Characterizing Longitudinal Antibody Responses in Recovered Individuals Following COVID-19 Infection and Single-Dose Vaccination: A Prospective Cohort Study

Andrea D. Olmstead 1,2,†, Aidan M. Nikiforuk 2,3,†, Sydney Schwartz 2, Ana Citlali Márquez 2, Tahereh Valadbeigy 2, Eri Flores 2, Monika Saran 1, David M. Goldfarb 1,4, Althea Hayden 5, Shazia Masud 1,6, Shannon L. Russell 1,2, Natalie Prystajecky 1,2, Agatha N. Jassem 1,2, Muhammad Morshed 1,2 and Inna Sekirov 1,2,*

1 Department of Pathology and Laboratory Medicine, University of British Columbia, 2211 Wesbrook Mall, Vancouver, BC V6T 1Z7, Canada
2 British Columbia Centre for Disease Control Public Health Laboratory, Provincial Health Services Authority, 655 West 12th Ave, Vancouver, BC V5Z 4R4, Canada
3 School of Population and Public Health, University of British Columbia, 2206 E Mall, Vancouver, BC V6T 1Z3, Canada
4 Department of Pathology and Laboratory Medicine, British Columbia Children’s and Women’s Hospital, 4500 Oak Street, Vancouver, BC V6H 3N1, Canada
5 Office of the Chief Medical Health Officer, Vancouver Coastal Health, Vancouver, BC V5Z 4C2, Canada
6 Department of Pathology and Laboratory Medicine, Surrey Memorial Hospital, Surrey, BC V3V 1Z2, Canada
* Correspondence: inna.sekirov@bccdc.ca
† These authors contributed equally to this work.

Abstract: Background: Investigating antibody titers in individuals who have been both naturally infected with SARS-CoV-2 and vaccinated can provide insight into antibody dynamics and correlates of protection over time. Methods: Human coronavirus (HCoV) IgG antibodies were measured longitudinally in a prospective cohort of qPCR-confirmed, COVID-19 recovered individuals (k = 57) in British Columbia pre- and post-vaccination. SARS-CoV-2 and endemic HCoV antibodies were measured in serum collected between Nov. 2020 and Sept. 2021 (n = 341). Primary analysis used a linear mixed-effects model to understand the effect of single dose vaccination on antibody concentrations adjusting for biological sex, age, time from infection and vaccination. Secondary analysis investigated the cumulative incidence of high SARS-CoV-2 anti-spike IgG seroreactivity equal to or greater than 5.5 log10 AU/mL up to 105 days post-vaccination. No re-infections were detected in vaccinated participants, post-vaccination by qPCR performed on self-collected nasopharyngeal specimens. Results: Bivariate analysis (complete data for 42 participants, 270 samples over 472 days) found SARS-CoV-2 spike and RBD antibodies increased 14–56 days post-vaccination (p < 0.001) and vaccination prevented waning (regression coefficient, B = 1.66 [95%CI: 1.45–3.46]); while decline of nucleocapsid antibodies over time was observed (regression coefficient, B = −0.24 [95%CI: −1.2–(−0.12)]). A positive association was found between COVID-19 vaccination and endemic human β-coronavirus IgG titer 14–56 days post vaccination (OC43, p = 0.02 & HKU1, p = 0.02). On average, SARS-CoV-2 anti-spike IgG concentration increased in participants who received one vaccine dose by 2.06 log10 AU/mL (95%CI: 1.45–3.46) adjusting for age, biological sex, and time since infection. Cumulative incidence of high SARS-CoV-2 spike antibodies (>5.5 log10 AU/mL) was 83% greater in vaccinated compared to unvaccinated individuals. Conclusions: Our study confirms that vaccination post-SARS-CoV-2 infection provides multiple benefits, such as increasing anti-spike IgG titers and preventing decay up to 85 days post-vaccination.

Keywords: SARS-CoV-2; COVID-19; cohort study; antibody waning; seroreactivity; electrochemiluminescence assay; fixed-effect models
1. Introduction

The coronavirus disease 2019 (COVID-19) pandemic, caused by the novel beta (β)-coronavirus, severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2), has resulted in significant morbidity, mortality, economic impact, and disruption of health care and societal systems. Prior to the emergence of COVID-19, four seasonal human coronaviruses (HCoV) were identified that typically cause self-limited respiratory infections with mild symptoms, i.e., the ‘common cold’ in otherwise healthy people [1]. Like SARS-CoV-2, HCoV-OC43, and HCoV-HKU1 are β-coronaviruses, while HCoV-229E and HCoV-NL63 are classified as alpha (α)-coronaviruses [2]. Coronavirus genera are separated by unique serological and genomic characteristics; viral species from the same genus share cross-neutralizing (non-specific) antibodies which arise from homology in viral genes and structural proteins [3].

In the province of British Columbia (BC), Canada, the first confirmed case of COVID-19 was reported on January 25, 2020; strict and swift public health measures were largely effective at controlling spread during the first wave, which peaked locally between the third week of March and late April in 2020 [4]. During the first epidemiological wave of the pandemic, little was known about antibody responses to SARS-CoV-2 infection and studies were needed to understand if and how quickly infected individuals develop a detectable, protective, and durable antibody-mediated immune response. Understanding the durability or waning of antibodies over time helps elucidate the risk of re-infection and inform vaccination schedules. Studies have shown that most SARS-CoV-2 infected individuals seroconvert within 14–28 days; the spike (S) and the nucleocapsid (N) proteins elicit the strongest humoral response [5,6]. Predictably, SARS-CoV-2 antibody concentrations wane over time; the rate of decline varies widely depending on various factors (e.g., age, biological sex, and disease severity) [7–9]. Neutralizing antibodies acquired naturally or from vaccination protect against infection and re-infection [10]. Several studies have shown a strong correlation between anti-S, anti-receptor binding domain (RBD) and neutralizing antibody titers, as such measuring anti-S and anti-RBD can be used as a proxy for antibody-mediated protection [8,11,12].

Despite the success of SARS-CoV-2 vaccines, many individuals are still hesitant to be immunized against COVID-19; supply shortages combined with social and economic inequity hamper global vaccination efforts [13–15]. The study of antibody dynamics following natural infection and the impact of vaccination on those who have been previously infected is needed, as novel SARS-CoV-2 variants with increasing capacity to escape pre-existing immunity continue to evolve and spread [16–18]. Observational studies agree that vaccination benefits those who have been previously infected, but the number of doses required for optimal protection remains unclear [19–21].

We describe a prospective cohort that was established to monitor antibody responses over three months in people that recovered from SARS-CoV-2 infection. Many participants were offered a single dose COVID-19 vaccine during the study; therefore, we expanded the aims to consider the dynamics of antibodies against both SARS-CoV-2, as well as endemic HCoVs, in recovered individuals pre- and post- vaccination against SARS-CoV-2, and investigated their relationship with age, biological sex, and time from qPCR diagnosis.

2. Materials and Methods

2.1. Study Design

A prospective observational cohort termed CARE (Characterizing the Antibody Response to Emerging COVID-19) was established from individuals who recovered from SARS-CoV-2 infection, for the purposes of investigating antibody responses against several SARS-CoV-2 epitopes (full spike (S), receptor binding domain (RBD) and nucleocapsid (N)), as well as against the S protein of endemic HCoVs (OC43, HKU1, NL63, 229E), at least 2 weeks post natural SARS-CoV-2 infection, with or without subsequent SARS-CoV-2 vaccination. Information on age, biological sex and date of real time PCR (qPCR) diagnosis was collected through medical records, while information on the duration of COVID-19 symptoms, hospitalization, and vaccination was collected using an online self-reporting
survey. The date of SARS-CoV-2 qPCR diagnosis was used to estimate ‘days post-infection’.
Survey data and participant informed consent were collected and managed using REDCap
electronic data capture tools hosted at BC Children’s Hospital (Vancouver, BC). REDCap
(Research Electronic Data Capture) is a secure, web-based application designed to sup-
port data capture for research studies [22]. Participants were enrolled in the cohort from
19 November 2020, to 7 September 2021. During this time the most prevalent SARS-CoV-2
variant in British Columbia transitioned between the Alpha, Beta, Gamma and Delta geno-
types. The Beta variant was detected in late 2020 and January 2021, increase prevalence of
the Alpha variant shortly followed and it remained dominant until June 2020. The Gamma
variant was first detected in late February 2021, it’s incidence surpassed Alpha in July 2021.
Public health surveillance first recorded the Delta variant in March, and it was responsible
for most sequenced cases over the summer from July to September 2021 [23]. Vaccinated
CARE Study participants received any of the three SARS-CoV-2 glycoprotein-based vac-
cines approved by Health Canada during the study period: COMIRNATY (BioNTech
(Mainz, Germany), Pfizer (New York, NY, USA)), Spikevax (Moderna, Cambridge, MA,
USA) and Vaxzevria (Oxford, UK, Astra Zeneca) vaccine. All data analysis was performed
in R version 4.0.4 using the packages: ‘DataExplorer’, ‘survival’, ‘survminer’, ‘dplyr’,
‘ggfortify’, ‘tableone’, ‘naniar’, ‘RCColorBrewer’, ‘lme4’, ‘mgcv’, ‘gam.check’ and ‘readr’ [24].

2.2. Recruitment Criteria
Adults 18 years of age and older from the greater Vancouver metropolitan area were
recruited if they had a confirmed real time PCR (qPCR)-positive SARS-CoV-2 infection and
if they were no longer required to self-isolate per the BC provincial public health guidelines
(i.e., tested positive for SARS-CoV-2 at least 14 days prior). Initial diagnostic qPCR testing
was done in accordance with standard laboratory practices in British Columbia during the
time of the study. Either nasopharyngeal swab or saline gargles were acceptable sample
types. Diagnostic laboratories in the Lower Mainland area, where participants were tested,
used a variety of both commercial (GeneXpert Cepheid, Sunnyvale, CA, USA; Cobas®
Roche Diagnostics, Indianapolis, IN, USA; Panther Fusion Hologic, Marlborough, MA,
USA) and laboratory developed assay (E gene and RdRP gene) [25]. Only positive qPCR
results were allowed as recruitment criterion (i.e., potential participants with indeterminate
or invalid test results were excluded). The study protocol was approved by the University
of British Columbia (UBC) Clinical Research Ethics Board (H20-01089).

2.3. Sample Collection and Processing
Participants were required to donate 10 mL blood samples collected by venipuncture
(for serological testing) and concurrent self-collected saline gargle samples (for SARS-CoV-2
qPCR testing by a laboratory-developed assay targeting the E gene and RdRP genes) [25],
every two weeks for 3 months post-recruitment (up to 7 collections total). Blood was drawn
in gold-top separator tube with polymer gel (BD, cat# 367989); after at least 30 min
of clotting at room temperature, the blood sample was then centrifuged at 1400 g by staff at
the collection site and sent to the British Columbia Centre for Disease Control Public Health
Laboratory (BCCDC PHL). At the BCCDC PHL the samples were divided into serum
aliquots that were frozen at −80°C within four hours of receipt. Blood collections occurred
at four sites in the Greater Vancouver Area, British Columbia Canada: BC Children’s
Hospital, St. Paul’s Hospital, Abbotsford Regional Hospital and Surrey Memorial Hospital.
Saline gargle samples were self-collected by the participants at home, in accordance with
well-validated instructions [26], on the day of blood collection and transported to the
BCCDC PHL by the blood collection site. Self-collected saline gargle samples were tested
for SARS-CoV-2 by qPCR.
SARS-CoV-2 whole genome sequencing was done on all available participants’ SARS-
CoV-2 positive diagnostic clinical specimens; detailed methods have been described else-
where [27]. Samples were sequenced on an Illumina NextSeq instrument (San Diego,
CA, USA) using a tiled 1200bp amplicon scheme and analyzed using a modified ARC-
TIC Nextflow pipeline (https://github.com/BCCDC-PHL/ncov2019-artic-nf, accessed on 15 September 2022). Called variants were kept if the variant allele frequency was above 0.25 with ≥10x coverage. Sequences passing QC (85% genome completeness, 10x depth of coverage and no quality flags) were included in the phylogenetic analysis. A phylogenetic tree was constructed using Fasttree [28] and visualized in Nextstrain [29] and lineage assignment was performed using the Phylogenetic Assignment of Named Global Outbreak Lineages tool (Pango/Usher Version 1.15.1) [30]. All molecular, genomic, and serological testing for participant specimens (described below) was conducted centrally at the BCCDC PHL, a College of American Pathologists accredited laboratory.

2.4. Measurements of Humoral Immunity

Serum samples were initially tested using a combination of three Health Canada approved chemiluminescent immunoassays: (1) total antibodies to SARS-CoV-2 RBD (Siemens SARS-CoV-2 Total Assay [COV2T], Munich, Germany), (2) total antibodies to SARS-CoV-2 S (Ortho VITROS™ Anti-SARS-CoV-2 Total, Raritan, NJ, USA) and IgG anti-N antibodies (Abbott ARCHITECT™ SARS-CoV-2 IgG, Abbott Park, IL, USA), as per manufacturer guidelines, with results interpreted as reactive or non-reactive using the manufacturer-recommended signal to cut-off ratios [31,32]. All available samples were then tested using the V-PLEX COVID-19 Coronavirus Panel 2 (IgG) (Mesoscale Diagnostics LLC (MSD): #K15369U, Rockville, MD, USA), the diagnostic accuracy of the MSD assay was previously validated through comparison with alternative Health Canada approved tests at the BCCDC PHL [33]. The MSD assay provides quantitative measures of IgG antibodies against RBD, S and N SARS-CoV-2 epitopes, as well as IgG antibodies against the glycoprotein (S) of the four seasonal endemic HCoVs. Serological specimens were processed as previously reported [33]. Quantitative antibody levels expressed as log10 antibody units (AU)/mL were recorded and evaluated for all tested samples. MSD results were interpreted as reactive or non-reactive using the MSD recommended signal thresholds for serum: SARS-CoV-2 anti-S IgG = 1960; anti-N IgG = 5000; anti-RBD IgG = 538. Cutoffs (derived at the BCCDC PHL) for seasonal HCoVs seropositive status are as follows: HCoV-229E anti-S IgG = 1700; HCoV-HKU1 anti-S IgG = 900; HCoV-NL63 anti-S IgG = 270; HCoV-OC43 anti-S IgG = 2000 [34]. Samples were stratified by collection time to <6 months and ≥6 months post infection and percent positivity was compared using a Chi-square test ($\chi^2$ test).

2.5. Power Analysis for Investigating Association between IgG Concentration and Vaccination

A power calculation was conducted to determine the minimum number of paired participant samples needed to estimate at least a 70% association between COVID-19 vaccination and HCoV anti-IgG antibody concentration. Antibody concentrations were assumed to be normally distributed with a standard deviation of one. A significance level of 5% and two-sided alternative were used [35].

2.6. Analytic Data Selection

To analyze antibody dynamics, an analytic dataset was selected from the CARE COVID-19 cohort. At least k = 18 paired participant samples are required to estimate a 70% or greater association between COVID-19 vaccination and HCoV anti-IgG antibody concentration. Exclusion criteria were applied to select an analytic dataset from k = 57 participants with n = 341 observations. One participant had no follow up samples and was omitted from the analytic dataset (k = 1, n = 1). Six participants were excluded because they were vaccinated before collection of their baseline sample (k = 6, n = 37). Eight participants were removed from the analytic dataset because of missing data in their survey results (k = 8, n = 33). After applying the exclusion criteria, the analytic dataset contained k = 42 participants with n = 270 observations (Supplementary Figure S1). There were k = 41 participants with >1 pre-vaccine sample (n = 210 pre-vaccine observations) used for analysis of antibody waning pre-vaccination.
2.7. Bivariate Data Analysis

2.7.1. Antibody Waning

Waning of SARS-CoV-2 antibodies prior to vaccination was investigated in independent participant specimens measured at baseline using linear regression. HCoV anti-IgG antibody signals at baseline were compared to signals 14–56 days post-vaccination using a paired Wilcoxon signed-rank test in a sample of n = 21 participants who received a COVID-19 vaccine during the study. Waning of SARS-COV-2 specific IgG was measured prior to vaccination between participants using linear regression. Participant’s baseline samples (defined as the first specimen taken after enrolling in the study) were plotted for anti-S and anti-N IgG over time.

2.7.2. Descriptive Statistics

Bivariate analysis was conducted between the exposure (a single dose of a Health Canada approved COVID-19 vaccine) and outcome (SARS-CoV-2 anti-S or anti-N IgG) of interest at baseline. Baseline represents the time of a participant’s first blood draw after enrollment. The bivariate relationship between vaccine status and covariates was examined by t-test or Chi-square test ($\chi^2$ test) depending on variable type. HCoV anti-IgG antibody signals were transformed to the logarithmic base ten scale for conformation to normality and ease of interpretation.

2.8. Primary Analysis

Primary analysis used a multivariable linear mixed-effects model to regress SARS-CoV-2 anti-IgG concentration on vaccine status adjusting for dependency within participant samples and covariates defined as potential confounders by the common cause criterion [36,37]. Separate models were fit for anti-S and anti-N IgG signals. Unconditional mean models were used to find the intraclass correlation coefficient (ICC) before covariates were added to build fixed effect models [38]. Effect modification terms were assessed by the Akaike information criterion and included in the fixed effect models to understand if time from infection influences SARS-CoV-2 antibody concentrations [39].

2.9. Secondary Analysis

Secondary analysis employed a Kaplan–Meier curve to estimate the cumulative incidence of seroreactivity stratified by vaccine status. The survival function was transformed to cumulative incidence by $1-S(\tau)$ [40]. Seroreactivity was defined from the distribution of SARS-CoV-2 anti-S IgG concentration at first blood draw (baseline); the 95th percentile was chosen as the threshold (5.5 log10 AU/mL). Participants were censored if they were not seroreactive before loss to follow-up (right censoring). A log-rank test was used to test the hypothesis that the cumulative incidence of seroreactivity between unvaccinated and vaccinated persons, who have been previously naturally infected with SARS-CoV-2, does not differ [41].

3. Results

3.1. CARE COVID-19 Cohort

Fifty-seven individuals recovered from COVID-19 infection were recruited into the CARE COVID-19 Cohort. Recruited subjects (17 male, 40 female; 18 to 76 years old) represented a range of COVID-19 disease severity. Most subjects had a mild case of COVID-19, defined as not requiring hospitalization; 6 reported being asymptomatic and 12 reported experiencing fever. Only four of the recruited subjects (7%) reported being hospitalized for COVID-19; one required intensive care. The observed case severity distribution was consistent with the general distribution of COVID-19 disease severity in BC (~5% of diagnosed cases hospitalized as of April 2022) [42].

Participants were required to have recovered from COVID-19 (i.e., 14 days post-qPCR diagnosis) before providing their first blood and saline gargoyle sample. Collection dates ranged from 18–490 days (median 152 days) since a positive qPCR test (used as proxy for
time since infection), with the baseline collection date ranging from 18–339 days (median 114 days). Participants submitted between 1 and 7 samples, with approximately 2 weeks (median 14 days; range 7–83 days) between each collection, with an average of 6 samples collected per participant and a total of 341 samples collected. No reinfections or persistent virus shedding were detected in self-collected saline gargle samples using qPCR (data not shown).

Virus whole genome sequencing was performed on all available primary diagnostic specimens obtained from recruited participants [27] to determine the SARS-CoV-2 variant responsible for infection. Forty-one sequences were obtained (28 from the analytic dataset) (Supplementary Tables S2 and S3). SARS-CoV-2 variants were classified by pangolin lineage (Supplementary Tables S2 and S3) and visualized as a phylogenetic tree (Supplementary Figure S2). Viral genomes detected in the study sample are representative of variants circulating at the time of respective participants’ diagnoses [29]. Whole genome sequencing data was missing for ~33% of participants in the analytic dataset and, therefore, was not included as a covariate in the analysis. Multiple studies corroborate no significant difference in neutralising antibodies between the alpha variant and the ancestral isolate post mRNA vaccination with BNT162b2 or mRNA-1273. Noteworthy reduction of post-vaccination neutralising sera was observed for the beta variant in persons vaccinated with mRNA-1273 [43].

3.2. Comparison of Anti-SARS-CoV IgG Antibody Responses across Four Commercial Assays

All available samples (n = 340; 1 missing) were initially tested using a combination of three commercial serology assays supplied by Siemens (COV2T), Abbott (ARCHITECT™), or Ortho (VITROS™) clinical diagnostics. Of the n = 340 samples tested, n = 338 were classified as reactive using at least one assay (Supplementary Table S1). All available samples (n = 339) were subsequently tested using a highly sensitive and multiplex electrochemiluminescent assay offered by Meso Scale Diagnostics (MSD). Percent positivity differed across the platforms and by antigenic target. Overall detection of anti-S was more sensitive than anti-N SARS-CoV-2 IgG. Comparing anti-S results, the Ortho assay had the highest positivity rate (100%) followed by Siemens (95%) and MSD (89%) (Supplementary Table S1). For anti-N results MSD (58%) outperformed Abbott (47%) with a 11% increase in positivity ($\chi^2$ test, $p = 0.01$). When samples were stratified by collection time to less than or greater than 6 months post-infection, the anti-N positivity rate decreased for both the Abbott (72% to 13%) and MSD, (76% to 33%) ($p < 0.001$). A 7% decline in positivity was observed for anti-S ($p = 0.06$) and 2% for anti-RBD ($p = 0.53$) when tested by MSD (Supplementary Table S1). Only antibody measurements from the MSD assay were used in the multivariable analysis as the anti-S IgG results compared well with Ortho and anti-N IgG results were superior to Abbott. Waning of anti-S and anti-N IgG concentrations over time were measured between participants using the first baseline observation for each of the $k = 42$ participants in the analytic dataset. Using linear regression analysis, overall waning was observed in both anti-N and anti-S and the slope did not differ significantly across the two measures ($p = 0.46$; Figure 1). On average SARS-CoV-2 antibodies wane at a rate of $-0.0029$ log10 AU/mL per day ($p < 0.001$) or ~4228 AU/mL per month. These results confirm waning of anti-SARS-CoV-2 antibodies over time in people who have recovered from natural SARS-CoV-2 infection before vaccination. Estimates of anti-SARS-CoV-2 IgG waning are calculated post-vaccination using a mixed-effects linear regression model and reported as the ‘primary analysis’.
Figure 1. Longitudinal decay of SARS-CoV-2 anti-N and anti-S IgG concentration over time in natural SARS-CoV-2 infected CARE participants prior to vaccination (k = 42, n = 42 samples). Participant samples were restricted to the first collection date (baseline) and plotted independently. Linear regression was used to estimate the decrease in anti-S and anti-N titer over time since qPCR test result.  

3.3. Serological Response to SARS-CoV-2 Vaccination

Bivariate analysis was conducted on the analytic dataset to compare participant antibody responses pre- and post-vaccination for COVID-19. Participant’s serology results and survey responses are summarized and stratified at baseline by the exposure of interest, one dose of a COVID-19 vaccine (Table 1). No difference in the distribution of covariates between participants who received and did not receive a COVID-19 vaccine (p = 0.014) (Table 1). Importantly, age, biological sex, days from positive qPCR test (diagnosis), symptom duration and endemic anti-coronavirus IgG signals did not differ by exposure at baseline; therefore, we expect limited confounding from these covariates when estimating the association between COVID-19 vaccination and anti-SARS-CoV-2 IgG signals. Covariates, which met the definition of a confounder by the common cause criteria, were adjusted for in the primary analysis using a linear mixed effects model.

In k = 21 paired participants, SARS-CoV-2 anti-S and anti-RBD IgG antibody concentrations increased post vaccination by 1.63 (p ≤ 0.001) and 1.82 (p ≤ 0.001) log10 AU/mL (Figure 2A,B). Anti-N antibody concentration continued to decrease post vaccination by −0.3 (p = 0.03) log10 AU/mL (Figure 2C), consistent with waning observed prior to vaccination. Most participants (>99%) were found to be seropositive for anti-S antibodies against the endemic HCoVs. Post vaccination, anti-S antibody concentrations for endemic human β-coronaviruses HCoV-HKU1 and HCoV-OC43 increased (p = 0.02 and p = 0.02) (Figure 3B,D). No increase in antibody concentration was observed for the endemic human α-coronaviruses HCoV-229E and HCoV-NL63 (p = 0.15 and p = 0.25) (Figure 3A,C).
Table 1. Descriptive statistics of study participants at the beginning of the study (baseline) in the analytic dataset with complete data (k = 42) 5.

| Variable Name                              | Level | Total (n) | No      | Yes     | p-Value * |
|--------------------------------------------|-------|-----------|---------|---------|-----------|
| Biological Sex (n [%])                     |       |           |         |         |           |
| Male                                       | 13    | 21        | 5 (23.8)| 8 (38.1)| 0.504     |
| Female                                     | 29    | 21        | 16 (76.2)| 13 (61.9)| 0.504     |
| Age (mean [SD])                            | 42    | 41.48 (11.66)| 46.33 (11.91)| 0.189 |
| Days Since Positive qPCR Test (mean [SD])  | -     | 127.62 (88.45)| 165.33 (115.90)| 0.243 |
| Pre-Vaccine Sample (n [%])                 |       |           |         |         |           |
| True                                       | 42    | 21 (100)  | 21 (100)|         | –         |
| Duration of COVID-19 Symptoms (n [%])      |       |           |         |         |           |
| ≤2 Weeks                                   | 26    | 13 (61.9) | 13 (61.9)| 1.00    |
| >2 Weeks                                   | 16    | 8 (38.1)  | 8 (38.1)|         | –         |
| SARS-CoV-2 anti-Spike-IgG                  |       |           |         |         |           |
| Log10 AU/mL                                | -     | -         | 4.00 (0.82)| 3.88 (0.50)| 0.583     |
| SARS-CoV-2 anti-RBD-IgG                    |       |           | 3.63 (0.81)| 3.61 (0.49)| 0.891     |
| Log10 AU/mL                                | -     | -         | 4.02 (0.82)| 4.00 (0.72)| 0.755     |
| SARS-CoV-2 anti-Nucleocapsid-IgG           |       |           | 4.33 (0.42)| 4.32 (0.53)| 0.971     |
| Log10 AU/mL                                | -     | -         | 4.14 (0.47)| 4.19 (0.52)| 0.779     |
| 229E-CoV anti-Spike-IgG                    |       |           | 3.60 (0.46)| 3.62 (0.41)| 0.891     |
| Log10 AU/mL                                | -     | -         | 4.75 (0.53)| 4.68 (0.54)| 0.642     |
| HKU1-CoV anti-Spike-IgG                    |       |           | 4.14 (0.47)| 4.19 (0.52)| 0.779     |
| Log10 AU/mL                                | -     | -         | 3.60 (0.46)| 3.62 (0.41)| 0.891     |
| NL63-CoV-2 anti-Spike-IgG                  |       |           | 4.75 (0.53)| 4.68 (0.54)| 0.642     |
| Log10 AU/mL                                | -     | -         | 4.14 (0.47)| 4.19 (0.52)| 0.779     |
| Follow Up Time (median [SD])               |       |           | 85 (25.87)| 84 (9.20)| 0.435     |
| Number of Follow Up Visits Per-Participant (mean [SD]) |       |           | 5.95 (1.60) | 6.90 (0.3) | 0.014     |

5 Participants are stratified by vaccine status (primary exposure) throughout the study period, k = 21 participants were vaccinated while under observation. Bivariate associations at baseline were examined by testing for a difference in the distribution of covariates between participants who did or did not receive one dose of a COVID-19 vaccine over the study period. * p-values are reported for parametric tests used for continuous (t-test) and categorical variables (χ² test).

Figure 2. SARS-CoV-2 anti-IgG pre- and post-vaccination. Antibody signals in k = 21 paired participants, who re-ceived a COVID-19 vaccine during the study, at baseline and 14 to 56 days post-vaccination, presented by individual SARS-CoV-2 antigen (k = 21): (A) anti-S, (B) anti-RBD, (C) anti-N. Differences in antibody signals were examined with a paired Wilcoxon signed rank test.
3.4. Primary Analysis

Linear mixed-effects regression models were used to estimate intraclass correlation within participant samples and the relationship between COVID-19 vaccination and SARS-CoV-2 anti-S or anti-N IgG antibody concentration. An unconditional mean model was fit to partition within participant variation from between participant variation (Table 2). The minority of variation in SARS-CoV-2 anti-S IgG concentration was attributable to differences between participants (ICC = 0.43) (Table 2). On average, anti-S IgG concentration increased over time in participants who received one dose of a COVID-19 vaccine during the study by 2.06 log10 AU/mL (95% CI: 1.45–3.46) adjusting for age, biological sex, days from positive qPCR test (time) and effect modification between COVID-19 vaccination and time (Table 2). In the adjusted model, the ICC increased to 0.89 indicating that between participant differences (e.g., COVID-19 vaccination) explains most of the variation in SARS-CoV-2 anti-S IgG antibody concentration. COVID-19 vaccination has a positive association with SARS-CoV-2 anti-S IgG antibody concentration, which increases over time. Variation in anti-N IgG concentration was due to differences between participants in the unconditional mean model (ICC = 0.88). The average, anti-N IgG concentration decreased in vaccinated participants over time.

\(-0.243 \log_{10} \text{AU/mL}, 95\% \text{CI: } -1.2-[-0.12]\) adjusting for age, biological sex, days from positive qPCR test (time) and effect modification between COVID-19 vaccination and time (Table 2). Variation in anti-N IgG concentration after fitting the adjusted model was explained by within participant variance (ICC = 30). Overall, these results indicate that waning of SARS-CoV-2 anti-N IgG is unaffected by COVID-19 vaccination. Anti-S IgG

**Figure 3.** Endemic human coronavirus anti-S IgG antibody signals pre- and post-vaccination. HCoV antibody signals in n = 21 paired participants, who received a COVID-19 vaccine during the study, measured at baseline (before vaccination) and 14 to 56 days post-vaccination, presented by HCoV species: (A) HCoV-229E anti-Spike (S), (B) HCoV-HKU1 anti-S, (C) HCoV-NL63 anti-S, (D) HCoV-OC43 anti-S. Difference in antibody signal was examined with a paired Wilcoxon signed rank test.
titers increase post vaccination; therefore, vaccination of recovered individuals benefits the durability of their humoral immune response.

**Table 2.** Summary of linear mixed effects models fit to examine the relationship between anti-S IgG log10 AU/mL (light grey) or anti-N IgG log10 AU/mL (dark grey) and COVID-19 vaccination status adjusting for: biological sex, age, and time from qPCR diagnosis.

| Unconditional Mean Model (S) | Intraclass Correlation Coefficient |
|-----------------------------|-----------------------------------|
| Participant ID (n = 42)     | 0.434                             |
| Residual                    | 0.566                             |

Random Intercept Model

| Variable                  | Fixed Effect Estimate | 95% CI       |
|---------------------------|-----------------------|--------------|
| Intercept                 | 4.84                  | 3.27–6.39    |
| Vaccine-Yes               | 0.40                  | −0.41–1.20   |
| Biological Sex-Male       | 0.93                  | 0.068–1.79   |
| Age (Years)               | −0.029                | −0.063–0.0057|
| Time from +ve qPCR Test * | −0.20                 | −0.47–0.054  |
| Vaccine: Time *           | 1.86                  | 1.39–2.21    |

Random Effects

Intraclass Correlation Coefficient 0.893

**Unconditional Mean Model (N)**

| Participant ID (n = 42) | 0.875 |
|-------------------------|-------|
| Residual                | 0.125 |

Random Intercept Model

| Variable                  | Fixed Effect Estimate | 95% CI       |
|---------------------------|-----------------------|--------------|
| Intercept                 | 3.14                  | 2.48–3.79    |
| Vaccine-Yes               | −0.080                | −0.42–0.26   |
| Biological Sex-Male       | 0.27                  | −0.095–0.63  |
| Age (Years)               | 0.016                 | 0.0017–0.03  |
| Time from +ve qPCR Test * | −0.40                 | −0.53–(−0.27)|
| Vaccine: Time *           | −0.077                | −0.25–0.11   |

Random Effects

Intraclass Correlation Coefficient 0.30

* An effect modification term was incorporated to explore how the effect of vaccination on antibody concentration differs by time since diagnosis with a qPCR test. Unconditional means models were fit to partition the variance by participant without inclusion of other exposure variables. Fixed effect models were built by applying the common cause criterion to select covariates which are a cause of the exposure, outcome, or both.

3.5. Secondary Analysis

Secondary analysis used the Kaplan–Meier method to estimate the cumulative incidence of seroreactivity above a defined threshold in vaccinated and unvaccinated participants over time. Seroreactive status was classified by the threshold of ≥5.5 log10 AU/mL SARS-CoV-2 anti-S IgG, as described in Methods. Participants with antibody measurements equal to or greater than the threshold were considered reactive. Over the 105 days follow up from baseline (first antibody measurement available for participants post-infection), 88% (95%CI: 42–98%) of vaccinated participants (n = 16) were seroreactive compared to 5% (95%CI: 0–14%) of unvaccinated participants (n = 1) (p = 0.03) (Figure 4). A single dose of COVID-19 vaccine increases the probability of a SARS-CoV-2 anti-S IgG antibody concentration ≥5.5 log10 AU/mL by 83%.
A prospective cohort study was carried out in Greater Vancouver, British Columbia to observe anti-SARS-CoV-2 and anti-endemic HCoV antibody dynamics in participants who were infected with SARS-CoV-2, a subset received the first dose of a Health Canada approved SARS-CoV-2 vaccine during the follow up. Several commercial serology assays were used to detect anti-coronavirus antibodies; detection of anti-SARS-CoV-2 antibodies was confirmed in all available samples, although both anti-S and anti-N antibodies decline over time post-infection. Bivariate analysis found that vaccination significantly increased the titer of SARS-CoV-2 anti-S IgG antibodies 14–56 days post vaccination. A positive association was found between SARS-CoV-2 vaccination and endemic human β-coronavirus (HCoV-OC43 and HCoV-HKU1) anti-S IgG antibodies. It cannot be ruled out that infection with an endemic HCoV is an alternative/competing cause of the observed increase; however, the incidence of influenza-like (syndromic) respiratory infections was low during the study period [44]. Vaccination was not observed to boost SARS-CoV-2 anti-N IgG titers, which waned overtime in both vaccinated and unvaccinated individuals. The rate of anti-N waning was approximately double that of anti-S. Secondary analysis used a Kaplan–Meier model to estimate the cumulative incidence of anti-S antibody titers equal to or above 5.5 log10 AU/mL (‘seroreactivity threshold’) in those vaccinated and unvaccinated. In the vaccinated group, 88% (95% CI: 42–98%) of participants had SARS-CoV-2 anti-S IgG...
titers greater than or equal to this threshold, while this level was achieved in only one unvaccinated participant measured twenty-seven days post infection. Asymptomatic and subclinical SARS-CoV-2 infections have been observed to prime the adaptive immune response and may explain observed increases in unvaccinated participants. In people with asymptomatic re-infections, anti-S IgG titers wane more slowly over time and anti-N may increase [45].

In unvaccinated participants, a few more substantial increases in anti-S titers were observed despite overall antibody waning. No reinfections were confirmed using qPCR in self-collected saline gargle samples throughout the study; however, one participant had a large (>8-fold) average increase in mean (anti-S, RBD and N) antibody levels seven months following initial SARS-CoV-2 diagnosis, which may be explained by a second exposure to SARS-CoV-2. A second participant had 6-fold increase of anti-S and anti-RBD levels, but not anti-N IgG levels, suggesting they may also have been re-exposed. Other detected increases in antibody levels were of much smaller magnitude and might be secondary to rising titers early in convalescence or be explained by technical variations rather than a biological mechanism.

4.2. Comparison with Literature

Previous studies have measured changes in SARS-CoV-2 antibody titers over time. Repeated exposure to SARS-CoV-2 antigens increases IgG titer, while antibodies generated from a single exposure wane overtime [46,47]. Following infection, SARS-CoV-2 specific antibody waning has been observed to decrease from the 8th to 9th week post symptoms onset, with detectable levels observed up to the end of the 12th week [48]. In those with multiple SARS-CoV-2 exposures or a hybrid immune response from infection and vaccination, decrease of antibody titers stops shortly after the secondary antigen exposure when stimulation of the memory B cell response produces additional antibodies [49]. A strong correlation between total lymphocyte count and SARS-CoV-2 anti-S IgG provides evidence that an ongoing/active immune response provides better protection than a dormant one [48]. Waning of SARS-CoV-2 antibodies differs by the SARS-CoV-2 epitope they target, anti-N IgG antibodies wane faster than anti-S. The difference in reactivity between anti-N to anti-S IgG was observed at the population level, anti-N seroprevalence underestimated the number of confirmed infections by 9–31% [50]. Vaccination post SARS-CoV-2 infection prevents waning of anti-S but has no effect on anti-N IgG [51]. Hybrid immunity also benefits the breadth of the antibody mediated response, increasing the probability that existing antibodies are effective against the novel variants. Persons who were infected prior to receiving one of two doses of a COVID-19 vaccine had more somatic mutations and antibody production from the IGHV2-5; IGHJ4-1 germline which was not active in the vaccinated but uninfected [52,53]. Additionally, hybrid immunity produces greater total and neutralizing anti-S titers than natural infection or vaccination alone [49]. Our study both supports and builds upon prior findings, as we show that SARS-CoV-2 IgG antibodies wane in SARS-CoV-2 infected people over time, with the rate of decline being greater for anti-N IgG than anti-S; vaccination post-infection boosts anti-S IgG titers and participants with hybrid immunity possess anti-S antibody levels which are not common in those infected but unvaccinated. We propose 5.5 log10 AU/mL anti-S IgG as a putative correlate of protection in persons convalescent for SARS-CoV-2 infection who have received a single dose of a Health Canada approved vaccine. Our calculated rates of antibody decline may be used to help estimate infection timing in seroprevalence studies.

4.3. Clinical and Epidemiological Interpretation

Our findings have important implications for clinical practice and public health guidelines as the pandemic progresses into its third year, novel viral variants continue to emerge, and vaccine doses are more widely distributed globally. Humoral immunity from natural infection wanes and vaccination with at least one dose of COVID-19 vaccine increases SARS-CoV-2 anti-S IgG titers immediately and over time. Therefore, we recommend that
naturally infected individuals receive COVID-19 vaccination to increase protection from re-infection and severe disease and the duration of their humoral immune response against SARS-CoV-2. We demonstrate that a single dose of SARS-CoV-2 vaccine is effective in boosting anti-S antibody titers to high levels, which has implications in distribution of vaccine supplies in those countries with scarce access and low vaccination levels in the setting of high numbers of natural infection.

4.4. Strengths and Limitations

The strength of the described study stems from the prospective design, use of multiple serological tests, including the quantitative MSD option, and thorough analysis. A prospective cohort design offers several benefits, which allowed us to observe SARS-CoV-2 antibody dynamics over time with minimal bias. Recruiting participants post-infection but prior to vaccination delineated the sequence of temporal events, limiting the probability that any changes in antibody titers observed post-vaccination were due to causes other than the vaccine. Selection bias was minimized as the participants exposure (vaccination status) and outcome (IgG titer) were not known when they were recruited into the study. At the beginning of the study, the measured covariates were exchangeable between participants who were unvaccinated or vaccinated during follow-up. Balance of the covariates allowed for estimation of the relationship between vaccination and anti-SARS-CoV-2 IgG antibody titer with minimal bias from confounding. Utilizing multiple serological tests increased the rigor of our observations, limited instrument bias and allows for our findings to be generalized between different types of serological testing methods. Statistical power was optimized by analysis with a mixed effects linear regression model, which accommodated multiple repeated measures per participant.

Limitations of the work include differential loss to follow up in the vaccinated and unvaccinated groups, a small sample size, and incomplete/missing survey responses and incomplete whole genome sequencing data. Unvaccinated participants were observed to have approximately one fewer visit than those who received a COVID-19 vaccine. Vaccines were not an originally planned intervention in the study and were made available in British Columbia on a stage roll-out basis about half-way through the study period. The difference in visit numbers between the vaccinated and unvaccinated groups may be related to surveillance bias- those who receive a medical intervention are more open to clinical follow up than those who do not. Obtaining a larger sample size initially planned for the study was difficult due to low enrollment uptake, likely related to the social and economic stress of the pandemic on the public and geographic limitations on recruitment related to the availability of sample collection sites.

Despite the small sample size, we detected a significant increase in human β-coronavirus (HCoV-HKU1 and HCoV-OC43) anti-S titers following SARS-CoV-2 vaccination, while no difference was observed for the α-coronaviruses (HCoV-NL63 and HCoV-229E). Antigenic cross-reactivity between human β-coronaviruses may allow for ‘back-boosting’ a phenomenon which has been well described for Influenza A viruses [54]. Antibodies are ‘back-boosted’ when a secondary exposure to a novel viral strain generates new antibodies and increases the titer of antibodies against a previously encountered strain. As a result, SARS-CoV-2 vaccinations may provide non-specific protection in children who have a high incidence of endemic coronavirus infections. The study design may have underestimated any association between existing endemic coronavirus IgG titers and COVID-19 vaccination as the sample was restricted to persons previously infected with SARS-CoV-2. COVID-19 has been shown to affect endemic coronavirus antibody levels and as such, the effect of vaccination should be observed in a cohort of SARS-CoV-2 naive persons prior to vaccination [55–57]. The overall effect of SARS-CoV-2 vaccination and/or infection on the circulating antibodies against endemic HCoVs in the population may have implications for their seasonal epidemiology.
5. Conclusions

In summary, we report that single dose vaccination in a British Columbia-based cohort after natural infection significantly increases SARS-CoV-2 anti-S IgG titer by 1.63 log_{10} units and that vaccination increases the durability of high anti-S titers over time. Vaccination post-natural infection had no significant association with SARS-CoV-2 anti-N IgG titer; a significant trend towards higher anti-S IgG against the endemic human β-coronaviruses (HCoV-HKU1 and HCoV-OC43) was observed. Our results provide support that vaccination is beneficial for achieving higher and more persistent SARS-CoV-2 anti-S IgG titers. We also report an estimated rate of decay of anti-N antibodies, which may be useful for measuring ongoing population seroprevalence estimates. Future studies should examine the impact of infection following vaccination on antibody dynamics, as vaccine breakthrough infections with omicron or other variants of concern continue to occur.

Supplementary Materials: The following supporting information can be downloaded at: https://www.mdpi.com/article/10.3390/v14112416/s1, Table S1: Summary of serological test results; Table S2: Descriptive summary of variables not included in the analysis for n = 21 vaccinated participants; Table S3: Descriptive summary of variables not included in the analysis for n = 21 unvaccinated participants; Figure S1: Exclusion criteria were applied to select an analytic data of n = 270 observations from k = 42 dependent participants (clusters); Figure S2: Phylogenetic analysis of CARE Study Participants’ SARS-CoV-2 genomes.

Author Contributions: Conceptualization, A.D.O., A.N.J., M.M. and I.S.; methodology, A.M.N., A.C.M., E.F., M.S., A.H. and S.M.; software, A.D.O. and A.M.N.; formal analysis, A.D.O. and A.M.N.; investigation, A.D.O. and I.S.; resources, A.D.O., A.N.J., M.M., I.S., A.C.M., E.F., M.S., A.H., D.M.G. and S.M.; data curation, S.S., A.C.M. and T.V.; writing—original draft preparation, A.D.O. and A.M.N.; writing—review and editing, All Authors; visualization, A.D.O. and A.M.N.; supervision, A.N.J. and I.S.; project administration, A.C.M., E.F. and A.H.; genomic data provision and analysis S.L.R. and N.P.; funding acquisition, A.N.J., M.M. and I.S. All authors have read and agreed to the published version of the manuscript.

Funding: This study was funded by Genome BC’s COVID-19 Rapid Response Funding Program (#COV-050).

Institutional Review Board Statement: The study was conducted in accordance with the Declaration of Helsinki, and approved by the Institutional Review Board (or Ethics Committee) of The University of British Columbia (H20-01089 on 4/21/2020).

Informed Consent Statement: Informed consent was obtained from all subjects involved in the study.

Data Availability Statement: The datasets generated during and/or analysed during the current study are available from the corresponding author on reasonable request.

Acknowledgments: We thank the dedicated staff at the BCCDC PHL for processing, testing, and sequencing clinical specimens. We also thank all the study participants who volunteered their time and specimens for this research. We also are thankful to all the staff who helped to collect and process samples for this study at Abbotsford Regional Hospital, Surrey Memorial Hospital and St. Paul’s Hospital. Notably, we thank the staff at BC Children’s Hospital for providing extra guidance and support to the researchers in conducting the study.

Conflicts of Interest: The authors declare no conflict of interest. The funders had no role in the design of the study; in the collection, analyses, or interpretation of data; in the writing of the manuscript; or in the decision to publish the results.
50. Fenwick, C.; Croxatto, A.; Coste, A.T.; Pojer, F.; André, C.; Pellaton, C.; Farina, A.; Campos, J.; Hacker, D.; Lau, K.; et al. Changes in SARS-CoV-2 Spike versus Nucleoprotein Antibody Responses Impact the Estimates of Infections in Population-Based Seroprevalence Studies. *J. Virol.* 2021, 95, e01828-20. [CrossRef] [PubMed]

51. Jalkanen, P.; Kolehmainen, P.; Hänninen, H.K.; Huttunen, M.; Tahtinen, P.A.; Lundberg, R.; Maljanen, S.; Reinholm, A.; Tauriainen, S.; Pakkanen, S.H.; et al. COVID-19 MRNA Vaccine Induced Antibody Responses against Three SARS-CoV-2 Variants. *Nat. Commun.* 2021, 12, 3991. [CrossRef] [PubMed]

52. Andreano, E.; Paciello, I.; Piccini, G.; Manganaro, N.; Pileri, P.; Hyseni, I.; Leonardi, M.; Pantano, E.; Abbiento, V.; Benincasa, L.; et al. Hybrid Immunity Improves B Cells and Antibodies against SARS-CoV-2 Variants. *Nature* 2021, 600, 530–535. [CrossRef]

53. Wang, Z.; Muecksch, F.; Schilder-Babajew, D.; Finkin, S.; Viant, C.; Gaebler, C.; Hoffmann, H.-H.; Barnes, C.O.; Cipolla, M.; Ramos, V.; et al. Naturally Enhanced Neutralizing Breadth against SARS-CoV-2 One Year after Infection. *Nature* 2021, 595, 426–431. [CrossRef]

54. Skowronski, D.M.; Sabaiduc, S.; Leir, S.; Rose, C.; Zou, M.; Murti, M.; Dickinson, J.A.; Olsha, R.; Gubbay, J.B.; Croxen, M.A.; et al. Paradoxical Clade- and Age-Specific Vaccine Effectiveness during the 2018/19 Influenza A(H3N2) Epidemic in Canada: Potential Imprint-Regulated Effect of Vaccine (I-REV). *Eurosurveillance* 2019, 24, 1900585. [CrossRef]

55. Guo, L.; Wang, Y.; Kang, L.; Hu, Y.; Wang, L.; Zhong, J.; Chen, H.; Ren, L.; Gu, X.; Wang, G.; et al. Cross-Reactive Antibody against Human Coronavirus OC43 Spike Protein Correlates with Disease Severity in COVID-19 Patients: A Retrospective Study. *Emerg. Microbes Infect.* 2021, 10, 664. [CrossRef]

56. Lin, C.Y.; Wolf, J.; Brice, D.C.; Sun, Y.; Locke, M.; Cherry, S.; Castellaw, A.H.; Wehenkel, M.; Crawford, J.C.; Zarnitsyna, V.I.; et al. Pre-Existing Humoral Immunity to Human Common Cold Coronaviruses Negatively Impacts the Protective SARS-CoV-2 Antibody Response. *Cell Host Microbe* 2022, 30, 83–96.e4. [CrossRef] [PubMed]

57. Anderson, E.M.; Goodwin, E.C.; Verma, A.; Arevalo, C.P.; Bolton, M.J.; Weirick, M.E.; Gouma, S.; McAllister, C.M.; Christensen, S.R.; Weaver, J.E.; et al. Seasonal Human Coronavirus Antibodies Are Boosted upon SARS-CoV-2 Infection but Not Associated with Protection. *Cell* 2021, 184, 1858. [CrossRef] [PubMed]