

# Epidemiological and evolutionary considerations of SARS-CoV-2 vaccine dosing regimes

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1       **As the threat of Covid-19 continues and in the face of vaccine dose shortages**  
2       **and logistical challenges, various deployment strategies are being proposed to**  
3       **increase population immunity levels. How timing of delivery of the second dose**  
4       **affects infection burden but also prospects for the evolution of viral immune**  
5       **escape are critical questions. Both hinge on the strength and duration (i.e.**  
6       **robustness) of the immune response elicited by a single dose, compared to nat-**  
7       **ural and two-dose immunity. Building on an existing immuno-epidemiological**  
8       **model, we find that in the short-term, focusing on one dose generally decreases**  
9       **infections, but longer-term outcomes depend on this relative immune robust-**  
10       **ness. We then explore three scenarios of selection, evaluating how different**  
11       **second dose delays might drive immune escape via a build-up of partially im-**  
12       **mune individuals. Under certain scenarios, we find that a one-dose policy may**  
13       **increase the potential for antigenic evolution. We highlight the critical need to**  
14       **test viral loads and quantify immune responses after one vaccine dose, and to**  
15       **ramp up vaccination efforts throughout the world.**

16       As the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) betacoronavirus ( $\beta$ -  
17       CoV) pandemic continues, the deployment of safe and effective vaccines presents a key inter-  
18       vention for mitigating disease severity and spread and eventually relaxing non-pharmaceutical  
19       interventions (NPIs). At the time of writing, eleven vaccines have been approved. We focus on  
20       vaccines from Pfizer/BioNTech, Moderna, and Oxford/AstraZeneca. The first two elicit adap-  
21       tive immunity against SARS-CoV-2 in response to the introduction of messenger ribonucleic  
22       acid (mRNA) molecules that encode the spike protein of SARS-CoV-2 (1), and appear to offer  
23       greater than 95% (Pfizer/BioNTech (2), approved in 55 countries) and 94% (Moderna (1), ap-  
24       proved in 37 countries) protection against symptomatic coronavirus disease 2019 (COVID-19).  
25       Both of these mRNA vaccines were tested in clinical trials according to a two-dose regime with

26 dose spacing of 21 and 28 days for the Pfizer/BioNTech and Moderna platforms, respectively.  
27 The Oxford/AstraZeneca vaccine uses a non-replicating adenovirus vector, and has also been  
28 tested in clinical trials according to a two-dose regime with a target 28-day inter-dose period  
29 (although for logistical reasons some trial participants received their second dose after a delay  
30 of at least 12 weeks). Clinical trials indicated 62% – 90% efficacy for this vaccine according  
31 to the specific dose administered (3). While we base our parameter choices and modeling as-  
32 sumptions on these three vaccines, our results are generic across platforms.

33

34 As these vaccines have been distributed internationally, several countries including the  
35 UK (4) and Canada (5) have chosen to delay the second dose in an effort to increase the num-  
36 ber of individuals receiving at least one or in response to logistical constraints (6). Although  
37 a number of participants dropped out after a single dose of the vaccine in the Pfizer/BioNTech  
38 and Moderna trials, these studies were not designed to assess vaccine efficacy under these cir-  
39 cumstances, and Pfizer has stated that there is no evidence that vaccine protection from a single  
40 dose extends beyond 21 days (4). The Oxford/AstraZeneca clinical trials did include different  
41 dose spacings, and limited evidence suggests that longer intervals (two to three months) did not  
42 affect and may even have improved vaccine efficacy (3, 4). Ultimately, the consequences of de-  
43 viating from manufacturer-prescribed dosing regimes at the population scale remain unknown,  
44 but will hinge on immune responses.

45

46 While there has been significant progress in quantifying host immune responses following  
47 infection (7), substantial uncertainty regarding the strength and duration of both natural and  
48 vaccinal SARS-CoV-2 immunity remains. Previous work suggests that these factors will play  
49 a central role in shaping the future dynamics of Covid-19 cases (8). Future cases also create  
50 an environment for the selection of novel variants (e.g. (9–11)). Of particular concern is the

51 possibility of antigenic drift (via immune escape from natural or vaccinal immunity), especially  
52 if immunity elicited after a single vaccine dose is weaker than that of the complete two-dose  
53 regime. Consequently, the longer term epidemiological and evolutionary implications of these  
54 different SARS-CoV-2 vaccine dosing regimes are not yet clear; the immediate need for effec-  
55 tive mass vaccination makes understanding them critical to inform policy.

56

Figure 1: Description of the extended immuno-epidemiological model with one- and two-dose vaccination regimes (based on (8)). (A) Model flow chart depicting transitions between immune classes (see main text and Supplementary Materials for a full description of the immune classes and parameters). (B) Diagram of the inter-dose period ( $\frac{1}{\omega}$ ) considered between the first and second vaccine doses and its relationship to the rate of administration of the first vaccine dose  $\nu$ . The maximum achievable rate is  $\nu_0$  for a fully one-dose strategy, and  $\nu$  is assumed to decrease exponentially to its lowest value  $\nu_0/2$  when a fully two-dose strategy with inter-dose period corresponding to the clinical recommendation ( $L_{\text{opt}}$ ) is employed. (C) Representative schematic of societal composition of various immune classes for the SIR(S) model with no vaccination (left), the extended model with a short inter-dose period (middle), and the extended model with a long inter-dose period (right).

57 Here, we explore these epidemiological and evolutionary considerations with an extension  
58 of a recent immuno-epidemiological model for SARS-CoV-2 dynamics (8), depicted schemati-  
59 cally in Figure 1. Without vaccination, our model reduces to the Susceptible-Infected-Recovered-  
60 (Susceptible) (SIR(S)) model (8, 12), where individual immunity after recovery from primary  
61 infection may eventually wane, leading to potentially reduced susceptibility to secondary in-  
62 fections, denoted by the fraction  $\epsilon$  relative to a baseline level of unity. This parameter  $\epsilon$  thus  
63 titrates between the SIR (lifetime immunity,  $\epsilon = 0$ ) and SIRS (hosts regain complete suscep-  
64 tibility,  $\epsilon = 1$ ) paradigms. In this model extension (Fig. 1 and Supplementary Materials) we  
65 incorporate two vaccinated classes;  $V_1$  accounts for individuals who have received one dose of  
66 a SARS-CoV-2 vaccine and  $V_2$  tracks individuals who have received two doses. In the short  
67 term, we assume that both dosing options decrease susceptibility by fractions  $(1 - \epsilon_{V_1})$  (one

68 dose) and  $(1 - \epsilon_{V_2})$  (two doses), inferred from the clinical trial data (though the nature of the  
69 infecting variant may influence this); we also assume that  $I_V$  tracks infection following vacci-  
70 nation. We allow for vaccinal immunity to wane at separate rates ( $\rho_1$  (one dose) and  $\rho_2$  (two  
71 doses)), moving individuals to the partially susceptible immune classes  $S_{S_1}$  and  $S_{S_2}$  character-  
72 ized by (possibly different) levels of immune protection  $\epsilon_1$  and  $\epsilon_2$ . Infection following waned  
73 one-dose or two-dose vaccinal immunity is tracked by the immune classes  $I_{S_1}$  and  $I_{S_2}$ , respec-  
74 tively. We consider a continuous spectrum for the inter-dose period ( $\frac{1}{\omega}$ ), with an infinite value  
75 corresponding to a “one-dose strategy”, and model the rate of administration of the first dose  
76  $\nu$  as an increasing function of the inter-dose period (Fig. 1 and Supplementary Materials) to  
77 reflect the increase in available doses due to a delayed second dose. Thus, dosing regimes with  
78 longer inter-dose periods allow for higher coverage with the first dose.

79

80 We begin our analysis by studying the epidemiological impacts of the different dosing  
81 regimes on the medium-term temporal dynamics of Covid-19 cases. We then examine the poten-  
82 tial evolutionary consequences of dosing regime through the quantification of a time-dependent  
83 relative net viral adaptation rate ( $I_3$ ). This term is related to the strength of the conferred natural  
84 and vaccinal immunity (via either inducing selection through immune pressure or suppressing  
85 viral replication) as well as the sizes of classes of individuals experiencing infections after im-  
86 mune waning.

## 87 Epidemiological impacts

Figure 2: Illustrative time series of the fraction of the population vaccinated with one or two doses (top) (see note (14)), the fraction of total and severe infections (see (15)) (middle), and area plots of the fraction of the population comprising each immune ( $S_P, R, S_S, V_1, V_2, S_{S_1}, S_{S_2}$ ) or infection ( $I_P, I_S, I_V, I_{S_1}, I_{S_2}$ ) class (bottom) from the introduction of vaccination until 5 years after the pandemic onset. The immune and infection class colors are the same as those defined in Figure 1A. In all plots, the maximum rate of administration of the first vaccine dose is taken to be  $\nu_0 = 2\%$  and the vaccine is introduced at  $t_{\text{vax}} = 48$  weeks. We take  $\epsilon_{V_1} = 0.1$  and  $\epsilon_{V_2} = 0.05$  in keeping with data from clinical trials (2). The fraction of severe cases for primary infections, secondary infections, infection after vaccination, and infection after waned two-dose immunity are taken to be  $x_{\text{sev,p}} = 0.14$ ,  $x_{\text{sev,s}} = 0.07$ ,  $x_{\text{sev,V}} = 0.14$ , and  $x_{\text{sev,2}} = 0$ . The transmission rates and periods of NPI adoption are defined in the Supplementary Materials. The leftmost column corresponds to a one-dose vaccine strategy ( $\omega = 0$ ), followed by inter-dose spacings of 24 weeks, 12 weeks, and 4 weeks (rightmost column). (A) corresponds to an overall more pessimistic natural and vaccinal immunity scenario, with  $\epsilon = \epsilon_2 = 0.7$  and  $1/\delta = 1/\rho_2 = 1$  year. For a less effective one-dose vaccine (top section), we take  $\epsilon_1 = 0.9$ ,  $1/\rho_1 = 0.25$  years, and the fraction of severe cases associated with infection after waned one-dose immunity is  $x_{\text{sev,1}} = 0.14$ . For an effective one-dose vaccine (bottom section), we take  $\epsilon_1 = 0.7$ ,  $1/\rho_1 = 1$  year, and the fraction of severe cases associated with infection after waned one-dose immunity is  $x_{\text{sev,1}} = 0$ . (B) corresponds to an overall more optimistic natural and vaccinal immunity scenario, with  $\epsilon = \epsilon_2 = 0.5$  and  $1/\delta = 1/\rho_2 = 2$  years. For a less effective one-dose vaccine (top section), we take  $\epsilon_1 = 0.9$ ,  $1/\rho_1 = 0.5$  years, and the fraction of severe cases associated with infection after waned one-dose immunity is  $x_{\text{sev,1}} = 0.14$ . For an effective one-dose vaccine (bottom section), we take  $\epsilon_1 = 0.5$ ,  $1/\rho_1 = 2$  years, and the fraction of severe cases associated with infection after waned one-dose immunity is  $x_{\text{sev,1}} = 0$ .

88 As a base case, we consider a high latitude European or North American city with initial con-  
89 ditions that qualitatively correspond to early 2021 (see Supplementary Materials and Figures  
90 S5 and S6 for other scenarios, e.g. a high initial attack rate or almost full susceptibility), in  
91 addition to a seasonal transmission rate (16) with NPIs (see Supplementary Materials). Fur-  
92 thermore, the UK and Canadian policy is for a delayed second dose; they are not aiming for  
93 an “exclusively” one-dose policy. However, we explore the one-dose strategy as an extreme  
94 case for the ‘two-dose’ vaccines; it also encompasses a pessimistic situation of waning public

95 opinion on vaccination and individuals' own decisions to forgo the second dose. Finally, this  
96 one-dose policy could capture vaccines which only require a single dose, e.g. the Johnson &  
97 Johnson vaccine.

98

99 In Figure 2, we present potential scenarios for medium-term SARS-CoV-2 infection and  
100 immunity dynamics contingent upon vaccine dosing regimes. We start by assuming that vacci-  
101 nation occurs at a constant rate, and assume a relatively optimistic maximum rate of adminis-  
102 tration of the first dose of  $\nu_0 = 2\%$  of the population per week (see Supplementary Materials  
103 for other scenarios). Figures 2A and 2B correspond, respectively, to scenarios with weaker (and  
104 shorter) and stronger (and longer) natural and vaccinal adaptive immune responses. Thus, the  
105 former represents a scenario with higher secondary susceptible density than the latter. In each  
106 panel, the top and bottom sections consider poor and robust one-dose vaccinal immunity, re-  
107 spectively. The leftmost column represents a one-dose vaccine policy (captured in the model by  
108 infinite dose spacing), with dose spacing decreasing to 4 weeks in the rightmost column (i.e. a  
109 strict two-dose policy with doses separated by the clinical trial window corresponding to Mod-  
110 erna's recommendations for their vaccine, hereafter referred to as the "recommended two-dose  
111 strategy").

112

113 As expected, we find that broader deployment of widely-spaced doses is beneficial. Specif-  
114 ically, a one-dose strategy (or a longer inter-dose period) may lead to a substantially reduced  
115 'first' epidemic peak of cases after the initiation of vaccination (compare the leftmost top pan-  
116 els of Figs. 2A and 2B with the no vaccination scenarios in Figs. S1A and S1B). This result  
117 applies even if immunity conferred by one vaccine dose is shorter and weaker than that follow-  
118 ing two-doses (top panels of Figures 2A and 2B). However under these conditions of imperfect  
119 immunity, an exclusively one-dose strategy then leads to an earlier subsequent peak due to the

120 accumulation of partially susceptible individuals. When the rate of administration of the first  
121 dose is very high (Fig S4,  $\nu_0 = 5\%$  per week), this subsequent infection peak may be larger than  
122 that expected in the scenario with no vaccination. In general, the accumulation of partially sus-  
123 ceptible individuals with waned one-dose vaccinal immunity can be mitigated by implementing  
124 a two-dose strategy and decreasing the time between doses. Thus, in situations of a less effec-  
125 tive first dose where the second dose is delayed, it is important to ensure individuals eventually  
126 do obtain their second dose.

127

128 In line with intuition, longer and stronger immunity elicited after a single dose heightens the  
129 benefits of a one-dose strategy or of delaying the second dose (compare the top and bottom left-  
130 most panels of Figs. 2A and 2B). Additionally, the protective effects of adopting these strategies  
131 instead of the two-dose regime are maintained in the medium-term, with decreased burden in  
132 all future peaks. This is further summarized in Figure 3A, where the cumulative number of total  
133 and severe cases (right and left panels, respectively), from the time of vaccine initiation through  
134 the end of the five year period considered normalized by the burdens with no vaccination, are  
135 plotted as a function of the inter dose period and the one- to two-dose immune response ratio  
136  $x_e$  (see figure caption for details). When the immune response conferred by a single dose is  
137 nearly or as robust as that following two doses, total case numbers (Figure 3A, right panel) can  
138 be substantially reduced by delaying the second dose. However, for smaller values of  $x_e$ , larger  
139 inter-dose periods are associated with more cases. The reduction in the cumulative burden of  
140 severe cases is even more sizeable (Figure 3A, left panel) due to the assumed reduction in the  
141 fraction of severe cases for partially immune individuals. When vaccination rates are substan-  
142 tially lower (Fig S2,  $\nu_0 = 0.1\%$  per week and Fig S3,  $\nu_0 = 1\%$  per week), the benefits of a  
143 single dose strategy diminish even for an effective first dose, as an insufficient proportion of the  
144 population are immunized. The effect of the vaccine on case numbers is sensitive to when it



145 is introduced in the dynamical cycle (Figs. S7, S8), highlighting the critical interplay between  
146 the force of infection and the level of population immunity (see Supplementary Materials for  
147 further details).

Figure 3: Heat maps depicting various epidemiological outcomes contingent on dosing regimes. (A) Cumulative severe (left) and total (right) case numbers relative to the scenario with no vaccine from the time of vaccine introduction through the end of the five-year time period following the onset of the pandemic as a function of the one- to two-dose immune response ratio  $x_e$  and the inter-dose period. Parameters correspond to the “weak” immunity scenario of Figure 2A, but  $x_e$  sets the value of  $\epsilon_1$ ,  $\rho_1$ , and  $x_{\text{sev},1}$ . Specifically, we take  $\epsilon_1 = \epsilon_2 + (1 - x_e)(1 - \epsilon_2)$  such that the susceptibility to infection after a waned single dose interpolates linearly between the value after waned two doses ( $\epsilon_2$ ) when the one and two dose immune responses are equally strong ( $x_e = 1$ ) and unity (full susceptibility) when a single dose offers no immune protection ( $x_e = 0$ ). Similarly, we take  $x_{\text{sev},1} = x_{\text{sev},2} + (1 - x_e)(x_{\text{sev},V} - x_{\text{sev},2})$  such that the fraction of severe cases for infections following a waned single dose interpolates linearly between the value after waned two doses ( $x_{\text{sev},2}$ ) when  $x_e = 1$  and the value after a (failed) vaccination  $x_{\text{sev},V}$  when  $x_e = 0$ . Finally,  $\rho_1$  is given by  $\rho_1 = \rho_2/x_e$ . (B) Values of  $\nu_{\min}$ , the minimal rate of first dose administration per day such that for any  $\nu > \nu_{\min}$  the basic reproduction  $\mathcal{R}_0[\nu] < 1$  and the disease cannot invade (see Supplementary Materials), as a function of the strength of immunity following one ( $\epsilon_1$ ) and two ( $\epsilon_2$ ) waned vaccines doses, for different inter-dose periods. We take the duration of one dose and two dose vaccinal immunity to be  $1/\rho_1 = 0.5$  years and  $1/\rho_2 = 1$  year, respectively, and set  $\epsilon_{V_1} = 0.1$  and  $\epsilon_{V_2} = 0.05$ .

148 Vaccinal immunity will be central to efforts to attain community immunity and prevent local  
149 spread due to case importation. We therefore analytically calculated the first vaccine dose ad-  
150 ministration rate for a given inter-dose spacing required for community immunity in our model  
151 (see Supplementary Materials). In the long term, however, individuals whose one- or two-dose  
152 immunity has waned will likely be able to be vaccinated again before infection, and so we incor-  
153 porated re-vaccination of these individuals into the extended model and computed an analogous  
154 minimal vaccination rate which we plot in Figure 3B. We find that as the inter-dose period  
155 grows, this minimal rate depends increasingly on the degree of reduction in susceptibility after  
156 the waning of one-dose vaccinal immunity  $\epsilon_1$  (Figure 3B and see Figure S13 for other parameter  
157 choices). Vaccine refusal (17) may also impact the attainment of community immunity through

158 vaccinal immunity in the longer-term (see Supplementary Materials).

159

160 While we have assumed that the inter-dose period is exponentially distributed, we have re-  
161 laxated this assumption and examined an Erlang-distributed inter-dose period (see Supplementary  
162 Materials). The model predictions are qualitatively and quantitatively similar (compare Figure  
163 [2](#) with Fig. S9), justifying our choice of the simpler model.

164

## 165 **Evolutionary impacts**

166 The recent emergence of numerous SARS-CoV-2 variants in still relatively susceptible popu-  
167 lations underline the virus's evolutionary potential (*18–20*). We focus here on the longer term  
168 potential for immune escape from natural or vaccinal immunity (*13*). For immune escape vari-  
169 ants to spread within a population, they must first arise via mutation, and then there must be  
170 substantial selection pressure in their favour. We expect the greatest opportunity for variants to  
171 arise in (and spread from) hosts with the highest viral loads, likely those with the least immunity.  
172 On the other hand, we expect the greatest selection where immunity is the greatest. Previous  
173 research on the phylodynamic interaction between viral epidemiology and evolution (based on  
174 seasonal influenza) predicts that partially immune individuals (permitting intermediate levels  
175 of selection and transmission) could maximize levels of escape (*13*), Figure [4A](#)). This is con-  
176 sistent with case reports of sustained antigenic evolution in immunocompromised patients with  
177 prolonged Covid-19 infections (*21*). Under this model, we would project that different cate-  
178 gories of secondarily infected people (after waning of natural immunity or immunity conferred  
179 from one or two doses of vaccine) would be key potential contributors to viral immune escape.

180

181 In Figure [4](#), we explore three potential evolutionary scenarios, each with their own assump-

182 tions regarding viral abundance and within-host selection for the different immune classes. In  
183 all scenarios, we assume for simplicity that immunity elicited after two doses of the vaccine is  
184 equivalent to that elicited after natural infection. We also assume that transmission rises with vi-  
185 ral abundance in hosts (13). In Scenario I (black borders on circles, top panel of Figure 4A), we  
186 assume that infections of all classes of partially susceptible individuals lead to strong selective  
187 pressures and low viral abundance (a marker of low transmission), and thus low rates of adap-  
188 tation, with only slightly reduced immune pressure for infections after a waned single vaccine  
189 dose relative to natural infection or two doses. Scenario II (blue borders on circles, middle panel  
190 of Figure 4A), considers a situation where natural and two-dose vaccinal immunity again lead to  
191 low viral abundance, but one-dose vaccinal immunity is associated with intermediate immune  
192 pressure that results in substantially higher rates of viral adaptation. Finally, in Scenario III  
193 (purple borders on circles, bottom panel of Figure 4A), adaptive immune responses following  
194 waned natural, one dose, and two dose vaccinal immunity all lead to similar intermediate levels  
195 of immune pressure and high rates of viral adaptation. In all cases, we assume for tractability  
196 that viral immune escape is not correlated with clinical severity (22).

197

198 The relative potential viral adaptation rates (see (13) for more details) corresponding to each  
199 scenario are presented in the top rows of Figures 4B and 4C. This relative rate is estimated as  
200 the sum of the sizes of the infection classes following waned immunity (i.e.  $I_S$  after  $S_S$ ,  $I_{S_1}$   
201 after  $S_{S_1}$ , and  $I_{S_2}$  after  $S_{S_2}$ ) weighted by the infection class-specific net viral adaptation rate  
202 assigned in each scenario. Therefore, this quantity reflects a weight-averaged potential rate  
203 for viral adaptation per-individual per-infection. The corresponding immune and susceptibility  
204 classes are plotted in the middle and bottom rows, respectively, according to the colour scheme  
205 defined in Figure 1A. The weaker immunity scenario of Figure 2A is considered, with Figures  
206 4B and 4C corresponding, respectively, to the situations of a weaker and more robust single

207 vaccine dose relative to two doses. The leftmost column corresponds to a one dose strategy, an  
208 inter-dose period of  $\frac{1}{\omega} = 24$  weeks is assumed in the middle column, and the rightmost column  
209 assumes a two dose strategy with doses separated by the clinical trial window of  $\frac{1}{\omega} = 4$  weeks.

210

211 Different assumptions regarding the strength and duration of adaptive immune responses to  
212 vaccines and natural infections result in different predictions for the proportions of individuals  
213 in the partially susceptible immune classes over time. When one dose vaccinal immunity is  
214 poor, a one-dose strategy results in the rapid accumulation of partially susceptible  $S_{S_1}$  individu-  
215 als (Figure 4B, bottom row) and a greater infection burden. When the assumed individual rates  
216 of evolutionary adaptation arising from these infection classes are high (Scenarios II and III),  
217 we find that a one-dose strategy could lead to substantially higher relative rates of adaptation.  
218 This effect can be mitigated by implementing a two-dose strategy even with a longer inter-dose  
219 period than the recommended duration, echoing our epidemiological findings.

220

221 When one dose vaccinal immunity is strong, reduced infection burdens result in lower rel-  
222 ative rates of adaptation when a one dose strategy is used, although the large fraction of  $S_{S_1}$   
223 individuals may still lead to evolutionary pressure, particularly when the potential viral adapta-  
224 tion rate associated with  $I_{S_1}$  infections is large. A two-dose strategy mitigates this effect, but  
225 the corresponding reduction in vaccinated individuals increases the infection burden from other  
226 classes. Thus, to avoid these potentially pessimistic evolutionary outcomes, our results high-  
227 light the importance of rapid vaccine deployment. More broadly, our results further underline  
228 the importance of equitable, global vaccine deployment (23, 24): immune escape anywhere will  
229 quickly spread.

Figure 4: Potential viral evolution scenarios under different vaccine regimes. (A) Schematic representations of the potential net viral adaptation rate associated with the  $I_S$ ,  $I_{S_1}$ , and  $I_{S_2}$  infection classes under three different scenarios. These are illustrated by the filled dots, with the central colour denoting the infection class and corresponding to the legend in Figure 1A. The dot outlines correspond to the three scenarios considered (Scenario I: black lines and top panel, Scenario II: blue lines and middle panel, and Scenario III: purple lines and bottom panel). The phylodynamic model for potential viral adaptation as a function of immune pressure is adapted from (13). (B) and (C): relative net rates of adaptation (top rows; colours correspond to the scenarios in (A)), and composition of associated infection ( $I_S$ : solid lines,  $I_{S_1}$ : dashed lines,  $I_{S_2}$ : dashed-dotted lines; middle rows) and susceptible ( $S_S$ : solid lines,  $S_{S_1}$ : dashed lines,  $S_{S_2}$ : dashed-dotted lines; bottom rows) classes. The colours in the middle and bottom rows correspond to the legend in Figure 1A. The leftmost column corresponds to a one dose strategy, an inter-dose period of  $\frac{1}{\omega} = 24$  weeks is assumed in the middle column, and the rightmost column assumes a two dose strategy with doses separated by the recommended window of  $\frac{1}{\omega} = 4$  weeks. Both (B) and (C) correspond to a “weak” natural and vaccinal immunity scenario, with the same parameters as those in Figure 2A. A weaker immune response after one vaccine dose is assumed in (B) (with parameters corresponding to those in the top section of Figure 2A), and a stronger immune response after one vaccine dose is assumed in (C) (with parameters corresponding to those in the bottom section of Figure 2A). The weights used to calculate the relative net rates of adaptation are  $w_{I_S,I} = 0.05$ ,  $w_{I_{S_1},I} = 0.3$ , and  $w_{I_{S_2},I} = 0.05$  in Scenario I,  $w_{I_S,II} = 0.05$ ,  $w_{I_{S_1},II} = 1$ , and  $w_{I_{S_2},II} = 0.05$  in Scenario II, and  $w_{I_S,III} = 0.8$ ,  $w_{I_{S_1},III} = 1$ , and  $w_{I_{S_2},III} = 0.8$  in Scenario III.

## 230 Impact of increasing vaccination through time

231 In Supplementary Materials (Figures S10, S11, S12), we explore the implications of ramping  
232 up vaccine deployment through two approaches. First, we examine a simple increase in the  
233 rate of administration of the first dose and unchanged dosing regimes (Fig. S10). Qualitatively,  
234 these results are largely analogous to our previous results, and reflect the benefits of increasing  
235 population immunity through an increase in vaccination deployment.

236

237 However, as vaccines become more widely available, policies on dosing regimes may change.

238 The second approach we consider is a timely shift to a two-dose policy with recommended inter-  
239 dose spacing as vaccine deployment capacity increases (Figs. S11, S12). Initially delaying (or

240 omitting) the second dose decreases the first epidemic peak after the initiation of vaccination.  
241 Such a reduction in first peak size would also reduce secondary infections, and thus potentially  
242 immune escape in most cases (i.e. an evolutionary advantage). Subsequently, the switch to a  
243 manufacturer-timed vaccine dosage regime mitigates the potential medium-term disadvantages  
244 of delaying (or omitting) the second dose that may arise if immunity conferred from a sin-  
245 gle dose is relatively poor, including the accumulation of partially susceptible  $S_{S_1}$  individuals  
246 whose one-dose vaccinal immunity has waned. These contrasts highlight the importance of  
247 data-driven policies that undergo constant re-evaluation as vaccination progresses.

## 248 **Caveats**

249 Our immuno-epidemiological model makes several assumptions. While heterogeneities (super-  
250 spreading, age, space, etc) (25–27) are important for the quantitative prediction of SARS-CoV-2  
251 dynamics, we previously found that these do not qualitatively affect our results (8). Never-  
252 theless, we again briefly explore heterogeneities in transmission and vaccine coverage in the  
253 Supplementary Materials. We have also assumed that the robustness of immune responses fol-  
254 lowing the second dose is independent of the inter-dose period, yet it is possible that delaying  
255 the second dose may actually enhance adaptive immune responses. Detailed clinical evaluation  
256 of adaptive immune responses after one and two vaccine doses with different inter-dose spacing  
257 is an important direction for future work.

258  
259 Additionally, we have assumed highly simplified scenarios for NPIs. The chosen scenario  
260 was selected to qualitatively capture current estimates of SARS-CoV-2 prevalence and seropos-  
261 itivity in large cities. However, these values vary substantially between locations, a notable  
262 example being recent estimates of a large infection rate in Manaus, Brazil during the first  
263 wave (28), or countries having almost no infections due to the successful implementation of

264 NPIs (29–31). We have examined these scenarios in the Supplementary Materials (Figures S5  
265 and S6). The qualitative projections of our model are sensitive to the composition of infection  
266 and immune classes at the onset of vaccination (including, therefore, the assumption of dramat-  
267 ically higher seropositivity levels, i.e. the sum of the  $S_S$  and  $R$  classes). We further explore this  
268 in the Supplementary Materials through the initiation of vaccination at different times in the dy-  
269 namic cycle (Figs. S7 and S8). Thorough explorations of various NPIs, seasonal transmission  
270 rate patterns, vaccine deployment rates, dosing regimes, and clinical burdens will be able to be  
271 investigated for broad ranges of epidemiological and immunological parameters with an online  
272 interactive application upon publication.

273

274 Finally, we have explored the simplest evolutionary model, which can only give a general  
275 indication of the potential for evolution under different scenarios. Including more complex  
276 evolutionary models (32, 33) into our framework is thus another important area for future work.  
277 A full list of caveats is presented in Supplementary Materials.

## 278 **Conclusion**

279 The deployment of SARS-CoV-2 vaccines in the coming months will strongly shape post-  
280 pandemic epidemiological trajectories and characteristics of accumulated population immunity.  
281 Dosing regimes should seek to navigate existing immunological and epidemiological trade-  
282 offs between individuals and populations. Using simple models, we have shown that different  
283 regimes may have crucial epidemiological and evolutionary impacts, resulting in a wide range  
284 of potential outcomes in the medium term. Our work also lays the foundation for a number  
285 of future considerations related to vaccine deployment during ongoing epidemics, especially  
286 preparing against future pandemics.

287

288 In line with intuition, spreading single doses in emergency settings (i.e. rising infections)  
289 is beneficial in the short term and reduces prevalence. Furthermore, we find that if immunity  
290 following a single dose is robust, then delaying the second dose is also optimal from an epi-  
291 demiological perspective in the longer term. On the other hand, if one-dose vaccinal immunity  
292 is weak, the outcome could be more pessimistic; specifically, a vaccine strategy with a very  
293 long inter-dose period could lead to marginal short-term benefits (a decrease in the short-term  
294 burden) at the cost of a higher infection burden in the long term and substantially more poten-  
295 tial for viral evolution. These negative longer term effects may be alleviated by the eventual  
296 administration of a second dose, even if it is moderately delayed. With additional knowledge of  
297 the relative strength and duration of one-dose vaccinal immunity and corresponding, clinically-  
298 informed policies related to dosing regimes, pessimistic scenarios may be avoided.

299

300 In places where vaccine deployment is delayed and vaccination rates are low, our results  
301 stress the subsequent negative epidemiological and evolutionary impacts that may emerge.  
302 Particularly since these consequences (e.g., the evolution of new variants) could emerge as  
303 global problems, there is an urgent need for global equity in vaccine distribution and deploy-  
304 ment (23, 24).

305

306 Current uncertainties surrounding the strength and duration of adaptive immunity in re-  
307 sponse to natural infection or vaccination lead to very broad ranges for the possible outcomes  
308 of various dosing regimes. Nevertheless, ongoing elevated Covid-19 case numbers stresses the  
309 rapid need for effective, mass vaccine deployment. Overall, our work emphasizes that the im-  
310 pact of vaccine dosing regimes are strongly dependent on the relative robustness of immunity  
311 conferred by a single dose. It is therefore imperative to determine the strength and duration  
312 of clinical protection and transmission-blocking immunity through careful clinical evaluations



313 (including, for instance, randomized control trials of dose intervals and regular testing of viral  
314 loads in vaccinated individuals, their contacts, and those who have recovered from natural in-  
315 fections) in order to enforce sound public policies. Our results underscore the importance of  
316 exploring the phylodynamic interaction of pathogen dynamics and evolution, from within host  
317 to global scales, for SARS-CoV-2, influenza, and other important pathogens (32–37).

## 318 **Supplementary Materials**

319 The Supplementary Materials contain technical details, expanded analyses, supplementary fig-  
320 ures, and references. See attached document.

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- 348 14. We qualitatively describe the fraction of population that have received vaccines by as-  
349 suming that vaccination occurs at random in the population. Specifically, suppose vacci-  
350 nation occurs at time  $\tau = 0$ . The rate of change of the fraction not vaccinated follows  
351  $\frac{dN}{d\tau} = -\nu N$ ,  $N(0) = 1$ , giving  $N(\tau) = e^{-\nu\tau}$  so that the fraction vaccinated with one dose  
352  $Y(\tau) = 1 - N(\tau) = 1 - e^{-\nu\tau}$ . The fractions of those vaccinated with one but not two doses

353 follows  $\frac{dW_1}{d\tau} = \nu N - \omega W_1$ , giving  $W_1(\tau) = \frac{\nu(e^{-\nu\tau} - e^{-\omega\tau})}{\omega - \nu}$ . Then, the fraction vaccinated  
354 with two doses is  $W_2(\tau) = Y(\tau) - W_1(\tau) = 1 - \frac{\omega}{\omega - \nu}e^{-\nu\tau} + \frac{\nu}{\omega - \nu}e^{-\omega\tau}$ .

355 15. We calculate the total number of cases at any time point as  $I_T = I_P + I_S + I_V + I_{S_1} +$   
356  $I_{S_2}$ . Similarly, the number of severe cases is given by  $I_{T,sev} = x_{sev,p}I_P + x_{sev,s}I_S +$   
357  $x_{sev,v}I_V + x_{sev,1}I_{S_1} + x_{sev,2}I_{S_2}$ . Cumulative case numbers for a give time period are  
358 calculated through  $\gamma \sum I_T$  (total cases) and  $\gamma \sum I_{T,sev}$  (severe cases), where the summation  
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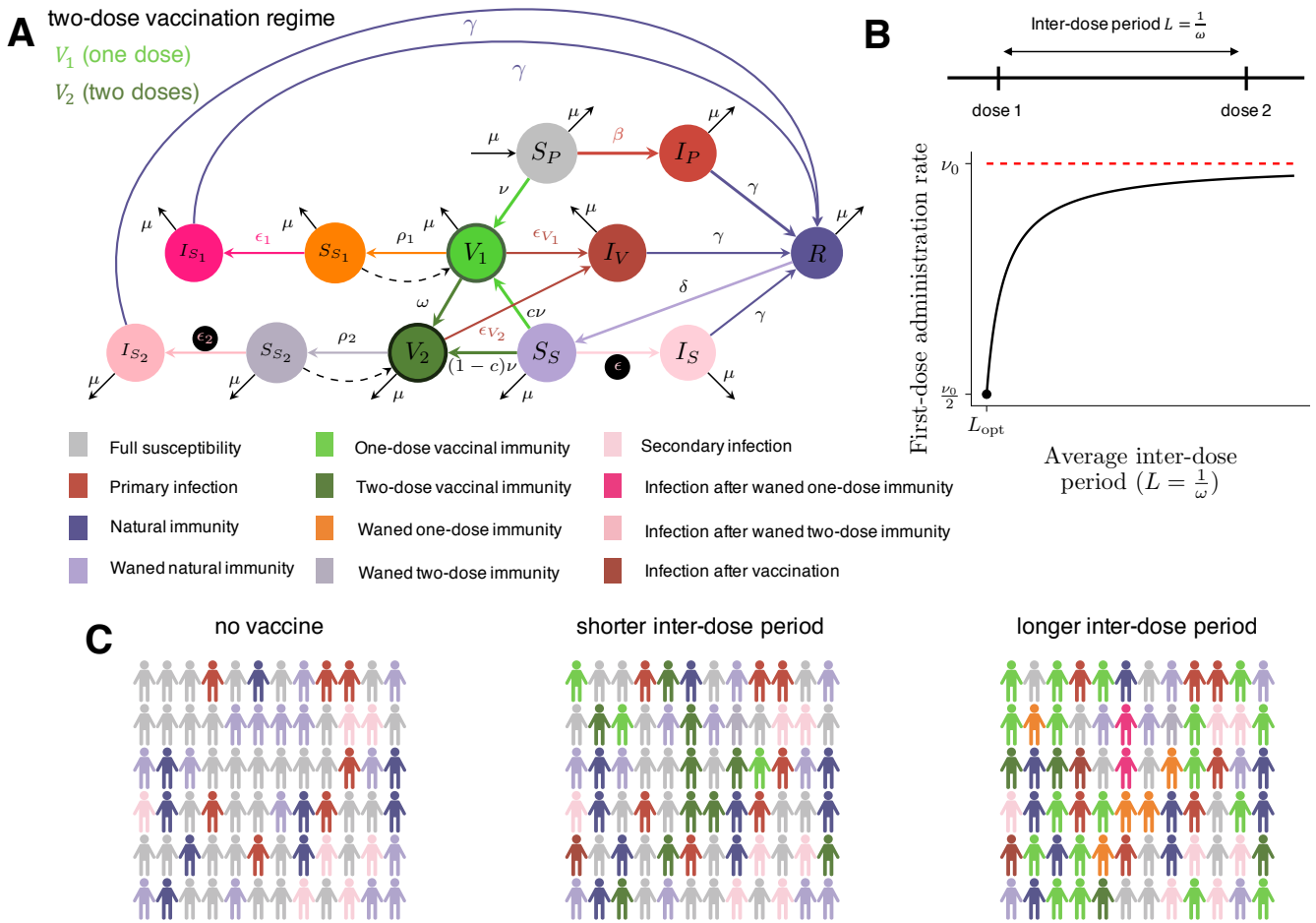
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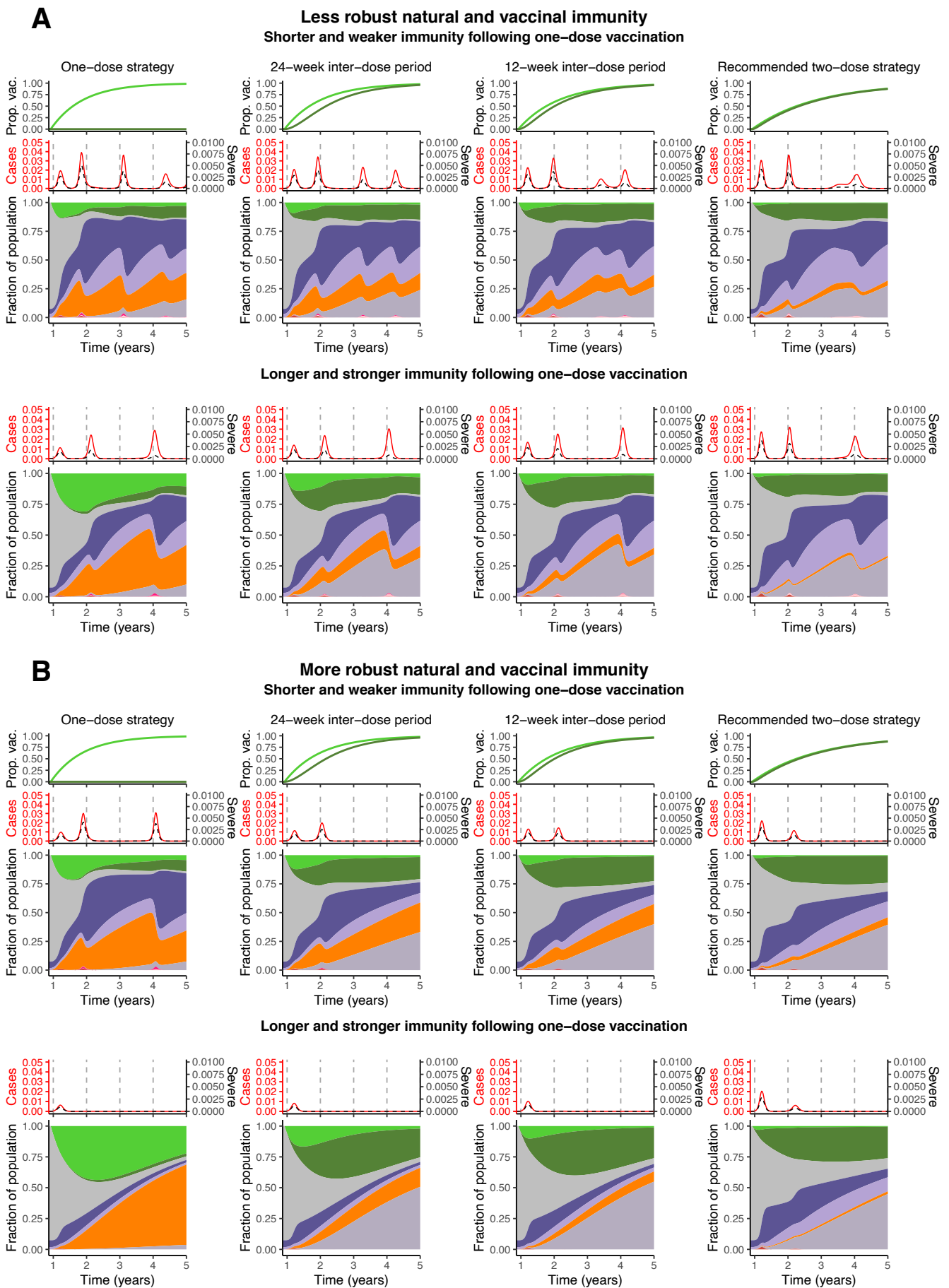
# Figure 1

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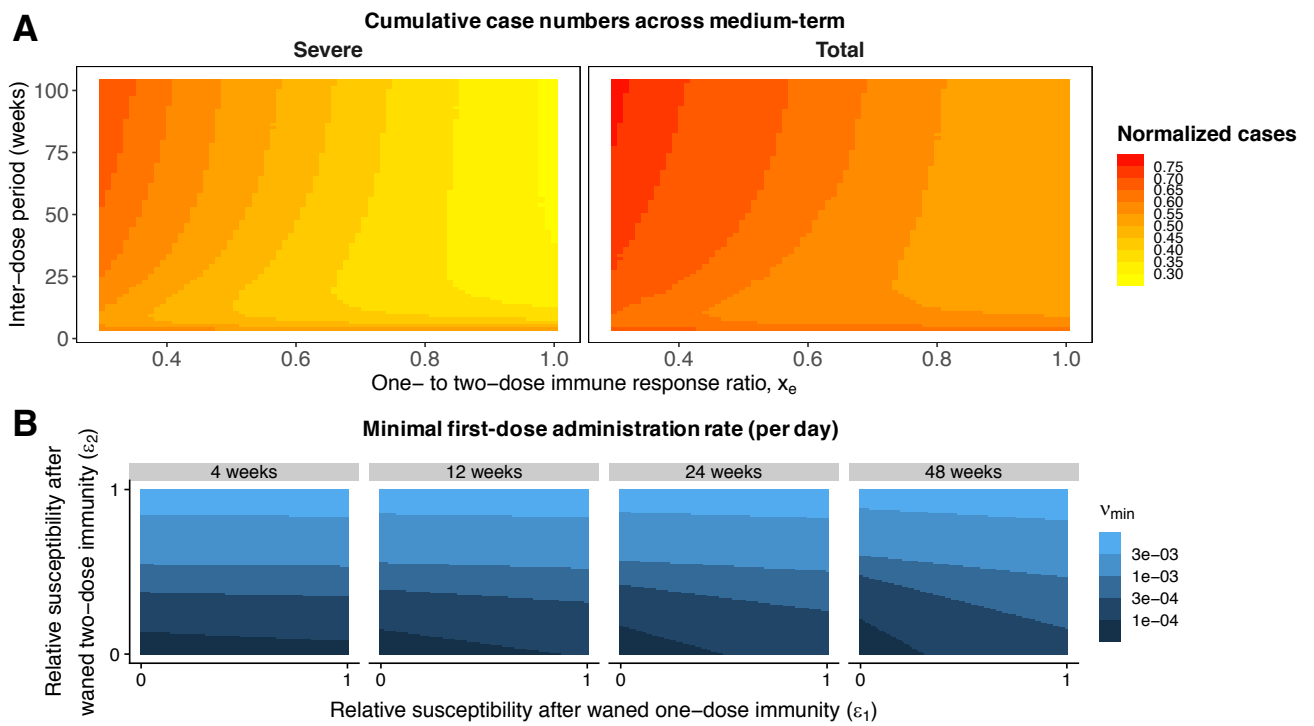
# Figure 2

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# Figure 3

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# Figure 4

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