Vaccination strategy for preventing the spread of SARS-CoV-2 in the limited supply condition: A mathematical modeling study

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Abstract
To mitigate SARS-CoV-2 transmission, vaccines have been urgently approved. With their limited availability, it is critical to distribute the vaccines reasonably. We simulated the SARS-CoV-2 transmission for 365 days over four intervention periods: free transmission, structural mitigation, personal mitigation, and vaccination. Sensitivity analyses were performed to obtain robust results. We further evaluated two proposed vaccination allocations, including one-dose-high-coverage and two-doses-low-coverage, when the supply was low. 33.35% (infection rate, 2.68 in 10 million people) and 40.54% (2.36) of confirmed cases could be avoided as the nonpharmaceutical interventions (NPIs) adherence rate rose from 50% to 70%. As the vaccination coverage reached 60% and 80%, the total infections could be reduced by 32.72% and 41.19%, compared to the number without vaccination. When the durations of immunity were 90 and 120 days, the infection rates were 2.67 and 2.38. As the asymptomatic infection rate rose from 30% to 50%, the infection rate increased 0.92 (SD, 0.16) times. Conditioned on 70% adherence rate, with the same amount of limited available vaccines, the 20% and 40% vaccination coverage of one-dose-high-coverage, the infection rates were 2.70 and 2.35; corresponding to the two-doses-low-coverage with 10% and 20% vaccination coverage, the infection rates were 3.22 and 2.92. Our results indicated as the duration of immunity prolonged, the second wave of SARS-CoV-2 would be delayed and the scale would be declined. On average, the total infections in two-doses-low-coverage was 1.48 times (SD, 0.24) as high as that in one-dose-high-coverage. It is crucial to encourage people in order to improve vaccination coverage and establish immune barriers. Particularly when the supply is limited, a wiser strategy to prevent SARS-CoV-2 is equally distributing doses to the same number of individuals. Besides vaccination, NPIs are equally critical to the prevention of widespread of SARS-CoV-2.

Keywords
coronavirus disease 2019 (COVID-19), nonpharmaceutical intervention (NPI), severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2), stochastic dynamic model, vaccination
1 | INTRODUCTION

The outbreak of COVID-19 caused by severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) has led to an unprecedented public health and economic crisis worldwide since December 2019. To mitigate the spread, a variety of nonpharmaceutical interventions (NPIs) have been implemented, including screening and isolation, travel restriction, remote schooling and work distancing. Although these efforts are beneficial to control the spread in a short term, globally, as of March 4, 2022, there have been 440807756 confirmed cases of COVID-19, including 5978096 deaths, reported by the World Health Organization (WHO). Additionally, in many countries, relaxation of NPIs has led to a resurgence of the epidemic as herd immunity has not been reached thus far. A long-term solution, such as vaccines that protect from SARS-CoV-2 infection, remains urgently needed.

The competition for developing vaccines against SARS-CoV-2 started in early 2020 and more than 50 companies began development of the first vaccine. At present, 14 vaccines have been approved for urgent use, including 3 nucleic acid vaccines (Curevac, Moderna, and BioNTech), 3 inactivated virus vaccines (Bharat, Sinovac, and Sinopharm), and 8 viral-vectored vaccines (Clover Biopharmaceutical, Serum institute, Novavax, Sanofi, AstraZeneca, Janssen, Gamaleya, and CanSino). As of March 5, 2022, a total of 10704043684 vaccine doses have been administered globally. The benefits of an effective vaccine for individuals and their communities have resulted in widespread demand, so it is critical that decision-making on vaccine distribution is well-motivated, particularly in the initial phases when vaccine availability is limited.

As the basis of regulatory approvals, the initial vaccine was released in early 2021, there are two main suggested approaches to vaccine prioritization: (i) directly vaccinate those at highest risk for severe outcomes and (ii) protect them indirectly by vaccinating those who do the most transmitting. Model-based investigations of the trade-offs between these strategies have found that the optimal balance between direct and indirect protection depends on both vaccine efficacy and supply, recommending direct vaccination of older adults for low-efficacy vaccines and for high-efficacy but supply-limited vaccines. Rather than comparing prioritization strategies, others have compared hypothetical vaccines, showing that even those with lower efficacy for direct protection may be more valuable if they also provide better indirect protection by blocking transmission. Prioritization of transmission-blocking vaccines can also be dynamically updated on the basis of the current state of the epidemic, shifting prioritization to avoid decreasing marginal returns. However, the strategies of prioritizing and optimizing doses complement are highly dependent on different vaccine efficacy (VE) and durability of immunity. An optimal resource allocation will largely reduce the transmission economically.

To evaluate the vaccine allocation strategies, we built an age and occupation stratified SEIRS (susceptible, exposed, infectious, recovered, and susceptible) model. Since age has been shown to be an important correlate of susceptibility, seroprevalence, severity, and mortality, this model includes an age-dependent contact matrix, susceptibility to infection, and infection fatality rate (IFR), allowing us to estimate the cumulative incidence of SARS-CoV-2 infections by means of forward simulations of disease dynamics.

2 | METHODS

An individual-based dynamic model, stratified by age and occupation, was built to simulate the transmission of SARS-CoV-2 based on the epidemiological progression of susceptible-exposed-infectious-removed-susceptible (SEIRS) structure. This model includes NPIs aimed at mitigating the epidemic.

2.1 | Model construction

In the model simulation, each healthy individual (susceptible) has a chance of being infected with SARS-CoV-2 under the force transmission rate depending on the number of daily contacts and the probability of SARS-CoV-2 being transmitted from an infected to uninfected contact. Once infected, the individual enters the exposed period. At the end of the exposure period, an individual will become infectious, either symptomatic or asymptomatic. Most infectious individuals recover but some will die (with IFR). We assumed that the recovered individual would be re-infected after waned immunity, including natural immunity.

The population was grouped by age and occupation, and interactions between groups (or individuals) were simulated, taking into account the number of daily contacts. The global age structure was from the United Nations (2019), people aged 0-100 years were divided into eight groups (0-4, 5-14, 15-24, 25-34, 35-44, 45-54, 55-64, and over 65). The occupations (nonworker, student, worker, and others) and social contact patterns (home community, school, workplace, and other contacts) were then assigned according to the economic structure from the Chinese 2010 and 2020 census data.

Each individual was assigned a social contact parameter (location-specific contact matrices, including home community, school, workplace, and other) by their occupation. This study assumed an individual has no age-dependence in transmissibility, and the likelihood of viral exposure varied by individuals depending on the number of infectious people in their social network. We considered age-stratified contact matrices from the BBC pandemic project in describing the average daily effective number of contacts that an individual has with others. Age-stratified IFRs were collected from a model-based analysis under New York City during the 2020 spring pandemic wave.

2.2 | Infection parameters

Each modeled individual was ascribed demographic characteristics (e.g., age and occupation) and epidemiological characteristics (e.g., exposed
period, infectious period, symptomatic or asymptomatic status, recovery or death from infection). We incorporated asymptomatic infections into this model, although it remains unclear as to the asymptomatic rate and the extent asymptomatic patients contribute to viral transmission.26–28 All the detailed information is presented in the supplementary materials.

Although existing studies have been focusing on the antibody responses to SARS-CoV-2, it is still vague in the natural and vaccine acquired immunity.8,30 Therefore, we considered 90 and 120 days of acquired immunity priority before getting susceptible again in our simulation.21–33 The transmission was dependent upon the period of exposure, period of infectiousness, and basic reproductive number (R0).34–37 We also incorporated an asymptomatic rate denoting the probability of an infected case being asymptomatic and assumed a reduced rate of infection for asymptomatic cases.24,26,35,36,38 After the infectious period, an individual has the chance to recover or die.34,39

2.3 | Interventions

In this model, we mimicked the transmission in four assumed periods. The first period is the free transmission period, during which no NPIs are implemented. The second period is the structural mitigation period, during which structural NPIs including isolation and quarantine are implemented. In this period, the infected individuals are isolated until they recover or die, and their close contacts are quarantined for 14 days. The accuracy rate of screening (sensitivity) was considered in this modeling.30,41 The third period is the personal mitigation period, during which personal NPIs including social distancing, mask-wearing, and hand washing are in place. The individuals begin to change their protection behavior depending on the government policy adherence rate. The assumed efficacies of mask-wearing and hand washing are introduced into the model.42 The final period is the vaccination period. In our study, we introduced VE into our simulation to predict the effect of the vaccine, which is associated with the NPIs adherence rate.43 Risk compensation was considered in our model. When the coverage rates reach the target value (≥50%), the adherence of personal NPIs becomes 50% lower among vaccinated people. In this situation, the theoretical immunity is elicited after all recommended doses of the vaccine are injected. We proposed two scenarios of vaccine-allocation strategies (assumed two injections) based on the limited vaccine supply. Scenario 1 (one-dose-high-coverage): distributing the two doses to two people, each person with one dose (lower VE); Scenario 2 (two-doses-low-coverage): distributing the two doses to one person, each person with two doses (higher VE). Among these scenarios, we assumed time varied supplies, increased by 0.5% per day.

2.4 | Simulation and sensitivity analysis

We simulated 100 million individual-level transmission events by repeatedly generating contact distributions for a primary case and randomly generating infections among these contacts. This process was repeated a thousand times, and each simulation was to generate a set of epidemic trends including (1) daily newly confirmed cases and (2) daily total cumulative confirmed cases. Our primary sensitivity analyses were to the level of vaccination coverage (60% and 80%), the adherence rate of NPIs (50% and 70%), and the asymptomatic rate (30% and 50%). We further assumed the (natural and vaccinated) immunity waned after 90 and 120 days.

3 | RESULTS

3.1 | Main analysis

Overall, in ages, the distribution of infections was 1.06% (0–4 years old), 7.13% (5–14 years old), 19.14% (15–24 years old), 23.16% (25–34 years old), 19.82% (35–44 years old), 16.05% (45–54 years old), 9.13% (55–64 years old), and 4.50% (over 65 years old). In all, 53.64% of the dead were people over 35 years old. Most of the dead were students (16.09%) and workers (70.00%). In occupations, 12.88%, 17.71%, 67.28%, and 2.12% were accounted for by nonworker, students, workers and others (Figure 1).

On average, the infection rate (in 100 million people) reduced from 2.68 (95% confidence interval [CI], 2.25–3.11) to 2.36 (95% CI, 1.94–2.78) as the NPIs adherence rate rose from 50% to 70% (Figure 2). 33.35% (95% CI, 22.22–43.78) and 40.54% (95% CI, 27.75–52.25) confirmed cases can be avoided if 50% and 70% of people followed the instruction.

As the vaccination coverage reached 60% and 80%, the number of infections can be reduced by about 32.72% (95% CI, 22.03–43.41) and 41.19% (95% CI, 30.32–52.06), compared to the number without vaccination. The average IFRs were 0.68% (95% CI, 0.66–0.69) and 0.58% (95% CI, 0.57–0.58) when the vaccination coverage were 60% and 80%. When the durations of immunity were 90 and 120 days, the infection rates were 2.67 (95% CI, 2.20–3.14) and 2.38 (95% CI, 1.99–2.77).

Furthermore, our model suggested that with the increase in the asymptomatic infection rate, the prevention and control of SARS-CoV-2 was becoming more and more unfavorable. As the asymptomatic infection rate rose from 30% to 50%, the infection rate increased from 1.73 (95% CI, 1.59–1.87) to 3.31 (95% CI, 3.14–3.47), which was 0.92 (SD, 0.16) times higher. However, no matter how high the asymptomatic rate was, higher NPIs adherence and higher vaccination coverage can always prevent more SARS-CoV-2 infections (Figure 2).

According to our simulation, relying on a 30% asymptomatic rate, 70% NPIs adherence rate, 80% vaccination coverage, and 180 days of immunity, the infection rate remained at 1.42% (total confirmed cases, 142 243; 95% CI, 134 598–152 471), which reduced the infections most.

3.2 | Vaccine distribution scenarios

The epidemiological impacts of the different dosing scenarios on mitigating the SARS-CoV-2 spread, when the vaccine supply was limited, are shown in Figure 3.
Assuming the adherence rate of NPIs was 70%, supposing the same amount of vaccines were available, under scenario 2 (completely vaccinated), relying on an immunity duration of 90 days, the IFR and infection rate were 0.74 (SD, 0.01%) and 3.38% (SD, 0.83%), whereas the values were 0.72% (SD, 0.01%) and 2.76% (SD, 0.90%) when immunity waning after 120 days. Considering scenario 1 (partially vaccinated), when the duration of immunity was 90 days, the IFR and infection rate were 0.74% (SD, 0.01%) and 2.69% (SD, 1.16%), whereas the values were 0.72% (SD, 0.01%) and 2.37% (SD, 1.12%) when immunity waning after 120 days (Table 1).

Figure 3 indicated as the duration of immunity prolonged, the second peak of SARS-CoV-2 will be delayed and the scale will be declined. On average, the total infections in scenario 2 was 1.48 times (SD, 0.24) higher than that in scenario 1.

Considering the same amount of supplies, and 20% and 40% vaccination coverage of scenario 1, the total number of infections were 270 256 (95% CI, 171 631–368 870) and 235 254 (95% CI, 113 440–357 060); corresponding to the scenario 2 with 10% and 20% vaccination coverage, the number of infections were 321 535 (95% CI, 241 342–401 658) and 291 523 (95% CI, 193 086–389 914).

**4 | DISCUSSION**

In this modeling, the findings showed that the vaccines and NPIs substantially contributed to the SARS-CoV-2 transmission control. With higher vaccination coverage and NPIs adherence rate, more infections can be avoided. Compared to no vaccination, the number of infections can be reduced by 40% or 26% if the vaccination coverage reached 80% or 60%). Furthermore, when the adherence rate increases from 50% to 70%, 28% of infected cases can be further saved. To acquire the theoretical vaccine immunity, all recommended vaccine doses should be injected. However, when the vaccine (assumed more than two injections) supply was limited, the partially vaccinated strategy was superior to the completely vaccinated since it helped to reduce the infections by 67.57%. Finally, the transmission...
of SARS-CoV-2 was more difficult to control as the asymptomatic infection rate increased.

Vaccines, which can induce and establish the immune response against SARS-CoV-2, are crucial to the prevention and mitigation of morbidity and mortality cases by infection.8,30 Multiple candidate vaccines, including nucleic acid vaccines (Moderna—mRNA-1273; BioNTech—BNT162b1 and BNT162b2), inactivated virus vaccines (Sinovac—CoronaVac; Sinopharm—BBIBP-CorV and Vero cells), live attenuated vaccines, protein or peptide subunit vaccines, and viral-vectored vaccines (AstraZeneca—AZD1222; Janssen—Ad26. COV2.S; Gamaleya—rAd5 and rAd26; CanSino—Ad5CoV), are being developed and tested.43 Each type has advantages and disadvantages and vaccine manufacturers have published articles to present the findings of their Phase III trials, each involving tens of thousands of participants, and commenced in various geographical locations.7,9–11 These results have shown the VE are over 90% with all recommended doses completed. Even after a partial dose, the VE can still reach 70%. In addition, the antibody responses from vaccination enhance the protection in preventing the disease progression.7,9–11

SARS-CoV-2 variants are circulating globally and quickly became dominant in countries, such as the United Kingdom (WHO label, Alpha; Pango lineage B.1.1.7), South Africa (Beta; B.1.351), Brazil (Gamma; P.1), India (Delta and Kappa, B.1.617.2 and B.1.617.1), America (Iota, B.1.526), Peru (Lambda; C.37), and multiple countries (Omicron; B.1.1.529).44 The reason for the strong infectivity of SARS-CoV-2 variants is that it can escape the neutralizing antibodies produced by the immune system, and the more lethal variants could substantially decrease the net benefit of vaccination.45–52 Therefore, the current issues on whether a third dose of enhanced vaccine is needed and when to vaccinate it has also become a topic of public concern.53–55 Moreover, how to cope with the evolution of the SARS-CoV-2 and develop a vaccine with a “broad-spectrum effect” in avoiding virus escape is an important problem that researchers need to urgently solve.56,57 The variants reduced the VE to varying degrees, but the vaccine is still protective.58,59

\[ \text{TABLE 1 Infection rate of SARS-CoV-2 among different scenarios} \]

| Scenario 1 (partially vaccinated) | Vaccine coverage | \(20\%\) | \(40\%\) |
| --- | --- | --- | --- |
| Asymptomatic rate | Days | 30% | 90 | 1.97 | 1.43 |
| | | 120 | 1.72 | 1.13 |
| | | 50% | 90 | 3.78 | 3.56 |
| | | 120 | 3.34 | 3.29 |

| Scenario 2 (completely vaccinated) | Vaccine coverage | \(10\%\) | \(20\%\) |
| --- | --- | --- | --- |
| Asymptomatic rate | Days | 30% | 90 | 2.84 | 2.50 |
| | | 120 | 2.30 | 1.71 |
| | | 50% | 90 | 4.19 | 3.97 |
| | | 120 | 3.53 | 3.48 |

FIGURE 3 Daily SARS-CoV-2 incidence rate among the population. Focusing on 70% NPIs adherence rate and considering the same amounts of limited vaccine supplies. The daily incidence rates with different asymptomatic rates and durations of immunity are represented as black and red curves, respectively. The panels in the left and right columns represent different scenarios.

Whichever vaccine appears, rational allocation of resources is very important. The current vaccination modeling research is mainly focusing on two directions. First, to minimize the deaths, Bubar et al.12 recommended older adults enjoy a vaccine priority due to its higher fatality rate. Second, to mitigate the spread, Yang et al.13 suggested several essential workers could be prioritized for...
vaccination to maintain essential services in the early phase of a vaccination program due to its higher contacts. In our model, we propose an interesting idea that when the vaccines are lacking in the early stage, to maximize the coverage, the total number of doses to be vaccinated should be equally distributed to the same amount of people even if it reduces the VE so that individuals can quickly establish an immune barrier.

All kinds of NPIs have been introduced in mitigating the transmission of SARS-CoV-2. The major NPIs, including isolation and quarantine, social distancing, mask, and hand washing, are recommended by WHO. Isolating confirmed cases stops the offspring generating and effectively blocks the transmission of SARS-CoV-2, and the contact tracing helps to minimize the potential transmission from second cases. However, this NPI is limited substantially by delays from testing of index cases to the tracing of their contacts, because secondary cases might have been transmitting for a number of days in the community during the time that contact tracing is taking place. In addition, the delays in isolating confirmed cases from the infected date to the symptom onset or hospitalization is another issue, not to mention the sensitivity of the screening test. Moreover, asymptomatic cases do not know they are carriers and will not attend hospitals to conduct self-screening. Thus, with the higher asymptomatic rate of SARS-CoV-2, it is not surprising that this NPI alone could not contain the SARS-CoV-2 pandemic.

Since the transmission routes, including droplet and contact transmission, of SARS-CoV-2, have been identified, several personal health behaviors, such as social distancing, mask-wearing, and hand washing, have been encouraged by WHO to avoid transmission. These NPIs have been proven effective against SARS-CoV-2 in our modeling, as shown in other previous research. By changing their personal behaviors, such as reducing social distancing, mask-wearing, and handwashing, people can lower the risk of potential SARS-CoV-2 infections or reduce the probability of spread to susceptible individuals. However, the effectiveness of these NPIs can be limited by low adherence. For example, due to the government policy, people who are living in China have a higher NPIs adherence rate and it achieves a lower total infection even with a larger population. For those countries with moderate or lower adherence rates, the governments should advocate the importance of these protective measures in preventing the SARS-CoV-2 transmission. In addition, maintaining a high level of mask utilization is also necessary and masks should be replaced frequently for a permanent protective effect. Proper use and disposal of masks is also essential to avoid increasing risk of transmission.

Our study also has several limitations. First, in this modeling, we did not consider the time-varied antibody responses and viral load dynamics. Due to its RNA virus structure, SARS-CoV-2 is continuously mutating. A much clear genomics-informed response and serology data should be further adopted for simulation. After adding the distribution of virus load and the immune waning into the model, we can much better predict the future epidemic trend and make a response early. Second, the risk compensation after vaccination should be considered. In another ongoing study, we have found that among the people who have been vaccinated, they have begun neither to use protections nor adhere to NPIs. In other words, we know that vaccines can protect against death after infection, but whether they can reduce the spread of SARS-CoV-2 seems to be an interesting research direction. Finally, the seasonal dependent transmission patterns should be introduced into the model. Based on little published data, the impact of unknown factors such as temperature or humidity changes could not be assessed in our model. However, two similar human coronaviruses, HCoV-OC43, and HCoV-HKU1, cause annual winter-time outbreaks of respiratory illness in temperate regions, suggesting that climate and host behaviors may facilitate transmission as is true for influenza.

In conclusion, although the world has taken many different NPIs and vaccines to control and prevent the SARS-CoV-2 epidemic, the epidemic of SARS-CoV-2 has led to the current global large-scale spread and there is no sign of complete control. This study investigates the effectiveness of vaccination and NPI strategies in various situations. It is crucial to encourage people to vaccinate in order to improve vaccination coverage and establish immune barriers. Particularly, when the vaccine supply is limited, an optimal strategy to prevent SARS-CoV-2 transmission is equally distributing doses to the same number of individuals. Besides vaccination, NPIs are equally critical to the prevention of widespread of SARS-CoV-2.

AUTHOR CONTRIBUTIONS
Huachun Zou and Yuelong Shu conceived the idea and protocol. Yi-Fan Lin built the model, collected data, finalized the analysis, interpreted the findings, and wrote the manuscript. Yuwei Li conducted the visualization. All authors have contributed to the interpretation of data, and study findings and approved the final manuscript.

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CONFLICTS OF INTEREST
The authors declare no conflicts of interest.

DATA AVAILABILITY STATEMENT
The data that support the findings of this study are available from the corresponding open accessed websites.

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