Disentangling post-vaccination symptoms from early COVID-19

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ABSTRACT

Background: Identifying and testing individuals likely to have SARS-CoV-2 is critical for infection control, including post-vaccination. Vaccination is a major public health strategy to reduce SARS-CoV-2 infection globally. Some individuals experience systemic symptoms post-vaccination, which overlap with COVID-19 symptoms. This study compared early post-vaccination symptoms in individuals who subsequently tested positive or negative for SARS-CoV-2, using data from the COVID Symptom Study (CSS) app.

Methods: We conducted a prospective observational study in 1,072,313 UK CSS participants who were asymptomatic when vaccinated with Pfizer-BioNTech mRNA vaccine (BNT162b2) or Oxford-AstraZeneca adeno-virus-vectorized vaccine (ChAdOx1 nCoV-19) between 8 December 2020 and 17 May 2021, who subsequently reported symptoms within seven days (N=362,770) (other than local symptoms at injection site) and were tested for SARS-CoV-2 (N=14,842) aiming to differentiate vaccination side-effects per se from superimposed SARS-CoV-2 infection. The post-vaccination symptoms and SARS-CoV-2 test results were contemporaneously logged by participants. Demographic and clinical information (including comorbidities) were recorded. Symptom profiles in individuals testing positive were compared with a 1:1 matched population testing negative, including using machine learning and multiple models considering UK testing criteria.

Findings: Differentiating post-vaccination side-effects alone from early COVID-19 was challenging, with a sensitivity in identification of individuals testing positive of 0.6 at best. Most of these individuals did not have fever, persistent cough, or anosmia/dysosmia, requisite symptoms for accessing UK testing; and many only had systemic symptoms commonly seen post-vaccination in individuals negative for SARS-CoV-2 (headache, myalgia, and fatigue).

Interpretation: Post-vaccination symptoms per se cannot be differentiated from COVID-19 with clinical robustness, either using symptom profiles or machine-derived models. Individuals presenting with systemic symptoms post-vaccination should be tested for SARS-CoV-2 or quarantining, to prevent community spread.

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Abbreviations: AUC, Area under the curve; bMEM, Bayesian mixed-effect model; BMI, Body mass index; CI, Confidence interval; COVID-19, Coronavirus disease 2019; CSS, COVID Symptoms Study; DI, Data invalid; IQR, inter quartile range; KCL, King’s College London; LFAT, Lateral flow antigen test; LR, Logistic Regression; NHS UK, National Health Service of the United Kingdom; O-AZ, Oxford-AstraZeneca adeno-virus-vectorized vaccine; PB, Pfizer-BioNTech mRNA vaccine; RF, Random forest; ROC, Receiver operating curve; rtPCR, Reverse transcription polymerase chain reaction; SARS-CoV-2, Severe acute respiratory syndrome-related coronavirus-2; UK, United Kingdom of Great Britain and Nothern Ireland

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Research in context

Evidence before this study

There are multiple surveillance platforms internationally interrogating COVID-19 and/or post-vaccination side-effects. We designed a study to examine for differences between vaccination side-effects and early symptoms of COVID-19. We searched PubMed for peer-reviewed articles published between 1 January 2020 and 21 June 2021, using keywords: “COVID-19” AND “Vaccination” AND (“mobile application” OR “web tool” OR “digital survey” OR “early detection” OR “Self-reported symptoms” OR “side-effects”). Of 185 results, 25 studies attempted to differentiate symptoms of COVID-19 vs. post-vaccination side-effects; however, none used artificial intelligence (AI) technologies (“machine learning”) coupled with real-time data collection that also included comprehensive and systematic symptom assessment. Additionally, none of these studies attempted to discriminate the early signs of infection from side-effects of vaccination (specifically here: Pfizer-BioNTech mRNA vaccine (BNT162b2) and Oxford-AstraZeneca adenovirus-vectorised vaccine (ChAdOx1 nCoV-19)). Further, none of these studies sought to provide comparisons with current testing criteria used by healthcare services.

Added value of this study

This study, in a large community-based cohort, uses prospective data capture in a novel effort to identify individuals with COVID-19 in the immediate post-vaccination period. Our results suggest that early symptoms of SARS-CoV-2 cannot be differentiated from vaccination side-effects robustly.

Implications of all the available evidence

Our study suggests that post-vaccination symptoms per se cannot be differentiated from COVID-19 with clinical robustness and therefore individuals presenting with systemic symptoms post-vaccination should be tested for SARS-CoV-2 to prevent community spread.

1. Introduction

The havoc wrought by SARS-CoV-2 is unprecedented in living memory, with >184 million cases of COVID-19 world-wide and ~4.0 million deaths by 8 July 2021 [1,2]. Extraordinary efforts directed towards rapid vaccine development meant that by late 2020 the UK Medicines and Healthcare products Regulatory Agency had authorized three vaccines: Pfizer-BioNTech mRNA vaccine (BNT162b2) [3,4] and Oxford-AstraZeneca adenovirus-vectorised [5–7] and Moderna mRNA (mRNA-1273) [8,9]. A fourth vaccine (Janssen adenovirus-vectorised Ad26.COV2.S) was authorised on 28 May 2021 [10]. Vaccination with BNT162b2 (herein, PB) and ChAdOx1 nCoV-19 (herein, O-AZ) started in the UK on 8 December 2020 [11] and 4 January 2021 [12] respectively, during which time the UK was experiencing its third pandemic wave with widespread community transmission (peak UK positive specimens reported on 29 December 2020) [13]. Since then, and in the context of social distancing and stay-at-home directives, new infections, hospitalisations, and deaths from SARS-CoV-2 have fallen rapidly and remained low until June 2021 [1,2,13].

Local and systemic reactions have been observed after all vaccines for SARS-CoV-2. Considering the two vaccines used predominantly in the UK to date (O-AZ and PB), local reactions were common during their pivotal trials (76% of younger (<55 years) O-AZ recipients reported tenderness; [5,6] 83% of younger PB recipients reported pain) [4]. Systemic reactions were also common and included fatigue (O-AZ 76%; PB 59%), headache (O-AZ 65%; PB 52%), and fever (O-AZ 24%; PB 16%) [4–6]. Observational data from the COVID Symptom Study (CSS) [14,15] also showed high incidence of local (62%) and systemic (26%) effects [16]. Other real-world experience has resulted in identification of some very rare but serious side effects, such as vaccine-induced thrombotic thrombocytopenia (VITT), associated with anti-PF4 antibody production, and myocarditis [17–19]. Saliently, most vaccine-related side-effects (including VITT) are more common in younger individuals, whereas COVID-19 clinical severity increases with age [4–6,16].

Prevention of SARS-CoV-2 dissemination requires rapid recognition followed by quarantining of infected individuals (along with appropriate health care). However, there is overlap between symptoms from COVID-19 [1,5,20,21] and early post-vaccination systemic symptoms [4–6,16]. Moreover, immunity to SARS-CoV-2 does not occur immediately post-vaccination, [22] with functional protection from approximately day 12 [23]. Quaratining and testing every individual with systemic symptoms early post-vaccination would be onerous, expensive, and labour-intensive – but given the impact of viral outbreaks might be unavoidable if SARS-CoV-2 infection cannot be excluded robustly [15,20].

Here we aim to determine whether symptom profiles can be used to differentiate individuals with systemic side-effects of vaccination alone from individuals with superimposed SARS-CoV-2 infection.

2. Methods

2.1. Study design and Participants

Data were acquired prospectively from the CSS, using a mobile health application launched by ZOE Limited and King’s College London in March 2020 (app details and development given in Supplementary Methods) [14,15]. Briefly, individuals are asked daily to log their health status, health care access, SARS-CoV-2 testing and results, and vaccination details, with direct questions about symptoms associated with COVID-19 (Supplementary Table S1) [14,15]. Symptomatic individuals are prompted to undergo testing, either through standard care or through test request from ZOE/CSS [24].

Data were acquired from UK participants aged 16-90 years, between 8 December 2020 (UK vaccination start date) and 17 May 2021, who were asymptomatic when vaccinated with PB or O-AZ (first or second dose), and subsequently reported: a) at least one predefined symptom (Supplementary Table S1) within seven days post-vaccination, and b) a SARS-CoV-2 test result (reverse transcription polymerase chain reaction [rtPCR] or lateral flow antigen test [LFAT]) within ten days post-vaccination. The seven-day cut-off for symptom presentation was informed by: a) serial interval for COVID-19 (the time interval between the primary and secondary case; mean, 5.2 days [25]); b) incubation period for SARS-CoV-2 (mean, 5.8 days [25,26]); c) the timeline for acute post-vaccination side-effects in both pivotal trials (one-week [4–6,8,10]) and d) reported real-world experience of post-vaccination symptoms (peak prevalence day 1 post-vaccination; mean duration one day [16]). The ten-day cut-off for testing allowed three days’ delay in accessing testing [27].

Early results indicated a large imbalance in numbers of individuals testing positive vs. negative post-vaccination (3525:11317, positive: negative SARS-CoV-2 tested individuals), sufficient to bias analysis. [28] A 1:1 population from the negative cohort (matching age, BMI, gender, occupation, week of testing, and comorbidities) was selected based on minimisation of Euclidean distance between positive and negative subjects considering these features, enabling a fair comparison between groups of equal size [29]. However, to ensure robustness, analyses were repeated using: a) a one-hundred bootstrapping
scheme selecting from the negative population; [30] and b) the entire negative population.

Individual symptoms (here, a symptom reported at any time within seven days post-vaccination, irrespective of duration) were compared between recently vaccinated individuals testing positive or negative for SARS-CoV-2, using Chi-squared tests per symptom, given that normality was not present for most of the symptom’s distribution. Duration of individual symptoms was calculated as days from first report of that symptom, until asymptomatic and/or seven days post-vaccination, noting that duration beyond seven days was not considered; however, as the number of individuals experiencing each symptom was low in both groups, no statistical comparison was made. Symptom burden, defined as total symptom count per person [irrespective of symptom duration] was compared between groups using Mann-Whitney-U tests. We also considered symptom manifestation across the week post-vaccination, by dynamic profiling for each symptom (symptom frequency). Correlation of individual symptoms within both positive and negative individuals was assessed by computing a Spearman-rank correlation test. Local symptoms due to vaccination per se (Supplementary Table S2) were excluded from analysis as unlikely to be indicative of, or influenced by, SARS-CoV-2 infection.

Machine learning was used to determine if post-vaccination symptoms per se could be separated from superimposed SARS-CoV-2 infection (including symptom combination, and cumulative symptom burden) [15,21,31]. We trained a set of binary classifiers to identify SARS-CoV-2-positive individuals. Models included random forest, logistic regression, and Bayesian mixed-effect models, exploiting their varying properties (Supplementary Table S3) to improve reliability of results. The models were trained using the outcome of the SARS-CoV-2 testing as response variable and ground truth for both training and model assessment (validation). We also considered whether symptoms occurred after the first or second vaccination, by including vaccination dose as a covariate. Models were trained on data without stratifying by vaccine type, due to small sample sizes. Further covariates included the age, gender, and BMI of the participants (details available in Supplementary Table S3). We also did not discriminate between type of SARS-CoV-2 testing (PCR vs. LFAT), or mode of testing access (NHS vs. ZOE-request), either in model development or other analyses.

We sought to reduce bias from assessing high numbers of individual symptoms by performing symptom-clustering using K-means [32]. However, a relevant/accurate number of clusters was not evident from the silhouette plot and entropy (data not shown); thus, further analyses using machine clustering were not pursued. Symptoms were clustered manually into clinical groupings (reviewed by ELD, MO, TS, AH, CJS) (Supplementary Table S4), and analysed using the above models similarly. Lastly, illness classification according to Figure 1. Flowchart of individuals included in this study. Symptoms* within 7 days excluded local symptoms related to injection site. SARS-CoV-2 test included both rtPCR and LFAT. Positive and negative refers to self-logged test results. DI: Data invalid. 1st and 2nd dose refer to the first and second doses of the two vaccines, respectively.
based on having at least one of the four symptoms required for accessing NHS testing during the timing of this study (viz., presence or absence of fever, persistent cough, anosmia and/or dysosmia) [24] were assessed.

Data were split into training and validation sets for random forest, logistic regression, and Bayesian mixed-effect models. Five folds were used on the training set, composed of 80% of the initial dataset randomly selected, to train the models in different subsamples of the population. The remaining 20% were then used to assess the performance of models, evaluating sensitivity, specificity, and balanced accuracy. The class ratio was maintained in both training and testing sets. Although models based on either clinical clustering or categorisation according to NHS criteria do not require training - and thus potentially both could be assessed using the full dataset - to ensure fair evaluation, both models (i.e., clinical clustering and categorisation according to NHS criteria) were assessed on 20% of the data of each cohort (i.e., the 20% of each fold corresponding to the testing set).

3. Ethical approval

The app and CSS were approved in the UK by KCL’s ethics committee (REMAS no. 18210, review reference LRS-19/20–18210). All app users provided informed consent for use of their data for COVID-19 research.

4. Role of the funding source

The funders, namely UK Government Department of Health and Social Care, Wellcome Trust, UK Engineering and Physical Sciences Research Council, UK National Institute for Health Research, UK Medical Research Council and British Heart Foundation and Chronic Disease Research Foundation and Zoe Limited had no role in study design, data analysis, data interpretation, or in content. ZOE Limited developed the app for data collection as a not-for-profit endeavour.

5. Results

Figure 1 shows the flow chart for this study. Overall, 1,072,313 UK CSS app users were vaccinated with either O-AZ or PB (O-AZ: 713,651; PB: 358,662). Of these, 362,770 (33.8% (O-AZ: 264,587; PB: 98,183)) reported at least one symptom early post-vaccination, with SARS-CoV-2 testing in 14,842 (4.09%) of these individuals (O-AZ: 10,765; PB: 4,077). A positive test was reported by 150 (1.01%) individuals (O-AZ: 73/10,765 (0.68%); PB: 77/4077 (1.89%)).

Within the tested group, 3,525/14,842 (23.75%) reported at least one requisite symptom fulfilling UK testing criteria; [24] 62 (1.76%) tested positive. Conversely, 11,317 tested individuals did not report any requisite symptom, of whom 88 (0.78%) tested positive. Individuals with requisite symptoms were more likely to test positive than those without (p-value<0.0001); none-the-less, the majority (88 of 150, 59%) who tested positive did not meet current UK testing criteria (Table 1).

For further analyses, one positive individual (vaccinated with PB) was excluded due to invalid data entry (invalid BMI), leaving 149 symptomatic positively-tested individuals. Table 2 describes the positive and matched negative cohorts.

Four symptomatic individuals who tested positive did so within 10 days of their second vaccination; their symptoms after first vaccination were thus disregarded. As the matched negative control cohort (N=149) included matching for vaccination order, the controls also included four individuals reporting symptoms after second vaccination. Given that small sample sizes compromise machine learning and model training, all 149 subjects in each of the positive and negative groups were included for either training or testing of the models (according to the percentages mentioned above). We regressed the impact of vaccination order, by including it as a co-variate of the models, or as a pre-processing step for the clinical clustering and NHS algorithm. However, we could not conduct a fair statistical analysis of symptoms after first vs. second vaccination for symptom profiling, given the very small numbers of infected individuals presenting after second vaccination (i.e., 4 subjects). Thus, as post-vaccination symptoms may vary after first vs. second dose, [16,31]
we present the symptom profiles after first vaccination (N=145) in the main text, with data from the entire cohort (N=149) in the Supplementary Materials (Supplementary Table S5).

Individual symptom prevalence after first vaccination is shown in Figure 2 and Supplementary Table S5. Although some symptoms were more common in individuals testing positive vs. negative (e.g., sore throat (p-value = 0.0187), sneezing (p value = 0.0474) and persistent cough (p-value = 0.0396)), others were more common in the negative group (palpitations (p-value = 0.0284). The numbers of individuals reporting each symptom were small (e.g., sore throat, n=17; persistent cough, n=12) (Supplementary Table S5).

Median day of onset for any symptom post-vaccination was Day 1 in both groups (noting all individuals were asymptomatic when vaccinated), with highest symptom burden on Day 3, again in both groups (Figure 3; and Supplementary Tables S6 and S7). There was no difference in symptom burden between individuals testing positive or negative (median: 12 in positive group, 10.5 in negative group, Mann-Whitney test p-value = 0.22).

Table 3 presents prevalence of each symptom over the week post-vaccination, divided into three windows. Some symptoms increased over time in both positive and negative individuals (e.g., headache, myalgia) whereas others increased in positive individuals only (e.g., sneezing, hoarse voice). Although fever and sore throat increased across the week in the negative individuals, there was a suggestion of a biphasic response in the positive individuals, also observed with persistent cough. The numbers of individuals were too small for formal testing; moreover, the exact date of infection in positive individuals was unknown.

Individual symptom duration is shown in Supplementary Table S8. There were no significant differences in symptom duration in
Figure 3. Symptom prevalence and distribution during the first week after the first dose of vaccination, in symptomatic individuals testing positive or negative for SARS-CoV-2. The colour bar represents the percentage of symptomatic individuals reporting each symptom.

Table 3
Symptom prevalence and timing during the first week post-vaccination (N=145), in symptomatic individuals testing positive or negative for SARS-CoV-2, grouped by days post-vaccination. Number of subjects and percentage (value in parenthesis).

| Symptom                  | Positive SARS-CoV-2 | Negative SARS-CoV-2 |
|--------------------------|---------------------|---------------------|
|                          | 0-2 days            | 2-4 days            | 4-7 days   | 0-2 days | 2-4 days | 4-7 days   |
| Headache                 | 22 (15.2)           | 44 (30.3)           | 66 (45.5)  | 31 (21.4) | 46 (31.7) | 66 (45.5)  |
| Chills or Shivers        | 13 (9.0)            | 25 (17.2)           | 34 (23.4)  | 21 (14.5) | 25 (17.2) | 26 (17.9)  |
| Myalgias                 | 4 (2.8)             | 12 (8.3)            | 26 (17.9)  | 7 (4.8)   | 7 (4.8)   | 20 (13.8)  |
| Fever                    | 14 (9.7)            | 10 (6.9)            | 21 (14.5)  | 11 (7.6)  | 14 (9.7)  | 20 (13.8)  |
| Dizziness                | 4 (2.8)             | 13 (9.0)            | 16 (11.0)  | 6 (4.1)   | 21 (14.5) | 20 (13.8)  |
| Sneezing                 | 2 (1.4)             | 8 (5.5)             | 18 (12.4)  | 1 (0.7)   | 8 (5.5)   | 6 (4.1)    |
| Nausea                   | 5 (3.4)             | 8 (5.5)             | 11 (7.6)   | 6 (4.1)   | 15 (10.3) | 19 (13.1)  |
| Sensitive Skin           | 0 (0.0)             | 6 (4.1)             | 3 (2.1)    | 0 (0.0)   | 2 (1.4)   | 2 (1.4)    |
| Rhinorrhoea              | 4 (2.8)             | 10 (6.9)            | 21 (14.5)  | 1 (0.7)   | 7 (4.8)   | 19 (13.1)  |
| Eye Soreness             | 1 (0.7)             | 6 (4.1)             | 11 (7.6)   | 4 (2.8)   | 5 (3.4)   | 8 (5.5)    |
| Loss of Appetite         | 6 (4.1)             | 3 (2.1)             | 19 (13.1)  | 5 (3.4)   | 6 (4.1)   | 7 (4.8)    |
| Chest Pain               | 0 (0.0)             | 3 (2.1)             | 14 (9.7)   | 1 (0.7)   | 2 (1.4)   | 5 (3.4)    |
| Sore Throat              | 9 (6.2)             | 5 (3.4)             | 30 (20.7)  | 2 (1.4)   | 6 (4.1)   | 14 (9.7)   |
| Brain Fog                | 0 (0.0)             | 4 (2.8)             | 16 (11.0)  | 5 (3.4)   | 5 (3.4)   | 10 (6.9)   |
| Hoarse Voice             | 1 (0.7)             | 2 (1.4)             | 15 (10.3)  | 1 (0.7)   | 2 (1.4)   | 3 (2.1)    |
| Anosmia                  | 0 (0.0)             | 4 (2.8)             | 2 (1.4)    | 0 (0.0)   | 2 (1.4)   | 0 (0.0)    |
| Abdominal Pain           | 3 (2.1)             | 3 (2.1)             | 16 (11.0)  | 2 (1.4)   | 7 (4.8)   | 15 (10.3)  |
| Diarrhoea                | 3 (2.1)             | 4 (2.8)             | 11 (7.6)   | 2 (1.4)   | 3 (2.1)   | 9 (6.2)    |
| Dysosmia                 | 1 (0.7)             | 3 (2.1)             | 12 (8.3)   | 3 (2.1)   | 1 (0.7)   | 8 (5.5)    |
| Ear Pain                 | 0 (0.0)             | 3 (2.1)             | 3 (2.1)    | 1 (0.7)   | 3 (2.1)   | 4 (2.8)    |
| Red Welts FL             | 0 (0.0)             | 1 (0.7)             | 0 (0.0)    | 0 (0.0)   | 0 (0.0)   | 1 (0.7)    |
| Fatigue                  | 1 (0.7)             | 3 (2.1)             | 10 (6.9)   | 6 (4.1)   | 2 (1.4)   | 2 (1.4)    |
| Persistent Cough         | 4 (2.8)             | 1 (0.7)             | 17 (11.7)  | 0 (0.0)   | 2 (1.4)   | 3 (2.1)    |
| Swollen Glands           | 0 (0.0)             | 1 (0.7)             | 6 (4.1)    | 3 (2.1)   | 3 (2.1)   | 1 (0.7)    |
| Blisters on Feet         | 0 (0.0)             | 0 (0.0)             | 4 (2.8)    | 0 (0.0)   | 0 (0.0)   | 0 (0.0)    |
| Dyspnoea                 | 0 (0.0)             | 0 (0.0)             | 0 (0.0)    | 0 (0.0)   | 0 (0.0)   | 0 (0.0)    |
positive vs. negative individuals after first vaccination. Importantly, symptom assessment was truncated at seven days, noting as above that some symptoms were increasing in prevalence with time. Amongst individuals testing negative, dysosmia and delirium had the longest duration (median, 2 days for each).

Similar results for symptom prevalence after first vaccination were obtained comparing the positive population with the constructed cohort (1:1 matched) of negative individuals selected by bootstrapping (Supplementary Table S9) (Supplementary Figure S1); and with the negative population as a whole (Supplementary Figure S2). Some symptoms were significantly more common in negative individuals when using the entire negative population (e.g., brain fog), driven by the extremely large negative sample size, which supports our use of a selected matched population to avoid bias from unbalanced sample size.

There was no significant correlation between symptoms in either the positive or negative populations, assessed using Spearman-rank test (Supplementary Figure S3). As a sensitivity analysis we assessed the impact of a (self-logged) previous COVID-19 diagnosis; this did not alter our conclusions.

### 6. SARS-CoV-2 test outcome prediction modelling

Model performance including receiver operator curves, using all reported symptoms, are shown in Table 4 and Figure 4. The best performance was obtained with random forest, followed by logistic regression; however, neither reached clinical utility (conventionally, 80%). Other models, including clinical symptom clustering (Supplementary Table S4) and categorisation of individuals using NHS screening criteria, were no better than chance.

### 7. Discussion

Here we aimed to develop a clinically useful algorithm predictive of SARS-CoV-2 infection early post-vaccination, by parsing symptoms according to proven infection status in symptomatic individuals. Such an algorithm would be extremely useful, particularly in countries with limited health resources, as testing could be targeted towards those predicted positive, with quarantining of these individuals until an available result. To our knowledge, this is the first study with this aim. However, we were unable to differentiate post-vaccination symptoms per se from superimposed SARS-CoV-2 infection robustly. Although two models, LR and RF, showed ROC AUC significantly greater than 0.5, neither came close to approaching clinical utility - for most clinical tests, conventionally given as 0.8; but for a highly infectious agent with devastating consequences from community spread the necessary AUC is much higher. We consider employing more complex model to improve the current results; however, we concluded that our sample size limited the use of such models. Were a larger dataset to be available, we agree this would be a potential approach.

Although one third of the one million vaccinated app users reported symptoms previously associated with COVID-19 early post-vaccination, only 4% of symptomatic individuals reported testing for SARS-CoV-2 even with allowance for delayed testing access. Considering those individuals who reported at least one of the symptoms fulfilling NHS criteria for testing (266,502 overall), 40% (107,929) were tested. During the study period, testing was widely available in the UK and it is unclear why more symptomatic people (including those with the widely advertised core symptoms of fever, persistent cough, and anosmia/dysosmia) were not tested [33]. Possible reasons for not testing, even among individuals presenting any of the core symptoms, were that some symptoms were increasing in prevalence with time, and some were significantly more common in negative individuals.

### Table 4

Model performance in the classification of COVID-19 status according to post-vaccination symptoms. Median values and percentiles (0.25 and 0.75) are obtained across five folds. AUC - area under curve, in a receiver operating characteristic analysis.

| Model                           | Sensitivity | Specificity | ROC - AUC |
|---------------------------------|-------------|-------------|-----------|
| Bayesian Mixed-Effect Model     | 0.52 [0.47; 0.56] | 0.55 [0.47; 0.60] | 0.52 [0.47; 0.56] |
| Logistic Regression             | 0.63 [0.58; 0.67] | 0.67 [0.60; 0.72] | 0.62 [0.58; 0.67] |
| Random Forest                   | 0.61 [0.58; 0.64] | 0.63 [0.56; 0.69] | 0.66 [0.61; 0.70] |
| Symptom clustering              | 0.51 [0.49; 0.56] | 0.67 [0.60; 0.73] | 0.50 [0.47; 0.55] |
| NHS screening criteria          | 0.48 [0.48; 0.48] | 0.62 [0.61; 0.63] | 0.47 [0.47; 0.48] |

Figure 4. ROC-AUC performance for the different models. Mean value (line) and 95% CI (shadow area) of the models’ performance, given the predictions obtained over the cross-validation scheme (5-folds) adopted for the validation. bMEM: Bayesian Mixed-effect Model (blue); LR: Logistic Regression (yellow); RF: Random Forest (red); Symptom clustering (green) and NHS screening criteria (grey).
symptoms, include the lack of knowledge about when and where to test, as well as the absence of severe and/or multiple symptoms [33]. Additionally, there is currently no specific guidance given to vaccine recipients either highlighting the possibility of post-vaccination infection or on when to access testing in the post-vaccine period, both of which might also affect the decision of going for testing among vaccinated individuals [33,34]. Conversely, of 149 individuals who tested positive, only 62 (41%) had symptoms that met current UK testing criteria. We do not know why the other 88 positive individuals were tested (e.g., contact tracing, routine workplace testing, direct personal request through the app).

Our data also suggest sensitivity of using core symptoms to justify testing for COVID-19 may be lower post-vaccination than in pre-vaccination times (here 48%, previously 73%). [35]. Although individuals with core symptoms were more likely to test positive than those without, the overall sensitivity and AUC suggests current UK testing policy is suboptimal for pandemic management particularly now that rapid testing capacity is much greater than when these criteria were established [35]. Notably, current UK testing criteria are more limited than WHO guidelines [2] and those of many other jurisdictions of similar GDP (including France, Germany, USA, and Australia).

Although there were some differences in symptom prevalence and distribution between positive and negative individuals, these could not be used robustly to discriminate between groups, including using machine-learning. We also considered time of symptom onset and symptom duration post-vaccination, (previous trials and post-marketing observational data have examined these parameters but not with respect to SARS-CoV-2 status) [4–6,8,16]. Whether positive or negative, median symptom peak burden was day 3 in both groups, concordant with vaccination side-effect profiles reported previously. [4,6,8,16] As time progressed, some symptoms appeared to become more common in the positive group only (e.g., persistent cough, hoarse voice), the timing of which coincides with the serial interval and incubation period of SARS-CoV-2 [26]. Note that no formal statistical analyses were undertaken on this point, and the results regarding symptom duration are descriptive only. Future work should assess statistical differences in symptom duration for the two groups, in larger cohorts. However, the critical public health importance of identifying and isolating cases early, and the lack of clear-cut differences between infected and non-infected symptomatic individuals, does not allow the luxury of a watch-and-wait approach.

We do not know the circumstances contributing to infection of the positive group (whether prior to, peri-, or immediately post-vaccination), noting here that our requirement for a positive rtPCR or LFAT result confirmed only recent infection. The serial interval and incubation period for SARS-CoV-2, and the high prevalence of asymptomatic infection, mean individuals could have been infected before vaccination. Although current UK vaccination guidelines do not require individuals to be completely asymptomatic at time of vaccination (only not “acutely unwell”), our inclusion criteria required individuals to be asymptomatic at time of vaccination. It is also possible that infection was contracted whilst getting vaccinated. Nosocomial infection with SARS-CoV-2 has been reported in the UK [37,38] including many health care workers infected at their workplace. [39,40]. Here we would emphasise strongly that data from ourselves and others indicate that even if infected peri- or post-vaccination the course of COVID-19 is much less severe in vaccinated vs. unvaccinated individuals; [41–45] and - acknowledging that a small percentage of vaccinated and symptomatic individuals were tested - our data demonstrated only 150 cases of confirmed infection (1%) in 14,842 tested individuals from over 1 million vaccinated app users. It is also possible that individuals changed their behaviour immediately after vaccination, increasing infection risk and contracting SARS-CoV-2 prior to acquiring adequate immunity. Here, data from previous infections suggest vaccination reduces adherence to other public health measures, [46] with pre-prints suggesting that this is also occurring after vaccination against SARS-CoV-2 [47,48].

Overall, CSS app users are not fully representative of the UK population (younger, more likely to be female, of higher educational status, lack of ethnical minorities, and over-representative of healthcare workers). [15] Although the population in the current study shares some of these biases, the median age of vaccinated individuals at the time of our analysis (64 years) was older than for app users overall (47 years), which is not surprising as the UK vaccination schedule began with the oldest individuals in the community. We considered the implications of this with respect to the likelihood of an infected person presenting for testing: although asymptomatic SARS-CoV-2 infection is well-recognised, it is less common in older people [35]. We also acknowledge that different economic and cultural experiences may influence presentation of SARS-CoV-2 infection and reporting of post-vaccination side-effects. [40] Our sample is mainly composed of White individuals, which may affect applicability of our results to populations with different ethnicities. The inclusion of other ethnicities was not possible, as we did not have any individuals who fulfilled the entry criteria who identified as Black, Asian or Minority Ethnicity. We are unsure whether this reflects the known population bias amongst app participants, the strict criteria for inclusion in this study, and/or other reason [33,46].

Our approach in comparing symptom profiles for individuals testing positive or negative for SARS-CoV-2 required a 1:1 matched population, so that comparison of symptom prevalence was fair and unbiased by the greatly different sample sizes of the two populations. However, this methodological choice is less reliable when used for the outcome of SARS-CoV-2 test prediction; and the forced balance of the classes can lead to an overestimation of the likelihood of being positive in the modelling. Thus, we have also presented extended analyses, using both boot-strapping, and entire-cohort approaches in the Supplementary Results. We acknowledge that the predictive power of our optimised models may be hampered if there are brand-specific post-vaccination side-effects, which were not considered during model optimisation [16,31]. However, although there were some differences in frequencies, most symptoms were reported in both PB and O-AZ pivotal trials [45,8,10]. We also did not consider type of SARS-CoV-2-testing (rtPCR vs. LFAT, noting here that LFAT has been confirmed to be an accurate alternative to rtPCR testing, [49,50] particularly in symptomatic individuals, with a specificity of 89.1% (95%CI: [86.3%, 91.9%]) and sensitivity of 5.4% (95%CI: [94%,96.8%]), or mode of testing access (NHS vs. ZOE-request), may also contribute variability to these models [50]. Here, our modest numbers preclude sensitivity testing. To avoid potential bias in the assessment of the performance of the models, we kept the same ratio of the population tested with either SARS-CoV-2 test in both training and validation sets. Lastly, we did not consider the different vaccine type in our models, since the number of individuals having each vaccine is approximately the same (O-AZ: 73, PB: 77), the bias caused by the vaccine type is almost insignificant.

Our analyses do not consider the impact of COVID-19 prevalence in UK at the time of the vaccination. Positive and negative predictive values (PPV, NPV) for a test depend not only on test sensitivity and specificity but also on population prevalence of disease. The rapidly changing prevalence of SARS-CoV-2 infection in the UK and the pace of vaccination delivery over the time period of this study limits our capacity to provide accurate PPV and NPV. Further analyses, particularly in populations with higher prevalence of infection and/or higher symptom burden and severity of COVID-19, may result in better differentiation of early signs of infection from post-vaccination side-effects.

A strength of our study was our very large cohort of vaccinated participants, in a country that was an early adopter of vaccination; and our timeframe included the UK pandemic “third wave”. Prospective real-time symptom logging through the app minimised recall
bias; and our symptom assessment included direct ascertainment of core symptoms for accessing UK testing. However, the sharp decline in cases in the first six months of 2021 resulted in only 149 positive cases to inform our modelling. We acknowledge this number is small - though we were also able to draw upon large numbers of tested negative individuals for comparisons, reinforcing the consistency and generalisation of our results. Additionally, in this observational study no formal a priori sample size calculation was computable; posterior analysis concluded the necessary sample size for this study was 31 individuals in each population (positive versus negative COVID-19 infected individuals), for 95% confidence interval for the difference in proportions who are positive versus negative with a margin of error no more than 5%; with our population size (149 individuals per group), our study has a power > 0.80. Power analysis for the comparison between doses was also not computed, since we considered that four subjects are not representative of the vaccinated population. We also acknowledge that the demographic features of the app population especially those parameters considered for model estimation (e.g., age, gender, BMI) may be different in other populations within the UK and elsewhere.

The implication of our results will vary depending on the population prevalence of SARS-CoV-2 and pace of vaccination roll-out. For example, at the time of writing New Zealand has negligible community spread of SARS-CoV-2 and is still early in vaccination roll-out. It would be relevant to repeat this study in these different circumstances to maximise its utility to different populations, noting that the translation of this study to other social and demographic contexts could be challenging. The use of new technologies (either the app used here [modified for country of use, according to language, cultural context and/or other unique population features] or other technologies) may not be feasible in some countries, particularly those without policies for safe use of personal and medical information for research. We also note that testing of all symptomatic individuals comes at a cost (e.g., testing kits, infrastructure). On the other hand, our data show that individuals manifesting symptoms post-vaccination cannot be assumed to be uninfected (and, thus, non-infectious). Providing a testing kit to all individuals post-vaccination, with guidance as to when to test, may or may not prove tenable, noting here the widespread use of lateral flow testing currently in the asymptomatic population. The UK is a resource-rich country; the impact of our results in countries with fewer health resources needs careful consideration.

In conclusion, post-vaccination symptoms cannot be distinguished with clinical confidence from early SARS-CoV-2 infection. Our study highlights the critical importance of testing symptomatic individuals - even if recently vaccinated - to ensure early detection of SARS-CoV-2 infection and help prevent future waves of COVID-19.

Authors’ contribution

LSC, MM and ELD contributed to study concept and design. CHS, JCP, BM, TS, CJF, SO contributed to acquisition of data. All the authors had access to the raw data underlying the study. LSC, JD, ELD and MM accessed and processed the raw data. LSC, JD, ELD contributed to data analysis and verified the underlying data. LSC, MO, MM and ELD contributed to drafting of the manuscript. All authors contributed to interpretation of data and critical revision of the manuscript. MM and ELD contributed to study supervision. LSC, MM and ELD decided to submit, under the Contributor’s statement.

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Data sharing:

Data collected in the COVID Symptom Study smartphone application are shared with other health researchers through the UK National Health Service-funded Health Data Research UK (HDRUK) and Secure Anonymised Information Linkage consortium, housed in the UK Secure Research Platform (Swansea, UK). Anonymised data are available to be shared with researchers according to their protocols in the public interest (https://web.who.is/healthdataatwork.org/dataset/fddcb382-3051-4394-8436-b92295f14259). The code is available in: https://gitlab.com/KCL-BMEIS/covid-zoe/vaccination.

Declaration of Competing Interest

CJS reports grants from the Chronic Disease Research Foundation (CDRF), Medical Research Council (MRC) and Wellcome Trust during the conduct of the study. ELD reports being a co-leader of the KCL COVID Symptom Study Biobank, a research-supported biobank of individuals with Long COVID, administered through King’s College London, during this study. EM reports grants from the MRC, as a personal fellowship. CHS reports grant from Alzheimer’s Society. CH, SS, LP, AM report other from ZOE Limited and are employed by ZOE Limited, during the conduct of the study. TS is a scientific advisor to ZOE Limited. CHS reports grants from Alzheimer’s Society, during the conduct of the study. SO reports grants from the Wellcome Trust, Innovate UK (UKRI), and Chronic Disease Research Foundation (CDRF), during the conduct of the study. All the other authors have no conflicts.

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Supplementary materials

Supplementary material associated with this article can be found, in the online version, at doi:10.1016/j.eclinm.2021.101212.

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