addressed to the Commissioner, Health and Family Welfare Services, Government of Karnataka.

## Role of funding source

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## Appendix A. Supplementary data

Supplementary material related to this article can be found, in the online version, at doi:<https://doi.org/10.1016/j.ijid.2021.05.043>.

## References

- Aarti N, Aurnab G, LS S. Epidemiological and serological surveillance of COVID-19 in Pune city organizations India; Pune. . p. 1–4.
- Bendavid E, Mulaney B, Sood N, Shah S, Bromley-Dulfano R, Lai C, et al. COVID-19 antibody seroprevalence in Santa Clara County, California. *Int J Epidemiol* 2021; dyab010.
- CDC COVID-19 Response Team. Preliminary estimates of the prevalence of selected underlying health conditions among patients with coronavirus disease 2019—United States, February 12–March 28, 2020. *MMWR Morb Mortal Wkly Rep* 2020;69(13):382.
- Chatterjee S, Sarkar A, Karmakar M, Chatterjee S, Paul R. How the asymptomatic population is influencing the COVID-19 outbreak in India?. *arXiv preprint arXiv* 2020;1–12 2006.03034.

- Financial Express. Sero-prevalence survey Delhi: here's why survey result in the capital is 'remarkable'. 2021.
- Guan W-J, Liang W-H, Zhao Y, Liang H-R, Chen Z-S, Li Y-M, et al. Comorbidity and its impact on 1590 patients with Covid-19 in China: a nationwide analysis. *Eur Respir J* 2020;55(5).
- Hallal PC, Hartwig FP, Horta BL, Silveira MF, Struchiner CJ, VIDALETTI LP, et al. SARS-CoV-2 antibody prevalence in Brazil: results from two successive nationwide serological household surveys. *Lancet Glob Health* 2020;8(11):e1390–e8.
- Hellewell J, Abbott S, Gimma A, Bosse NI, Jarvis CI, Russell TW, et al. Feasibility of controlling COVID-19 outbreaks by isolation of cases and contacts. *Lancet Glob Health* 2020;e488–96.
- Herzog S, De Bie J, Abrams S, Wouters I, Ekinci E, Patteet L, et al. Seroprevalence of IgG antibodies against SARS coronavirus 2 in Belgium: a prospective cross-sectional study of residual samples. *medRxiv* 2020;1–15.
- Hu Y, Sun J, Dai Z, Deng H, Li X, Huang Q, et al. Prevalence and severity of corona virus disease 2019 (COVID-19): a systematic review and meta-analysis. *J Clin Virol* 2020;104371.
- ICMR COVID Study Group, COVID Epidemiology & Data Management Team, COVID Laboratory Team, VRDLN Team. Laboratory surveillance for SARS-CoV-2 in India: performance of testing & descriptive epidemiology of detected COVID-19, January 22–April 30, 2020. *Ind J Med Res* 2020;151(May (5)):424.
- Korth J, Wilde B, Dolf S, Anastasiou OE, Krawczyk A, Jahn M, et al. SARS-CoV-2-specific antibody detection in healthcare workers in Germany with direct contact to COVID-19 patients. *J Clin Virol* 2020;104437.
- Kumar N, Hameed SKS, Babu GR, Venkataswamy MM, Dinesh P, Bg PK, et al. Descriptive epidemiology of SARS-CoV-2 infection in Karnataka state, South India: transmission dynamics of symptomatic vs. asymptomatic infections. *EClinicalMedicine* 2021;32:100717.
- La Marca A, Capuzzo M, Paglia T, Roli L, Trenti T, Nelson SM. Testing for SARS-CoV-2 (COVID-19): a systematic review and clinical guide to molecular and serological in-vitro diagnostic assays. *Reprod Biomed Online* 2020;483–99.
- Laxminarayan R, Wahl B, Dudala SR, Gopal K, Neelima S, Reddy KJ, et al. Epidemiology and transmission dynamics of COVID-19 in two Indian states. *Science* 2020;370(6517):691–7.
- Lewis FI, Torgerson PR. A tutorial in estimating the prevalence of disease in humans and animals in the absence of a gold standard diagnostic. *Emerg Themes Epidemiol* 2012;9(1):1–8.
- Malani A, Shah D, Kang G, Lobo GN, Shastri J, Mohanan M, et al. Seroprevalence of SARS-CoV-2 in slums versus non-slums in Mumbai, India. *Lancet Glob Health* 2021;9(2):e110–1.
- Meyerowitz-Katz G, Merone L. A systematic review and meta-analysis of published research data on COVID-19 infection-fatality rates. *Int J Infect Dis* 2020;138–48.
- Ministry of Health and Family Welfare Services. #IndiaFightsCorona COVID-19 in India, Corona Virus Tracker. COVID-19 Dashboard. 2020. <https://www.mygov.in/covid-19>.
- Murhekar MV, Bhatnagar T, Selvaraju S, Rade K, Saravanakumar V, Thangaraj JWV, et al. Prevalence of SARS-CoV-2 infection in India: Findings from the national serosurvey, May–June 2020. *Ind J Med Res* 2020;152(1):48.
- Noh JY, Seo YB, Yoon JG, Seong H, Hyun H, Lee J, et al. Seroprevalence of anti-SARS-CoV-2 antibodies among outpatients in southwestern Seoul, Korea. *J Korean Med Sci* 2020;35(33).
- Rinaldi G, Paradisi M. An empirical estimate of the infection fatality rate of COVID-19 from the first Italian outbreak. *medRxiv* 2020;1–18.
- Rogan WJ, Gladen B. Estimating prevalence from the results of a screening test. *Am J Epidemiol* 1978;107(1):71–6.
- Sapkal G, Shete-Aich A, Jain R, Yadav PD, Sarkale P, Lakra R, et al. Development of indigenous IgG ELISA for the detection of anti-SARS-CoV-2 IgG. *Ind J Med Res* 2020;151(5):444.
- Siddiqui S, Naushin S, Pradhan S, Misra A, Tyagi A, Looma M, et al. SARS-CoV-2 antibody seroprevalence and stability in a tertiary care hospital-setting Available at SSRN 3696827. 2020.
- Statistics DoEa. Projected population of Karnataka (2012–2021). Government of Karnataka; 2013 p. 60.
- Stringhini S, Wisniak A, Piumatti G, Azman AS, Lauer SA, Baysson H, et al. Seroprevalence of anti-SARS-CoV-2 IgG antibodies in Geneva, Switzerland (SEROCoV-POP): a population-based study. *Lancet* 2020;395(10237):1587.
- The Lancet. The plight of essential workers during the COVID-19 pandemic. *Lancet* (London, England) 2020;395(10237):1587.
- To KK-W, Tsang OT-Y, Leung W-S, Tam AR, Wu T-C, Lung DC, et al. Temporal profiles of viral load in posterior oropharyngeal saliva samples and serum antibody responses during infection by SARS-CoV-2: an observational cohort study. *Lancet Infect Dis* 2020;565–74.
- World Health Organization. Solidarity II global serologic study for COVID-19. 2020 Accessed from: <https://www.who.int/emergencies/diseases/novel-coronavirus-2019/global-research-on-novel-coronavirus-2019-ncov/solidarity-2-global-serologic-study-for-covid-19>.
- Zou L, Ruan F, Huang M, Liang L, Huang H, Hong Z, et al. SARS-CoV-2 viral load in upper respiratory specimens of infected patients. *N Engl J Med* 2020;382(12):1177–9.