

# Public Health Impact of Paxlovid as Treatment for COVID-19, United States

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We evaluated the population-level benefits of expanding treatment with the antiviral drug Paxlovid (nirmatrelvir/ritonavir) in the United States for SARS-CoV-2 Omicron variant infections. Using a multiscale mathematical model, we found that treating 20% of symptomatic case-patients with Paxlovid over a period of 300 days beginning in January 2022 resulted in life and cost savings. In a low-transmission scenario (effective reproduction number of 1.2), this approach could avert 0.28 million (95% CI 0.03–0.59 million) hospitalizations and save US \$56.95 billion (95% CI US \$2.62–\$122.63 billion). In a higher transmission scenario (effective reproduction number of 3), the benefits increase, potentially preventing 0.85 million (95% CI 0.36–1.38 million) hospitalizations and saving US \$170.17 billion (95% CI US \$60.49–\$286.14 billion). Our findings suggest that timely and widespread use of Paxlovid could be an effective and economical approach to mitigate the effects of COVID-19.

Antiviral drugs can substantially reduce illness and deaths from human infections. For example, antiretroviral therapy has prevented millions of HIV/AIDS deaths globally since the late 1980s (1). During the 2009 influenza A(H1N1) pandemic, oseltamivir was widely administered in the United States (28.4 prescriptions/1,000 persons) (2); rapid treatment after symptom onset reduced the risk for hospitalization by an estimated 63% (95% CI 17%–81%) (3). The reduction in viral load might reduce the risk for onward transmission while accelerating recovery. A counterfactual analysis suggests that treating even 10% of infected patients with baloxavir shortly

after symptom onset would have prevented millions of infections and thousands of deaths in the United States during the severe 2017–18 influenza season (4). A fast-acting SARS-CoV-2 antiviral could similarly be deployed to curtail transmission on a population scale and directly save lives (5).

Paxlovid (Pfizer, <https://www.pfizer.com>), which received Food and Drug Administration Emergency Use Authorization on December 22, 2021, for treating SARS-CoV-2 infections in persons >12 years of age, combines 2 different antiviral agents, nirmatrelvir and ritonavir. Treating symptomatic COVID-19 patients with Paxlovid reduces hospitalization risks by an estimated 0.59 (95% CI 0.48–0.71) for adults 18–49 years of age, 0.40 (95% CI 0.34–0.48) for adults 50–64 years of age, and 0.53 (95% CI 0.48–0.58) for adults >64 years of age (6). Paxlovid has proven effective against the Omicron variant (7). In January 2022, the United States ordered 20 million courses of Paxlovid to be delivered within 9 months (8).

In this study, we analyzed the population-level benefits of expanding the clinical use of Paxlovid to treat COVID-19. By fitting a within-host model of viral replication to viral titer data from >2,000 COVID-19 patients, we provide early estimates for the efficacy of Paxlovid in curtailing viral load, depending on the timing of treatment after infection. Then, using a population-level SARS-CoV-2 transmission model, we estimated the effects of Paxlovid-based interventions on reducing the healthcare and economic burden of future COVID-19 epidemics. Specifically, we estimated the number of cases, hospitalizations,

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and deaths, as well as healthcare costs averted under a range of transmission scenarios, in which we vary both the between-individual transmission rate of the virus and the proportion of case-patients who receive rapid treatment with Paxlovid. This 2-level analytic framework can broadly support the rapid evaluation of antiviral-based mitigation strategies against COVID-19 and other respiratory viruses (4).

## Materials and Methods

### Within-Host Model of SARS-CoV-2 Replication Dynamics

We simulated SARS-CoV-2 virus kinetics in an infected person and the effect of Paxlovid treatment on viral growth using a standard target-cell limited virus kinetic model that tracks the number of uninfected cells, infected cells, and free viral particles (9,10) (Appendix, <https://wwwnc.cdc.gov/EID/article/30/2/23-0835-App1.pdf>). We used individual patient viral load data from a Paxlovid clinical trial data to estimate the 5 key parameters of the model: the infection rate of susceptible cells ( $b$ ), the rate at which infected cells die ( $\delta$ ), the rate at which active viruses were cleared ( $c$ ), the virus production rate ( $p$ ), and the efficacy of Paxlovid at suppressing viral replication ( $\epsilon$ ). Specifically, we used a stochastic approximation expectation-maximization algorithm to fit the model to 14-day viral titer data from 1,126 infected adults treated with a placebo and 1,120 infected adults treated with Paxlovid during a clinical trial in late 2021 (11) (Appendix).

### Modeling the Infectiousness of Treated and Untreated Cases

On the basis of previous studies (12,13), we assumed that a person's infectiousness is logarithmically related to their viral titer (Appendix). In this transmission model, we assumed that the daily infectiousness of a case-patient depends on whether they received treatment and, if so, the time at which treatment was initiated after symptom onset. To estimate the daily infectiousness of a given untreated or treated case-patient, we first used the within-host model to simulate the viral load on each day of the infection and set the viral load to zero when the estimated value dropped below the detection threshold of 100 (14). We then used a logarithmic equation (Appendix) to estimate the corresponding daily infectiousness.

### Modeling Population-Level SARS-CoV-2 Transmission Dynamics and Effects of Antiviral Treatment

We developed a stochastic individual-based network model of SARS-CoV-2 transmission dynamics in which susceptible persons can be infected by infected contacts

(Appendix Figure 1). The underlying contact network included 9,961 persons living in 5,000 households with sociodemographic characteristics provided in the 2017 National Household Travel Survey (15,16) (Appendix).

At every time point, each person was in one of 11 possible states: unvaccinated susceptible ( $S_U$ ), vaccinated susceptible ( $S_V$ ), exposed (E), presymptomatic (P), symptomatic infectious before becoming eligible for Paxlovid treatment (Y), symptomatic treated ( $Y_T$ ), symptomatic untreated ( $Y_U$ ), asymptomatic infectious (A), recovered (R), hospitalized (H), or deceased (D). We assumed that hospitalized patients were isolated and not able to infect others. Upon infection, a susceptible person progresses to the exposed state and then to either the presymptomatic state (probability  $\psi$ ) or asymptomatic state (probability  $1 - \psi$ ). Asymptomatic case-patients recover without experiencing symptoms or seeking treatment. Presymptomatic case-patients progress to the symptomatic state at a rate  $\omega$ , where they might be hospitalized according to published age-specific infection hospitalization rates ( $h_a$ ) and eventually recover or die from the infection, according to age-specific infection fatality rates ( $\mu_a$ ). A fraction  $\rho$  of symptomatic case-patients receive Paxlovid, initiated an average of 3 days after symptom onset, which is assumed to reduce the risk for hospitalization ( $\phi_a$ ), as well as the infectiousness of the person. The infectiousness of a case-patient depends on the timing of Paxlovid administration after infection, according to the daily infectiousness curves described in the previous section. Vaccinated persons initially have vaccine-derived immunity against infection  $\omega_{B'}$ , symptomatic disease  $\psi_{B'}$ , and death  $\theta_{B'}$ , which wanes gradually after vaccination. Similarly, recovered persons initially have infection-derived immunity against reinfection  $\omega_{N'}$ , symptomatic disease  $\psi_{N'}$ , and death  $\theta_{N'}$ , which wanes more slowly than vaccine-derived immunity. Persons who are vaccinated and previously infected are assumed to have the higher of the 2 levels of immunity (i.e., infection-acquired vs. immune-acquired) (Table 1; Appendix Tables 1, 2).

### Antiviral Treatment and Transmission Scenarios

We analyzed 24 different scenarios, each with an effective reproduction number ( $R_e$ ) (1.2, 1.5, 1.7, 2, 3, or 5) and Paxlovid treatment rate (20%, 50%, 80%, or 100%). For each scenario ( $s$ ), we compared 4 variations of the antiviral strategy: no treatment (i.e., treatment rate set to zero); treatment with Paxlovid at the given treatment rate; treatment with a hypothetical antiviral that reduces infectiousness with the same efficacy as Paxlovid but does not reduce severity; and treatment with a hypothetical antiviral that reduces

**Table 1.** Between-host parameter estimates used in study of public health impact of Paxlovid in treatment of COVID-19, United States\*

Key parameter	Estimated value
Symptomatic proportion, % ( $\psi$ )	75
Transition rate out of exposed state ( $d^{-1}$ ) ( $\sigma$ )	1/3
Time lag between infection and recovery in days for asymptomatic patients ( $d^{-1}$ ) ( $\gamma_A$ )	1/9
Time lag between symptom onset and recovery in days for symptomatic patients ( $d^{-1}$ ) ( $\gamma_T$ )	1/4
Transition rate from the presymptomatic to the symptomatic stage ( $d^{-1}$ ) ( $\omega$ )	1/2
Age-specific efficacy of Paxlovid in reducing the hospitalization rate, $\gamma$ ( $\phi_a$ )	
0–4	0.59 (95% CI 0.48–0.71)
5–17	0.59 (95% CI 0.48–0.71)
18–49	0.59 (95% CI 0.48–0.71)
50–64	0.40 (95% CI 0.34–0.48)
>65	0.53 (95% CI 0.48–0.58)
Life expectancy, $y$ , for age group $a$ , adjusted assuming a 3% yearly discount rate ( $\lambda_a$ )	
0–4	30.3
5–17	29.3
18–49	25.8
50–64	1837
>65	12.9

\*We use the between-host model to project population-level impacts of Paxlovid treatment. Key parameter values used in the model are listed below, with more details in Appendix Table 1 (<https://wwwnc.cdc.gov/EID/article/30/2/23-0835-App1.pdf>).

severity with the same efficacy as Paxlovid but does not reduce infectiousness. The last 2 variations enabled us to separate the direct therapeutic benefits from the indirect transmission-blocking benefits of Paxlovid. To estimate the health and economic costs associated with each scenario, we ran 100 stochastic simulations of each of the 4 strategy variations and calculated the mean and 95% CI across simulations of the years of life lost (YLL) averted and monetary costs attributable to Paxlovid treatment.

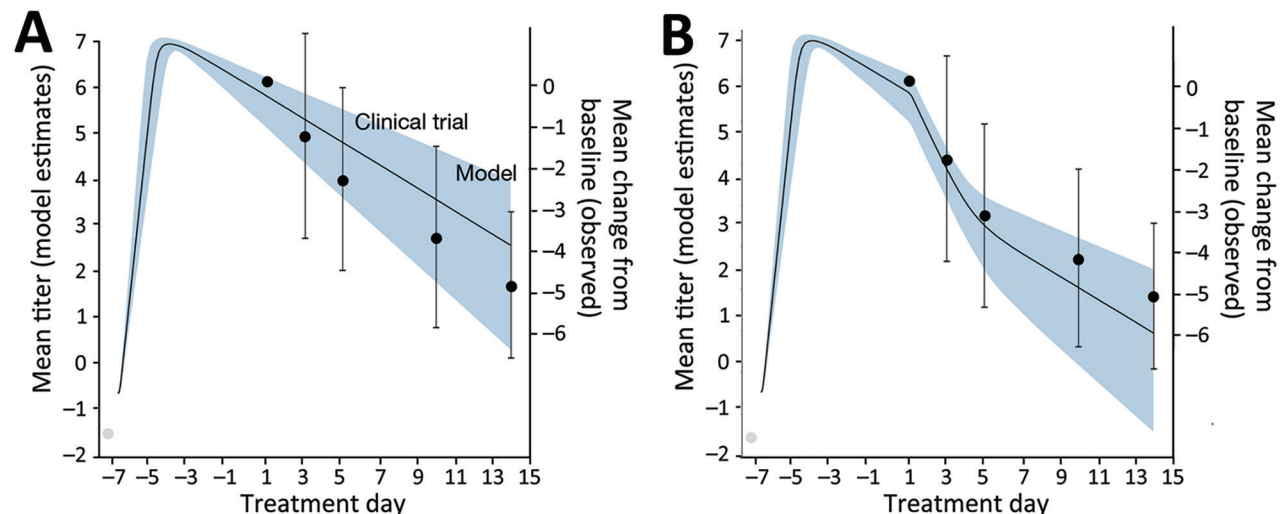
### Estimating YLL Averted and Monetary Costs

For each set of stochastic simulations, we estimated YLL averted for each antiviral strategy  $\tau$  by

comparing it to the no treatment strategy (Appendix). The willingness to pay per YLL averted is the maximum price a society is willing to pay to prevent the loss of 1 year of life. Health economists have inferred from healthcare expenditure that the United States is willing to pay US \$100,000 per quality-adjusted life-year (17), of which YLL is 1 component. For a given willingness to pay for a YLL averted ( $\theta$ ), we calculated the net monetary benefit (NMB) of each strategy (Appendix).

### Sensitivity Analyses and Model Validation

We assessed the robustness of the results with respect to the relationship between infectiousness and viral



**Figure 1.** Estimated and observed viral load following treatment with placebo (A) or Paxlovid (B) in large-scale campaign treating COVID-19, United States. The left y-axes, black lines, and blue shading indicate the means and 95% CI of SARS-CoV-2 viral load (RNA log<sub>10</sub> copies/mL) as estimated by the fitted within-host model. The right y-axes, black dots, and error bars indicate the means and 95% CI of the decrease in viral load since the initiation of treatment as reported in a clinical trial in which 1,126 patients received a placebo and 1,120 patients received Paxlovid during July 16–December 9, 2021 (11). Day one corresponds to the initiation of treatment. Gray circles denote the assumed initial viral load upon infection ( $V_0$ ) corresponding to 1 infectious virus particle in the upper respiratory tract (18).

**Table 2.** Within-host parameter estimates used in study of public health impact of Paxlovid in treatment of COVID-19, United States\*

Parameter	Mean (95% CI)
Cell infection rate in $10^{-6}$ mL/copies/day (b)	3.92 (2.82–5.38)
Infected cell death rate per day ( $\delta$ )	0.62 (0.42–0.92)
Virus production rate in copies/mL/day/cell (p)	3.19 (2.35–4.35)
Virus death rate per day (c)	2.21 (2.10–2.33)
Antiviral efficacy ( $\epsilon$ )	0.9937 (0.9917–0.9952)

\*We fit the within-host model to the mean viral load dynamics reported from a clinical trial involving 2,246 infected adults treated with either Paxlovid or a placebo (11) using nonlinear mixed-effects model method (19). This method allows between-subject variability to improve the precision and accuracy of estimates (20). Values are means and 95% CI of parameter values in population, assuming that antiviral efficacy follows logit-normal distribution, and all other individual parameters follow log-normal distributions.

load by investigating 3 alternative functions (i.e., sigmoid, log-proportional, and step) (Appendix Tables 5, 6). To validate our within-host viral replication mode, we compared model-estimated mean viral load trajectories for untreated and treated case-patients to corresponding clinical trial data for patients receiving placebo or Paxlovid treatment (1). We found that the observed mean decreases in viral load fall within the estimated 95% CI and vice versa (Figure 1; Appendix Figure 3).

To validate our transmission dynamic model, we compared model projections to observed incidence data during the early 2022 and late 2022 Omicron waves in the United States (Appendix Figure 2). For each of these waves, we fitted the model to reported case data to estimate the initial  $R_t$  and then simulated

the expected reported infections, assuming a 25% case-reporting rate (7).

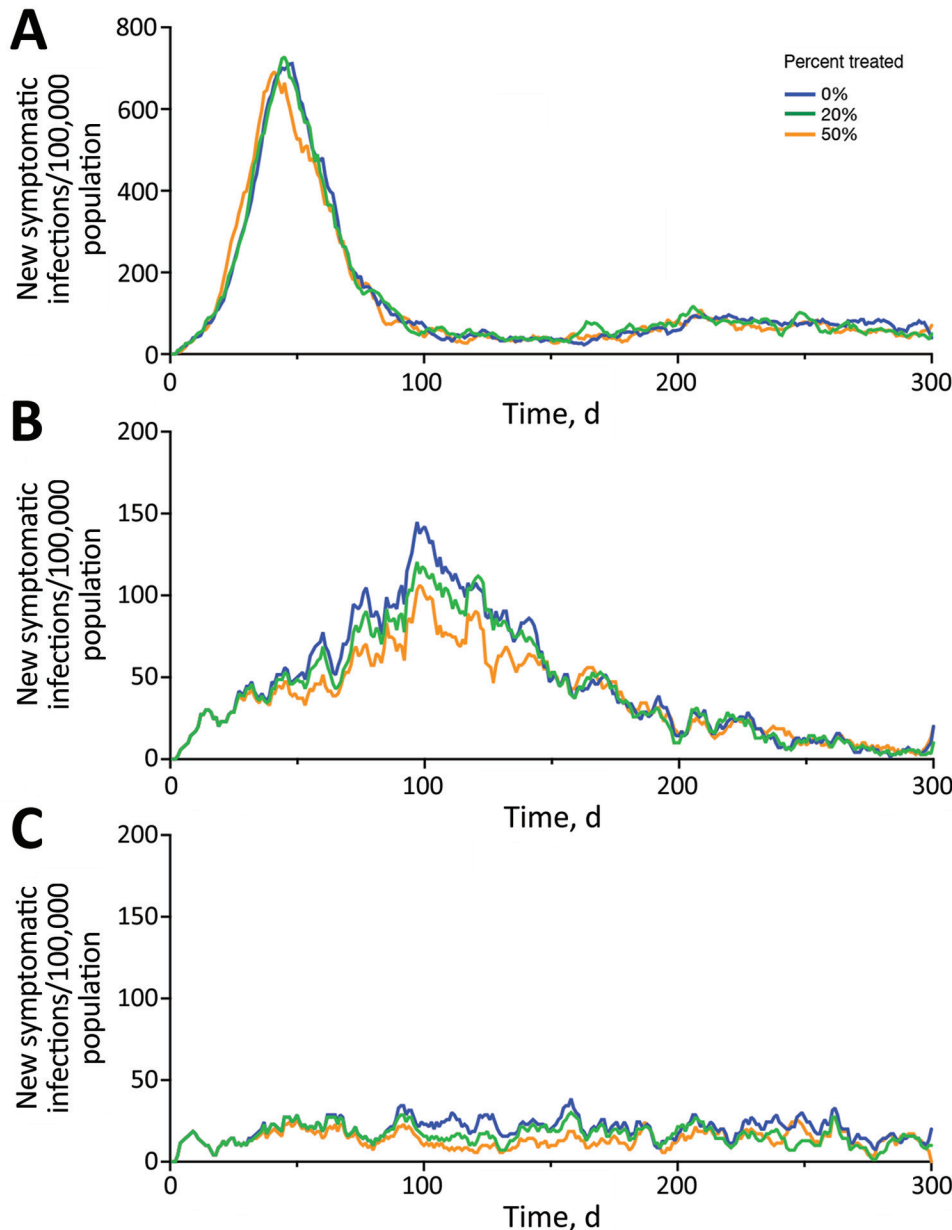
## Results

By fitting the within-host model to the mean viral load dynamics reported from a clinical trial (Table 2; Figure 1), we estimated that the rate at which viral particles infect susceptible cells (b) is 3.92 (95% CI 2.82–5.38)  $\times 10^{-6}$  mL/copies/day), the clearance rate for infected cells ( $\delta$ ) is 0.62 (95% CI 0.42–0.92) per day, the rate at which infected cells release virus (p) is 3.19 (95% CI 2.35–4.35) copies/mL/day/cell, and the rate at which free virus particles are cleared (c) is 2.21 (95% CI 2.10–2.33) per day. Treatment with Paxlovid is estimated to repress viral replication by 99.37% (95% CI 99.17%–99.52%) per day.

**Table 3.** Projected health and economic impacts of a large-scale SARS-CoV-2 Paxlovid campaign, United States

Outcome	$R_t$	Treatment rate, %	Mean (95% CI)
Infections averted, millions	1.2	20	10.54 (3.03–21.12)
		50	25.65 (12.59–41.19)
	1.7	20	4.25 (0.00–8.30)
		50	10.65 (5.77–16.70)
	3	20	0.67 (–0.13 to 1.45)
		50	1.68 (0.79–2.77)
Hospitalizations averted, millions	1.2	20	0.28 (0.03–0.59)
		50	0.67 (0.33–1.25)
	1.7	20	0.48 (0.07–0.92)
		50	1.16 (0.49–1.85)
	3	20	0.85 (0.36–1.38)
		50	2.08 (1.12–2.83)
Deaths averted, thousands	1.2	20	33.85 (1.69–71.15)
		50	79.11 (35.78–146.51)
	1.7	20	59.43 (9.13–129.86)
		50	145.44 (45.60–221.34)
	3	20	109.67 (35.95–179.83)
		50	266.69 (156.71–362.77)
NMB, USD billions	1.2	20	\$56.95 (\$2.62–\$122.63)
		50	\$135.60 (\$62.52–\$261.32)
	1.7	20	\$95.66 (\$8.54–\$196.23)
		50	\$232.35 (\$80.45–\$379.51)
	3	20	\$170.17 (\$60.49–\$286.14)
		50	\$417.18 (\$208.34–\$580.13)
Courses of treatment used, millions	1.2	20	5.77 (4.38–7.15)
		50	12.13 (8.86–14.89)
	1.7	20	13.57 (12.42–15.12)
		50	32.85 (30.87–34.76)
	3	20	24.41 (22.34–26.56)
		50	60.21 (57.07–63.16)

\*For each combination of treatment rate and reproduction number, the table provides the estimated mean and 95% CI of cases, hospitalizations, and deaths averted in the United States, NMB, and number of courses of treatment administered based on 100 pairs of stochastic simulations (treatment vs. no treatment simulations). NMB, net monetary benefit;  $R_t$ , effective reproduction number; USD, US dollars.



**Figure 2.** Projected symptomatic SARS-CoV-2 infections over 300 days in the United States across a range of transmission and Paxlovid treatment scenarios. Estimated incidence of symptomatic SARS-CoV-2 infections are shown assuming an effective reproduction number of 3.0 (A), 1.7 (B), or 1.2 (C). Colors correspond to 3 different treatment scenarios: 0% (blue), 20% (green), or 50% (orange) of symptomatic cases received a 5-day Paxlovid regimen initiated within 3 days of symptom onset.

We estimated the number of cases, hospitalizations, and deaths, as well as healthcare costs, averted under a range of transmission scenarios, in which we varied both the between-individual transmission rate of the virus and the proportion of case-patients who received rapid treatment with Paxlovid (Table 3; Figures 2, 3). Under a low-transmission scenario in which the  $R_t$  of the virus is 1.2, we estimated that treating 20% of symptomatic cases with Paxlovid would avert 10.54 million (95% CI 3.03–21.12 million) cases, 280,000 (95% CI 30,000–590,000) hospitalizations, and 33,850 (95% CI 1,690–71,150) deaths in the United States over a 300-day period (Appendix Table 4). Assuming a cost of US \$530 per course of

treatment (22) and willingness to pay per YLL averted of US \$100,000, we estimated that the optimal strategy is always the highest achievable treatment rate. A 20% treatment rate would be expected to yield an NMB of US \$56.95 billion (95% CI \$2.62–\$122.63 billion) averted.

To separate the direct (therapeutic) benefits of Paxlovid treatment from its indirect (transmission-reducing) effects, we conducted 2 additional analyses, 1 assuming the drug reduces severity but not infectivity and another assuming the opposite (Appendix Table 4). Assuming an  $R_t$  of 1.2, we estimated that direct therapeutic effects of treating 20% of symptomatic cases with Paxlovid would not affect the overall

attack rate but would avert 140,000 (95% CI -130,000 to 400,000) hospitalizations and 16,470 (95% CI -19,470 to 48,110) deaths over a 300-day period, resulting in an NMB of US \$25.35 (95% CI -\$34.98 to \$84.22) billion. The reduced infectivity of the treated cases would be expected to avert an additional 10.57 (95% CI 3.03–21.19) million infections, 160,000 (95% CI -130,000 to 530,000) hospitalizations, and 19,460 (95% CI -14,140 to 58,520) deaths, resulting in an NMB of US \$31.17 (95% CI -\$32.77 to \$103.74) billion.

## Discussion

Our results show that the widespread administration of Paxlovid would not only improve outcomes in treated patients but also concomitantly reduce risks of onward transmission. In this population-level assessment of expanding rapid treatment of symptomatic COVID-19 infections with Paxlovid, we found that the direct (therapeutic) effects of treatment would substantially reduce both deaths and socioeconomic costs. Of note, the indirect (transmission-blocking) effects would be expected to reduce burden by just as much, as well as substantially reducing the overall attack rate (Appendix Table 4). We would expect mass treatment campaigns to have even greater health and economic effects in countries that have adopted zero-COVID strategies and thus have lower levels of population-level immunity than the United States (23).

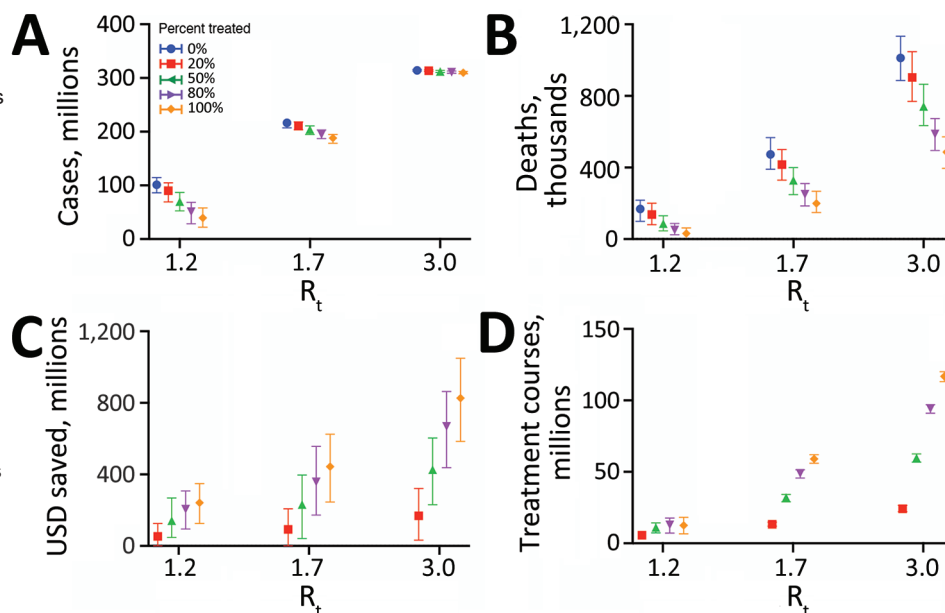
Drugs like Paxlovid could profoundly reduce the severity of COVID-19 and enable a global transition

to manageable coexistence with the virus. However, providing equitable and effective global access to SARS-CoV-2 antiviral drugs would require both ample supplies and broad-reaching test-and-treat programs. The pharmaceutical industry and global health agencies are working to produce enough Paxlovid to treat a large fraction of symptomatic cases (8). Online healthcare services (e.g., telemedicine) and community test-to-treat programs (24), such as those piloted in Pennsylvania and New Jersey (25), could be expanded nationally, and even globally, to accelerate and broaden access to antiviral drugs (26). For example, in 2020, China began an initiative to expand remote internet-based COVID-19 care (27). The country established 1,500 internet hospitals (either by extending existing hospitals or by opening new institutions) during 2019–2021 (28). The new services included follow-up consultations for common ailments (29) and served >239 million patients during December 2020–June 2021 (30). In addition, avoiding testing and treating infected individuals in person reduces the risk for SARS-CoV-2 transmission by patients to healthcare providers.

We highlight 3 limitations of our analyses that could be addressed as additional epidemiologic and clinical trial data become available. First, our fitted within-host model slightly overestimated viral levels for patients treated with placebo and underestimated those for patients receiving Paxlovid. The discrepancies might stem from limitations in the model structure or from unmodeled variation in

**Figure 3.** Projected health and economic impacts of a large-scale campaign using Paxlovid to treat COVID-19 over 300 days in the United States, across a range of transmission and treatment scenarios. Points and error bars correspond to means and 95% CI in number of infections in millions (A), number of deaths in millions (B), net monetary benefit in billions USD assuming a treatment course cost of US \$530 and willingness to pay per year of life lost averted of US \$100,000 (C), and number of courses of Paxlovid administered in millions (D). Each graph provides results for 3  $R_t$  and 5 different treatment scenarios: 0% (blue), 20% (red), 50% (green), 80% (purple), or 100% (orange) of symptomatic cases started a 5-day course

of Paxlovid within 3 days of symptom onset. Distributions are based on 100 stochastic simulations for each scenario. The results are scaled assuming a US population of 328.2 million (21).  $R_t$ , effective reproduction number; USD, US dollars.



viral kinetics and treatment efficacy across age or risk groups. In estimating model parameters, we considered only the mean in viral load of patients from 20 countries (4,31) (Figure 1). Incorporating such variability would enable us to analyze age-prioritized or risk-prioritized interventions and improve our estimates of the health and economic benefits of mass treatment. Second, we did not consider the emergence and spread of Paxlovid-resistant viruses, which could substantially undermine the utility of new drugs and exacerbate epidemics on a population level (32). Conversely, suppressed viral replication attributable to Paxlovid might limit viral evolution in treated patients. Depending on the immunological conditions of the individual person and population, reducing opportunities for viral growth and mutations could hinder the emergence of new variants (33). Third, we did not incorporate several economic, social, and logistical factors that might affect the expansion of Paxlovid treatment, including commercial impediments faced by the pharmaceutical companies that manufacture the drug (34); the costs of administering tests before treatment; and low levels of uptake stemming from misinformation, limited healthcare access, or pandemic fatigue. For example, in the 2009 H1N1 pandemic, only 40% of case-patients sought medical care within 3 days after symptom onset (35).

In conclusion, fast-acting antiviral drugs like Paxlovid can serve as invaluable tools to mitigate COVID-19 epidemics. By increasing supplies and improving infrastructure to enable rapid and equitable distribution, such drugs could substantially mitigate the health and societal burdens of COVID-19.

The computer code referenced in this study is available from Github (<https://github.com/ZhanweiDU/Pax>).

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comments. Y.B., Z.D., L.W., E.H.Y.L., I.C.H.F., P.H., B.J.C., A.P.G., R.M.K., and L.A.M. wrote the paper.

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# Public Health Impact of Paxlovid as Treatment for COVID-19, United States

## Appendix

### Within-Host Model of SARS-CoV-2 Replication Dynamics

The deterministic model given by

$$\frac{dU_i}{dt} = -bU_iV_i$$

$$\frac{dI_i}{dt} = bU_iV_i - \delta I_i$$

$$\frac{dV_i}{dt} = (1 - \epsilon)pI_i - cV_i$$

tracks the number of target cells at risk of infection ( $U_i$ ), infected cells ( $I_i$ ), and free viral particles ( $V_i$ ) ( $I$ ) (Appendix Figure 1). The rate at which free viral particles infect target cells is governed by the number of susceptible target cells, the number of free viral particles, and a fixed rate  $b$ . Viruses replicate at a rate  $p$  in infected cells; infected cells die at rate  $\delta$  and free viral particles die at rate  $c$ . The model assumes that Paxlovid inhibits the replication of viruses within infected cells, with efficacy  $\epsilon$ .

### Estimating the within-host model parameters

We fix the initial number of viruses ( $V_0$ ) at 1/30 copy/mL (corresponding to a single viral particle per 30 mL of nasal wash in the upper respiratory tract (2)) and the initial number of target cells ( $U_0$ ) at  $10^7$  ( $I$ ). For the Delta variant, the estimated average time from infection to symptom onset is 5 days (3); in recent clinical trials, the estimated average time from symptom

onset to initiation of treatment is 3 days (4). Based on these estimations, we assume that treatment is initiated 8 days after infection for treated patients.

To estimate the five model parameters governing the viral load dynamics (i.e., the infection rate of susceptible cells [ $b$ ], the rate at which infected cells die [ $\delta$ ], the rate at which active viruses were cleared [ $c$ ], the virus production rate [ $p$ ] and antiviral efficacy [ $\epsilon$ ]), we fit the within-host model to the mean SARS-Cov-2 RNA titer (log<sub>10</sub> copies/mL) at five time points (1, 3, 5, 10 and 14 days post initiation of treatment) measured across 1126 infected adults treated with a placebo during a clinical trial in late 2021 (4) and the mean SARS-Cov-2 RNA titer (log<sub>10</sub> copies/mL) at five time points (1, 3, 5, 10 and 14 days post initiation of treatment) measured across 1120 infected adults who received Paxlovid in the same clinical trial (5). We set the initial viral load upon infection,  $V_0$ , to correspond to one infectious virus particle in the upper respiratory tract (2). We assume that the average viral load at the initiation of treatment is  $10^6$  log<sub>10</sub> copies/mL (4). We use the Stochastic Approximation Expectation-Maximization (SAEM) algorithm to estimate the five parameters (MONOLIX 2021R1) (6,7) and confirm the convergence of estimates via trace plots.

## Modeling the daily infectiousness of treated and untreated cases

Our between-host SARS-CoV-2 transmission model assumes that the infectiousness of an infected individual depends on the number of days elapsed since they became infected ( $t$ ), whether or not they receive Paxlovid, and, if so, how quickly treatment is initiated. We use  $\iota$  to indicate treatment initiation time in days after symptom onset and  $\iota = \infty$  to denote that a case remains untreated. Specifically, we assume that an individual's infectiousness is *proportional to*:

$$\beta_\iota = \log(V_\iota(t))$$

Where  $V_\iota(t)$  represents the individual's viral load  $t$  days after infection. We use the fitted within-host model above to generate the  $V_\iota(t)$ , depending on whether and when the infected individual receives Paxlovid and assuming that infectiousness drops to zero when the viral load drops below the detection threshold of 100 (8).

## Between-host model of SARS-CoV-2 transmission and treatment with Paxlovid-like drug

Our between-host agent-based model assumes that an infected individual's daily infectiousness toward one of their contacts depends on: (i) the time elapsed since infection, (ii) whether and when Paxlovid treatment is initiated, and (iii) whether the case and contact live in the same household. We use the infectiousness equation above ( $\beta_i$ ) to account for the first two variables, and then solve for household and non-household scaling constants that yield target secondary infection rates. Specifically, we first calibrate the daily within-household transmission rates for untreated cases to match reported estimates for household secondary attack rates (which is constant across all scenarios). We then calibrate the daily non-household transmission rates for untreated cases so that the model produces a specified overall initial reproduction number (which depends on the scenario analyzed).

To calibrate the within-household transmission rate scaling constant ( $\xi_h$ ), we assume that a household secondary attack rate of 35% (9) and solve for the  $\xi_h$  that satisfies  $\xi_h \sum_{t=0}^{50} \beta_{\infty}(t) = 0.35$ . To calibrate the non-household transmission rate scaling constant ( $\xi_{nh}$ ), we set the target initial effective reproduction number ( $\bar{R}_e$ ) and then apply an interior point algorithm to find the value of  $\xi_{nh}$  that minimizes the mean square error between  $\bar{R}_e$  and the average initial  $R_e$  across 100 simulated epidemics. For each simulation, we estimate  $R_e$  by calculating the average number of secondary infections across a random sample consisting of 1% of individuals infected during the first 100 days of the simulation.

At the start of a simulation, we set the proportions of the population with infection-acquired and vaccine-acquired immunity to values estimated from data provided by the U.S. Centers for Disease Control and Prevention (Appendix Table 1). To estimate the number of previously vaccinated individuals and the date of their most recent dose, we simulated vaccination rates based on reported uptake in the U.S. from 2020 to 2022 (10). For each previously vaccinated individual, we randomly selected the date of their first dose ( $t_1$ ) based on the reported age-specific vaccine administration rates, starting on October 29, 2021 (11) for children between 5 and 11 years old, May 10, 2021 for children between 12 and 15 years old (12), and December 13, 2020 for all others. We then randomly determined whether and when an individual receives their second primary dose and first booster based on CDC-recommended

waiting periods and reported rates of uptake. Specifically, we assumed second doses are administered beginning 3 weeks after the first dose, and the window for boosters depends on the timing of the booster dose, with a minimum gap of 8 months for individuals receiving their booster dose before September 23, 2021 (13), 6 months between September 24, 2021 and January 3, 2022 (14), and 5 months after January 4, 2022 (15). We initialized immunity in our simulations using the dates of the last dose received for each vaccinated individual (Appendix Table 1) (16).

For the previously infected individuals, we estimated their times of recovery. Specifically, we collected the daily population proportion of confirmed cases in the USA from 2021 to 2022 from Our World In Data (10). For each individual infected previously at the start of the simulation, we estimated the date of the previous infection ( $t_{\text{infect}}$ ) by taking draws from the distribution of the daily population proportion of cases between January 29, 2021 to January 29, 2022. We considered the time of recovery as ( $t_{\text{infect}} + 9$ ), where 9 days is the average time lag between infection and recovery (17).

At the start of each simulation, we also assume that 1% of the unvaccinated susceptible and recovered populations are newly infected (exposed), which corresponds to  $\approx 0.6\%$  of the total population. We assumed age-stratified estimates for Paxlovid's efficacy at preventing hospitalizations (18) (Appendix Table 1) and incorporated uncertainty by sampling the Paxlovid efficacy parameters for each simulation from triangular distributions with mean, lower bound, and upper bound equal to the estimated mean, 95% CI lower bound, and 95% CI upper bound, respectively. To estimate therapeutic benefits of the drug via pairs of simulations, we enforced the same sequence of random numbers in each simulation.

## **Individual-based network construction**

The individual-based SARS-CoV-2 infection dynamic model assumes that the virus spreads through a fixed contact network consisting of 9961 individuals and 124,878 contacts between those individuals. We populated our network by first constructing 5000 households. The size and age composition of each household is based on a randomly sampled household from among the 129,697 households included in 2017 National Household Travel Survey (19). We assumed that households are fully connected (i.e., all nodes in the same household are linked by

edges). We constructed random links between individuals in different households based on reported age-specific contact rates in the U.S., stratified into age bins of 5–17, 18–49, 50–64, and over 65 years (20). Specifically, to determine the number of contacts a node in age group  $a_i$  has with nodes in age group  $a_j$ , we draw random deviates from Poisson distributions centered at the mean number of contacts between  $a_i$  and  $a_j$ . The resulting network includes 5000 households, 2019 nodes (people), and degrees (numbers of edges per node) that roughly follow a gamma distribution with shape and scale parameters of 3.69 and 3.41, respectively. We directly scaled our results to the 2019 U.S. population of 328 million (21).

### Estimating the Years of Life Lost (YLL) Averted and Monetary Costs

For each set of stochastic simulations, we estimated the years of life loss (YLL) averted for each antiviral strategy  $\tau$  as follows:

1. Calculate the difference in incidence by age group as  $\Delta_{a,\tau} = D_{a,0} - D_{a,\tau}$ , where  $D_{a,0}$  and  $D_{a,\tau}$  are total deaths in age group  $a$  produced by the no treatment and strategy  $\tau$  simulations, respectively.

2. Estimate the YLL prevented by the strategy  $\tau$  as  $B_\tau = \sum_a (\lambda_a - a) \Delta_{a,\tau}$  where  $\lambda_a$  denotes the future-discounted life expectancy for individuals of age  $a$ .

Similarly, we determined the incremental monetary costs for each strategy  $\tau$  as given by

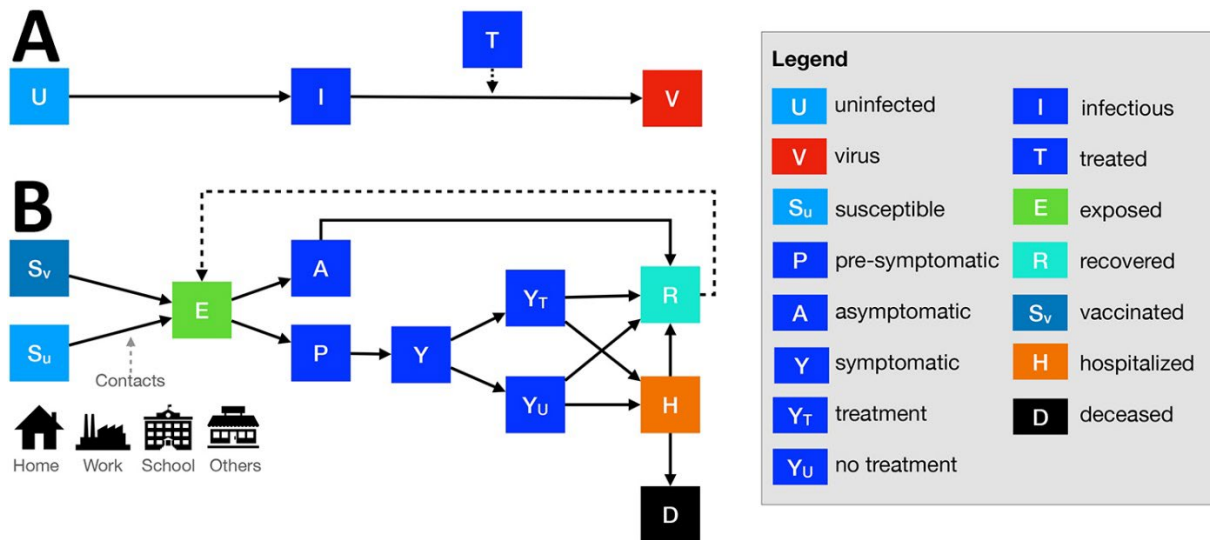
$$C_\tau = (T_\tau - T_0)c_T + \sum_a c_{H,a}(H_{\tau,a} - H_{0,a})$$

where  $T_0$  and  $T_\tau$  are the total number of treatment courses administered in the no treatment and strategy  $\tau$  simulations, respectively,  $c_T$  is the price of administering one course of antivirals,  $H_{\tau,a}$  and  $H_{0,a}$  are the total number of hospitalizations in age group  $a$  in each simulation, and  $c_{H,a}$  is the median COVID-19 hospitalization cost for age group  $a$ . The cost parameter values are given in Appendix Table 3.

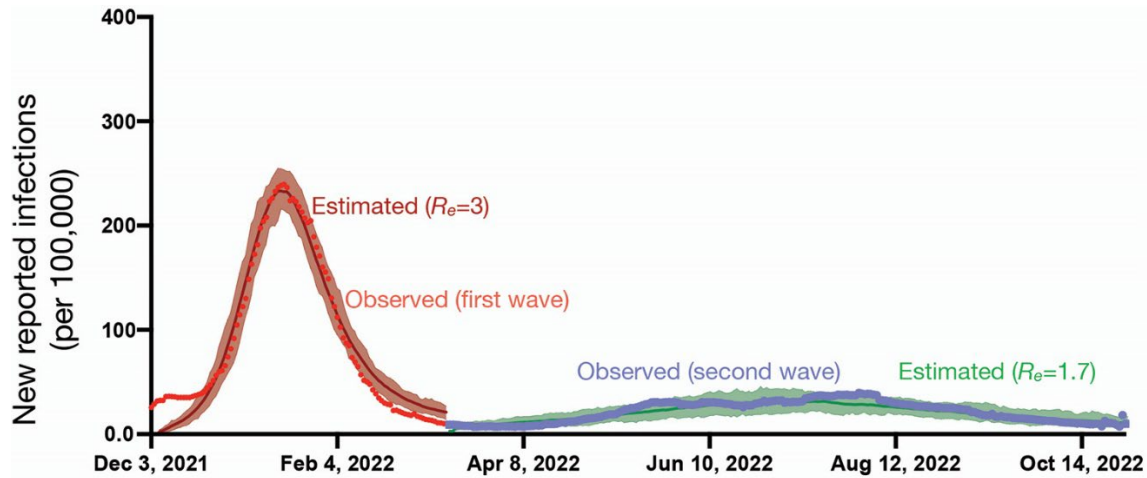
For a given willingness to pay for a YLL averted ( $\theta$ ), we calculated the net monetary benefit (NMB) of a strategy as

$$NMB_\tau = \theta \cdot B_\tau - C_\tau.$$

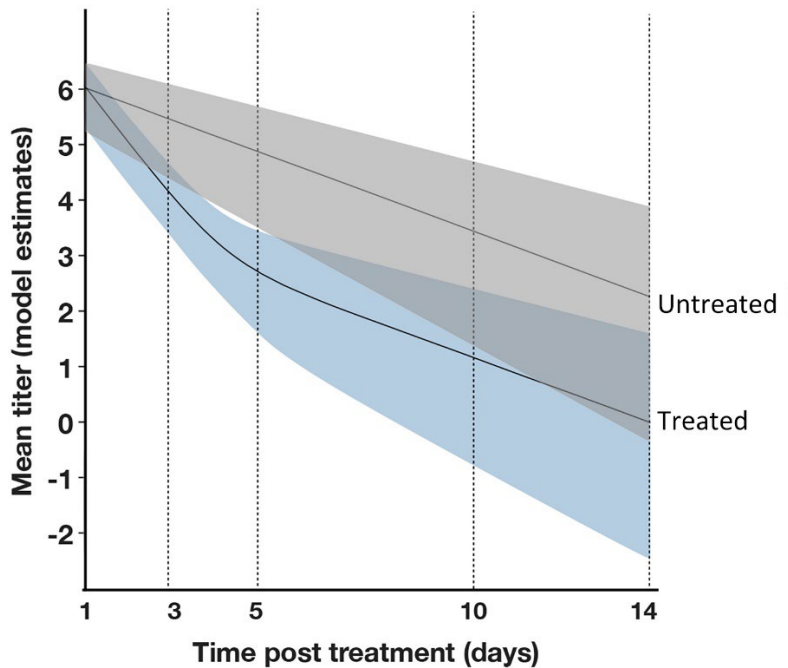
We determined the optimal strategy across a range of scenarios, each defined by the effective reproduction number ( $R_e$ ), willingness to pay, and cost of a vaccine.



**Appendix Figure 1.** Diagrams of the within-host and between-host models. (A) To estimate changes in infectivity following treatment with Paxlovid (T), we used a model that tracks the changing number of uninfected cells (U), infectious cells (I) and free viral particles (V) in an infected case, with and without treatment. (B) We projected population-level impacts of Paxlovid treatment using a stochastic individual-based model of SARS-CoV-2 transmission that considers age-specific risks and contact patterns in households, schools, workplaces, and other venues. Upon infection, susceptible ( $S_u$ ) and vaccinated ( $S_v$ ) individuals progress to exposed (E), asymptomatic infectious (A) or presymptomatic (P) and then to either symptomatic infectious (Y) with ( $Y_T$ ) or without ( $Y_U$ ) Paxlovid treatment. A fraction of symptomatic cases with or without treatment will be recovered (R) or hospitalized (H), and a subset of those will die (D). All asymptomatic cases eventually progress to a recovered class (R), where they remain protected from future infection. As immunity wanes, recovered individuals return to exposed (E) by reinfection.



**Appendix Figure 2.** Comparison of observed (22) to estimated SARS-CoV-2 incidence during two Omicron waves in the U.S.. Model estimates assume an initial effective reproduction number of 3.0 for the first large Omicron wave in the U.S. (December 3, 2021 to March 12, 2022) and a reproduction number of 1.7 for the second smaller wave (March 13, 2021 to October 31, 2022), and assume a case reporting rate of 25% (23). Points indicate reported incidence data; lines and shading correspond to the mean and 95% confidence interval across 100 stochastic simulations for each wave.



**Appendix Figure 3.** Comparison between the estimated mean viral loads for treated versus untreated cases. Lines and shading indicate the means and 95% confidence intervals of SARS-CoV-2 viral load (RNA log<sub>10</sub> copies/mL) as estimated by the fitted within-host model. Day one corresponds to the initiation of treatment.

**Appendix Table 1.** Parameter values.

Parameters	Values	Source
$N$ : number of individuals in the population	9961 individuals in 5000 households	
Initial percent vaccinated as of January 29, 2022 for [0–4y, 5–17y, 18–49y, 50–64y, >65y]	At least one dose: [0, 42.4%, 77.2%, 89.6%, 97.4%] One dose only: [0%, 4.50%, 2.30%, 1.70%, and 1.20%] Primary series only (two doses): [0%, 37.90%, 29.60%, 25.80%, and 21.30%] Primary and first booster: [0%, 0%, 45.30%, 62.10%, and 74.90%]	(16)
Initial infections (in exposed state)	1% of the susceptible population	Assumed
Initial proportion of recovered	[39.34%, 60.81%, 50.61%, 48.38%, and 35.88%] for [0–4y, 5–17y, 18–49y, 50–64y, >65y]	(i) 77M, 129M, and 220M reported cases in the U.S. on 29 January 2021, 30 September 2021, and 29 January 2022, respectively (10). (ii) U.S. CDC estimates [35.49%, 54.86%, 45.66%, 43.65%, 32.36%] for [0–4y, 5–17y, 18–49y, 50–64y, >65y] from February 2020 to September 2021 (24). (iii) Thus we set the age-specific proportions of the U.S. population infected between January 29, 2021 and January 29, 2022 to $(220-77)/129 \cdot [35.49\%, 54.86\%, 45.66\%, 43.65\%, 32.36\%]$ .
$\psi$ : symptomatic proportion (%)	75	(25)
$\rho$ : treatment proportion (%)	Depends on scenario	Assumed
$\xi_{nh}$ : baseline transmission rate	$\xi_{nh}$ is 0.0004, 0.0005, 0.0006, 0.0008, and 0.0013 for $R_e$ of 1.2, 1.5, 1.7, 2, and 3, respectively.	Calibrated to $R_e$ .
$\sigma$ : transition rate out of exposed state ( $d^{-1}$ )	1/3	(3)
$\gamma_A$ : asymptomatic recovery rate ( $d^{-1}$ )	1/9	(17)
$\gamma_Y$ : symptomatic recovery rate ( $d^{-1}$ )	1/4	(26,27)
$\omega$ : transition rate from the pre-symptomatic to the symptomatic stage ( $d^{-1}$ )	1/2	(3)
$h_a$ : age-specific proportion of symptomatic cases that are hospitalized	[0%, 0.025%, 2.672%, 9.334%, 15.465%] for [0–4y, 5–17y, 18–49y, 50–64y, >65y]	(28)
$\phi_a$ : age-specific efficacy of Paxlovid in reducing the hospitalization rate	[0.59 (95% CI: 0.48, 0.71), 0.59 (95% CI: 0.48, 0.71), 0.59 (95% CI: 0.48, 0.71), 0.40 (95% CI: 0.34, 0.48), 0.53 (95% CI: 0.48, 0.58)] for [0–4y, 5–17y, 18–49y, 50–64y, >65y]	(18) The study provides estimates for adults over age 18y. We assumed that efficacy for children under 18 is the same as that for adults aged 18–49y.
$\eta$ : transition rate from treatment to hospitalized ( $d^{-1}$ )	$1/(5.9-1/\gamma_T)$	5.9 d on average from symptomatic to hospitalized (29)
$\gamma_T$ : transition rate from symptomatic to treatment ( $d^{-1}$ )	1/3	an average of 3 d between COVID-19 symptom onset and the initiation of Paxlovid treatment (4)
$\mu_a$ : age-specific mortality rate for hospitalized cases	[0.48%, 0.48%, 5.68%, 10.82%, 16.15%] for [0–4y, 5–17y, 18–49y, 50–64y, >65y]	(30)
$\gamma_d$ : transition rate from hospitalized to deceased for cases that succumb ( $d^{-1}$ )	0.128	(31)
$\gamma_h$ : transition rate from hospitalized to recovered for cases discharged alive ( $d^{-1}$ )	0.091	(31)
$\lambda_a$ : life expectancy (years) for age group $a$ , adjusted assuming a 3% yearly discount rate	[30.3, 29.3, 25.8, 18.7, 12.9] for [0–4y, 5–17y, 18–49y, 50–64y, >65y]	(32)



**Appendix Table 2.** Parameters governing waning of immunity following vaccination and infection with respect to the SARS-CoV-2 Omicron variant. Vaccine-related parameter values are based on estimates for the BNT162b2 (Pfizer) vaccine.

Time since most recent vaccination	Immunity following vaccination		
	Reduction in susceptibility to infection ( $\omega_B$ )	Reduction in likelihood of developing symptoms following infection ( $\psi_B$ )	Reduction in risk of mortality following infection ( $\theta_B$ )
1 week	64%*	66.9% (73)	91.9%*
2 to 4 weeks	64%*	67.2% (73)	91.9%*
5 to 9 weeks	42.8%*	55.0% (73)	91.9%*
> = 10 weeks	22%*	45.7% (73)	91.9%*
Time since recovery	Immunity following infection		
	Reduction in susceptibility to infection ( $\omega_N$ )	Reduction in likelihood of developing symptoms following infection ( $\psi_N$ )	Reduction in risk of mortality following infection ( $\theta_N$ )
1 week	83.1%+	92.1% <sup>++</sup>	98.1% <sup>++</sup>
2 to 4 weeks	83.1%+	92.1% <sup>++</sup>	98.1% <sup>++</sup>
5 to 9 weeks	73.1%+	89.2% <sup>++</sup>	98.1% <sup>++</sup>
> = 10 weeks	63.3%+	87.0% <sup>++</sup>	98.1% <sup>++</sup>

\* Since direct estimates of  $\omega_B$  and  $\theta_B$  for vaccine booster doses against the Omicron variant are not available, we extrapolated from Ref (33), which estimates  $\omega_B$  for boosters against Omicron and other studies that simultaneously estimate vaccine efficacy against infection, symptomatic infection, and mortality for the Pfizer-BioNTech BNT162b2 vaccine earlier in the pandemic. In Ref (34), vaccine efficacy against infection is estimated to be 64% when efficacy against symptomatic disease reaches 67% (21 d after vaccination). In Ref (35), vaccine efficacy against infection is estimated to be 42.8% when efficacy against symptomatic disease reaches 52.4% (14 d after the initial vaccine dose). In Ref (34), vaccine efficacy against infection is estimated to be 22% when efficacy against symptomatic disease reaches 44% (14 d after the initial vaccine dose). In Ref (36), vaccine efficacy against mortality is estimated to be 91.9% when efficacy against symptomatic disease reaches 66.3% (20 d after the second vaccine dose).

\*\* Since direct estimates for  $\psi_N$  against the Omicron variant are not available, we extrapolated from Ref (37), which estimates  $\omega_N$  against Omicron and Ref (38), which reports that vaccine efficacy against symptomatic disease is 97% when efficacy against infection reaches 79% (14 d after the second vaccine dose).

+ Using an estimated adjusted hazard ratio of SARS-CoV-2 infection following natural infection versus BNT162b2 vaccination of 0.47 (95% CI: 0.45–0.48), we scaled  $\omega_N$  based on  $\omega_B$  (39).

\*\* Using an estimated adjusted hazard ratio of severe SARS-CoV-2 infection following natural infection versus BNT162b2 vaccination of 0.24 (95% CI: 0.08–0.72), we scaled  $\psi_N$  based on  $\psi_B$  (39) and  $\theta_N$  based on  $\theta_B$ .

**Appendix Table 3.** Cost parameters.

Parameter	Value (USD)
Cost of administering a treatment course ( $c_T$ ) by taking Paxlovid	\$530 per course (40)
Median COVID-19 hospitalization cost by age group ( $c_{H,a}$ )	[\$21,847, \$21,847, \$19,681, \$23,157, \$18,806] for [0–4y, 5–17y, 18–49y, 50–64y, >65y] (41)

**Appendix Table 4.** Projected cases, hospitalizations and deaths averted in the U.S., net monetary benefit (billion USD), and treatment courses administered in a large-scale SARS-CoV-2 antiviral campaign, across four treatment rates and five transmission scenarios (effective reproduction numbers from 1.2 to 5). Assuming a treatment course cost of U.S.\$530 and willingness to pay (WTP) per year of life lost (YLL) averted of U.S.\$100,000, for each reproduction number and treatment rate, we estimated the mean and 95% confidence intervals based on 100 pairs of stochastic simulations (treatment vs. no treatment simulations). To separate the direct therapeutic benefits of the drug from the indirect transmission-blocking impacts of treatment, we analyzed an alternative model in which the drug improves patient outcomes but does not impact infectivity.

Outcome	$R_e$	Treatment rate (% symptomatic cases)	Mean (95% CI)		
			Alternative model: indirect (transmission reducing) effect only	Alternative model: direct (therapeutic) effects only	Base model
No. of cases averted (million)	1.2	20%	10.57 (3.03, 21.19)	-0.22 (-3.82, 1.91)	10.54 (3.03, 21.12)
		50%	25.99 (12.88, 45.11)	-0.32 (-4.84, 2.37)	25.65 (12.59, 41.19)
		80%	43.13 (25.30, 80.72)	-0.61 (-5.34, 2.90)	42.58 (25.24, 67.45)
		100%	54.21 (36.77, 81.94)	-0.84 (-6.19, 3.26)	53.74 (36.67, 81.88)
	1.5	20%	4.92 (0.07, 9.32)	-0.28 (-2.14, 0.99)	4.85 (-0.03, 9.29)
		50%	13.76 (7.55, 19.54)	-0.44 (-3.69, 1.98)	13.55 (7.51, 19.21)
		80%	23.71 (16.38, 30.21)	-0.76 (-4.68, 2.54)	23.43 (16.21, 29.92)
		100%	31.33 (23.43, 41.35)	-0.84 (-4.61, 2.17)	30.84 (23.39, 40.69)
	1.7	20%	4.26 (0.07, 8.37)	-0.13 (-2.14, 1.45)	4.25 (0.00, 8.30)
		50%	10.87 (6.16, 16.70)	-0.33 (-2.64, 1.91)	10.65 (5.77, 16.70)
		80%	18.39 (10.54, 24.41)	-0.58 (-2.73, 1.52)	18.15 (10.51, 24.09)
		100%	23.87 (16.47, 30.11)	-0.57 (-2.73, 1.32)	23.50 (15.85, 29.98)
2	20%	2.93 (-0.23, 6.26)	-0.10 (-1.65, 1.55)	2.86 (-0.26, 6.16)	
	50%	7.14 (3.03, 11.66)	-0.39 (-1.94, 1.05)	6.96 (3.00, 11.73)	
	80%	12.01 (7.15, 18.62)	-0.71 (-2.77, 0.82)	11.63 (6.89, 18.48)	
	100%	15.55 (11.27, 21.38)	-0.86 (-3.00, 1.19)	15.19 (10.81, 20.99)	
3	20%	0.71 (-0.07, 1.45)	-0.05 (-0.49, 0.26)	0.67 (-0.13, 1.45)	

Outcome	R <sub>e</sub>	Treatment rate (% symptomatic cases)	Mean (95% CI)		
			Alternative model: indirect (transmission reducing) effect only	Alternative model: direct (therapeutic) effects only	Base model
Deaths averted (thousand)	5	50%	1.74 (0.79, 2.83)	-0.13 (-0.76, 0.40)	1.68 (0.79, 2.77)
		80%	2.92 (1.94, 4.18)	-0.21 (-0.82, 0.43)	2.80 (1.78, 4.02)
		100%	3.77 (2.67, 5.07)	-0.24 (-0.79, 0.46)	3.67 (2.67, 5.01)
	1.2	20%	0.03 (-0.10, 0.13)	0.00 (-0.07, 0.03)	0.03 (-0.07, 0.13)
		50%	0.08 (-0.07, 0.20)	0.00 (-0.10, 0.10)	0.07 (-0.07, 0.20)
		80%	0.11 (-0.07, 0.33)	-0.01 (-0.10, 0.07)	0.11 (-0.07, 0.30)
	1.5	100%	0.14 (-0.07, 0.33)	-0.02 (-0.13, 0.10)	0.12 (-0.07, 0.33)
		20%	19.46 (-14.14, 58.52)	16.47 (-19.47, 48.11)	33.85 (1.69, 71.15)
		50%	48.14 (-8.82, 117.86)	43.36 (1.58, 91.31)	79.11 (35.78, 146.51)
	1.7	80%	76.95 (18.23, 148.21)	65.25 (17.88, 121.49)	113.96 (53.68, 166.28)
		100%	95.19 (20.40, 165.92)	81.27 (17.71, 143.13)	133.31 (60.98, 185.92)
		20%	13.07 (-49.99, 78.46)	36.87 (-13.91, 86.89)	50.23 (-1.93, 114.14)
	2	50%	35.75 (-48.24, 121.17)	94.77 (5.56, 164.45)	123.79 (39.24, 201.48)
		80%	61.77 (-71.70, 162.12)	150.07 (66.01, 238.88)	188.19 (89.04, 277.98)
		100%	82.50 (-48.31, 178.02)	186.98 (83.38, 276.20)	231.40 (124.93, 335.46)
	3	20%	11.93 (-44.68, 78.58)	48.49 (7.54, 104.72)	59.43 (9.13, 129.86)
		50%	35.10 (-62.19, 134.72)	120.77 (28.96, 197.79)	145.44 (45.60, 221.34)
		80%	52.69 (-43.96, 174.65)	190.28 (94.44, 297.07)	221.79 (115.52, 315.25)
	5	100%	70.01 (-40.05, 211.69)	238.48 (136.21, 349.11)	272.44 (180.06, 392.08)
		20%	9.02 (-78.52, 74.64)	61.73 (5.45, 117.52)	70.75 (4.05, 144.58)
		50%	21.92 (-76.91, 129.39)	153.74 (64.08, 240.41)	174.36 (83.69, 271.33)
	1.2	80%	44.70 (-75.05, 166.76)	244.56 (108.42, 341.85)	275.73 (158.47, 377.52)
		100%	57.57 (-94.86, 173.98)	303.84 (158.23, 408.09)	338.93 (209.58, 456.58)
		20%	7.76 (-109.69, 76.28)	103.03 (41.24, 174.82)	109.67 (35.95, 179.83)
1.5	50%	16.97 (-91.08, 123.79)	257.11 (148.62, 362.85)	266.69 (156.71, 362.77)	
	80%	31.25 (-113.81, 158.07)	404.90 (266.90, 515.39)	425.24 (277.20, 552.68)	
	100%	41.88 (-109.69, 182.87)	507.09 (359.15, 646.19)	525.51 (384.94, 663.72)	
1.7	20%	0.90 (-120.85, 116.60)	154.32 (66.30, 259.38)	161.81 (56.01, 261.83)	
	50%	-0.16 (-175.10, 154.01)	387.76 (231.08, 547.49)	399.03 (246.21, 562.19)	
	80%	2.05 (-263.34, 205.53)	620.23 (444.13, 801.75)	632.56 (447.52, 812.49)	
NMB (billion USD)	100%	-6.88 (-310.29, 245.90)	768.62 (569.70, 997.59)	782.03 (586.95, 1002.60)	
	20%	31.17 (-32.77, 103.74)	25.35 (-35.19, 84.22)	56.95 (2.62, 122.63)	
	50%	80.36 (-24.60, 205.73)	68.03 (-6.92, 151.96)	135.60 (62.52, 261.32)	
1.2	80%	130.70 (26.22, 254.08)	102.12 (13.07, 189.75)	197.15 (97.62, 293.82)	
	100%	163.44 (22.57, 288.36)	126.96 (16.21, 228.25)	232.26 (107.17, 332.85)	
	20%	15.55 (-90.17, 134.55)	58.18 (-19.35, 127.21)	81.07 (-10.22, 194.06)	
1.5	50%	45.58 (-99.51, 193.46)	149.22 (-1.33, 276.58)	201.38 (50.16, 327.12)	
	80%	83.46 (-140.08, 282.27)	238.03 (85.13, 389.40)	307.42 (125.44, 449.21)	
	100%	115.04 (-99.49, 299.60)	295.70 (111.03, 447.24)	378.72 (198.82, 560.47)	
1.7	20%	12.67 (-82.30, 133.89)	76.14 (5.94, 157.33)	95.66 (8.54, 196.23)	
	50%	39.00 (-130.16, 170.20)	190.12 (32.17, 332.79)	232.35 (80.45, 379.51)	
	80%	60.66 (-112.13, 257.44)	300.90 (109.99, 451.75)	356.40 (176.38, 499.36)	
2	100%	86.49 (-102.48, 328.03)	377.20 (196.17, 547.91)	439.82 (266.32, 610.84)	
	20%	4.57 (-122.98, 126.50)	96.74 (3.28, 194.97)	111.39 (4.57, 246.24)	
	50%	14.18 (-169.44, 191.31)	240.68 (78.62, 396.23)	276.52 (97.96, 459.36)	
3	80%	41.73 (-184.27, 232.06)	385.48 (139.63, 559.60)	439.08 (208.27, 618.77)	
	100%	54.86 (-194.98, 282.01)	478.73 (205.19, 687.67)	539.73 (309.83, 764.58)	
	20%	-2.66 (-172.88, 125.58)	159.56 (62.97, 284.89)	170.17 (60.49, 286.14)	
5	50%	-4.31 (-179.49, 193.58)	401.32 (219.88, 578.78)	417.18 (208.34, 580.13)	
	80%	1.64 (-239.00, 234.46)	633.73 (384.92, 820.60)	665.92 (414.82, 878.32)	
	100%	5.43 (-272.37, 270.19)	789.52 (541.89, 982.78)	821.12 (595.37, 1059.33)	
1.2	20%	-24.98 (-212.90, 142.87)	234.19 (101.47, 417.79)	247.08 (92.89, 409.43)	
	50%	-53.96 (-348.55, 188.46)	596.98 (351.54, 838.32)	616.81 (366.77, 863.68)	
	80%	-77.94 (-500.47, 261.01)	960.35 (646.44, 1267.53)	982.74 (685.66, 1298.68)	
Treatment courses used (million)	100%	-112.37 (-612.92, 327.29)	1189.80 (860.17, 1567.85)	1214.09 (885.29, 1605.70)	
	20%	5.77 (4.38, 7.08)	6.45 (4.78, 7.68)	5.77 (4.38, 7.15)	
	50%	12.08 (7.31, 14.89)	16.24 (12.55, 18.42)	12.13 (8.86, 14.89)	
1.5	80%	14.90 (5.93, 20.92)	26.07 (21.35, 29.42)	15.04 (6.79, 21.02)	
	100%	15.15 (4.94, 21.78)	32.68 (26.66, 36.80)	15.30 (5.93, 21.91)	
	20%	11.64 (10.31, 12.88)	11.92 (10.84, 12.88)	11.64 (10.35, 12.92)	
1.7	50%	27.54 (25.60, 29.39)	29.74 (27.51, 31.73)	27.57 (25.57, 29.36)	
	80%	41.50 (38.68, 44.88)	47.60 (44.88, 50.28)	41.58 (38.78, 45.04)	
	100%	49.52 (45.17, 53.01)	59.62 (56.41, 63.26)	49.69 (45.34, 52.98)	
1.2	20%	13.58 (12.42, 15.09)	13.88 (12.62, 15.42)	13.57 (12.42, 15.12)	
	50%	32.81 (31.04, 34.73)	34.60 (32.55, 36.70)	32.85 (30.87, 34.76)	

Outcome	$R_e$	Treatment rate (% symptomatic cases)	Mean (95% CI)			
			Alternative model: indirect (transmission reducing) effect only	Alternative model: direct (therapeutic) effects only	Base model	
	2	80%	50.70 (47.61, 53.54)	55.38 (52.09, 58.38)	50.74 (47.61, 53.90)	
		100%	61.49 (58.02, 63.95)	69.26 (65.30, 73.05)	61.61 (58.22, 64.28)	
		20%	16.32 (14.73, 17.96)	16.54 (14.83, 18.09)	16.32 (14.73, 18.02)	
	3	50%	40.03 (37.96, 42.40)	41.40 (39.11, 44.35)	40.07 (37.86, 42.47)	
		80%	62.78 (58.68, 66.36)	66.37 (62.57, 70.02)	62.89 (58.58, 66.19)	
		100%	77.32 (72.98, 81.94)	83.17 (78.45, 87.68)	77.43 (72.98, 81.81)	
	5	20%	24.42 (22.31, 26.62)	24.58 (22.50, 26.72)	24.41 (22.34, 26.56)	
		50%	60.16 (57.00, 63.13)	61.43 (58.45, 64.41)	60.21 (57.07, 63.16)	
		80%	94.96 (91.56, 98.71)	98.13 (94.92, 101.78)	95.06 (91.70, 98.68)	
	Hospitalizations reduced (million)	1.2	100%	118.06 (114.50, 121.78)	122.84 (119.41, 126.75)	118.21 (114.89, 122.17)
			20%	38.42 (35.98, 41.09)	38.57 (36.08, 41.15)	38.43 (35.98, 41.19)
			50%	95.25 (91.20, 99.27)	96.23 (92.29, 100.16)	95.34 (91.50, 99.34)
1.5		80%	151.64 (146.42, 156.47)	154.18 (150.34, 159.50)	151.84 (146.62, 156.83)	
		100%	189.18 (184.81, 194.49)	193.11 (188.47, 198.65)	189.44 (185.27, 195.02)	
		20%	0.16 (-0.13, 0.53)	0.14 (-0.13, 0.40)	0.28 (0.03, 0.59)	
1.7		50%	0.41 (-0.07, 0.99)	0.36 (0.00, 0.76)	0.67 (0.33, 1.25)	
		80%	0.65 (0.16, 1.22)	0.55 (0.10, 0.96)	0.96 (0.49, 1.42)	
		100%	0.80 (0.16, 1.35)	0.68 (0.20, 1.15)	1.13 (0.56, 1.58)	
2		20%	0.10 (-0.40, 0.66)	0.30 (-0.03, 0.63)	0.41 (-0.03, 0.92)	
		50%	0.28 (-0.40, 0.96)	0.77 (0.07, 1.38)	1.01 (0.30, 1.58)	
		80%	0.48 (-0.43, 1.38)	1.23 (0.49, 1.91)	1.53 (0.66, 2.14)	
3	100%	0.65 (-0.36, 1.52)	1.53 (0.66, 2.24)	1.88 (0.99, 2.67)		
	20%	0.09 (-0.33, 0.66)	0.39 (0.07, 0.79)	0.48 (0.07, 0.92)		
	50%	0.26 (-0.49, 0.86)	0.97 (0.26, 1.68)	1.16 (0.49, 1.85)		
5	80%	0.40 (-0.43, 1.32)	1.54 (0.69, 2.27)	1.78 (0.89, 2.44)		
	100%	0.55 (-0.36, 1.61)	1.93 (1.15, 2.73)	2.19 (1.38, 2.93)		
	20%	0.06 (-0.46, 0.59)	0.49 (0.03, 0.92)	0.55 (0.10, 1.19)		
3	50%	0.16 (-0.72, 1.02)	1.22 (0.46, 1.94)	1.38 (0.59, 2.24)		
	80%	0.34 (-0.76, 1.19)	1.95 (0.86, 2.80)	2.19 (1.12, 3.00)		
	100%	0.44 (-0.69, 1.42)	2.43 (1.19, 3.43)	2.69 (1.65, 3.72)		
5	20%	0.05 (-0.69, 0.59)	0.80 (0.40, 1.32)	0.85 (0.36, 1.38)		
	50%	0.13 (-0.63, 1.12)	2.01 (1.15, 2.83)	2.08 (1.12, 2.83)		
	80%	0.24 (-0.82, 1.35)	3.17 (2.08, 4.05)	3.30 (2.14, 4.22)		
5	100%	0.31 (-0.89, 1.48)	3.95 (2.90, 4.84)	4.08 (3.03, 5.11)		
	20%	-0.03 (-0.89, 0.72)	1.17 (0.59, 2.04)	1.23 (0.49, 1.94)		
	50%	-0.02 (-1.32, 1.05)	2.99 (1.88, 3.99)	3.08 (1.94, 4.25)		
5	80%	0.00 (-1.91, 1.61)	4.81 (3.26, 6.19)	4.91 (3.59, 6.36)		
	100%	-0.06 (-2.14, 1.98)	5.97 (4.48, 7.74)	6.07 (4.68, 7.78)		

**Appendix Table 5. The optimal choice for Paxlovid treatment under a range of SARS-CoV-2 transmission scenarios.** We estimated the mean (95% confidence interval [CI]) of the number of cases infected averted (millions), number of deaths averted (thousands), number of hospitalizations averted (million), number of courses administered (millions), and net monetary benefit (NMB) in billions of USD, in contrast with baseline, which is scaled to a U.S. population of 328.2 million (21) (Appendix Table 6). Each scenario V1-V3 changes one of the base assumptions, as indicated in the second column. Values in the third column are mean and 95% CI in the transmission scenarios ( $R_e = 1.2$  and Treatment rate = 20%) as examples.

Scenarios	$R_e = 1.2$ Treatment rate = 20%	
	Incremental net monetary benefits (\$ billion), mean (95% CI)	
Base	Log relationship between infectiousness and viral load	
V1	Log-proportional relationship between infectiousness and viral load <sup>+</sup>	
V2	Step relationship between infectiousness and viral load <sup>&amp;</sup>	
V3	Sigmoid relationship between infectiousness and viral load <sup>*</sup>	
	56.95	(2.62, 122.63)
	58.42	(-22.99, 147.37)
	65.31	(0.71, 141.95)
	53.78	(-24.03, 125.87)

<sup>+</sup> Infectiousness is proportional log<sub>10</sub> of viral load for values above 10<sup>6</sup>, as given by log<sub>10</sub>(Viral load)-6, and is set to zero otherwise (42).

<sup>&</sup> Infectiousness is a constant for viral loads above 10<sup>6</sup>, and is set to zero otherwise (42).

<sup>\*</sup> Infectiousness has the sigmoid relationship with viral load following the association between viral load and cell culture isolation success rate (43).

**Appendix Table 6.** Dataset

Scenarios	Infectiousness	Measures	$R_t$	Treatment %	Direct	Indirect	All (Direct + Indirect)
Base	log10	Number of cases averted in U.S. (million)	1.2	20%	-0.22 (-3.82, 1.91)	10.57 (3.03, 21.19)	10.54 (3.03, 21.12)
Base	log10	Number of cases averted in U.S. (million)	1.2	50%	-0.32 (-4.84, 2.37)	25.99 (12.88, 45.11)	25.65 (12.59, 41.19)
Base	log10	Number of cases averted in U.S. (million)	1.2	80%	-0.61 (-5.34, 2.90)	43.13 (25.30, 80.72)	42.58 (25.24, 67.45)
Base	log10	Number of cases averted in U.S. (million)	1.2	100%	-0.84 (-6.19, 3.26)	54.21 (36.77, 81.94)	53.74 (36.67, 81.88)
Base	log10	Number of cases averted in U.S. (million)	1.5	20%	-0.28 (-2.14, 0.99)	4.92 (0.07, 9.32)	4.85 (-0.03, 9.29)
Base	log10	Number of cases averted in U.S. (million)	1.5	50%	-0.44 (-3.69, 1.98)	13.76 (7.55, 19.54)	13.55 (7.51, 19.21)
Base	log10	Number of cases averted in U.S. (million)	1.5	80%	-0.76 (-4.68, 2.54)	23.71 (16.38, 30.21)	23.43 (16.21, 29.92)
Base	log10	Number of cases averted in U.S. (million)	1.5	100%	-0.84 (-4.61, 2.17)	31.33 (23.43, 41.35)	30.84 (23.39, 40.69)
Base	log10	Number of cases averted in U.S. (million)	1.7	20%	-0.13 (-2.14, 1.45)	4.26 (0.07, 8.37)	4.25 (0.00, 8.30)
Base	log10	Number of cases averted in U.S. (million)	1.7	50%	-0.33 (-2.64, 1.91)	10.87 (6.16, 16.70)	10.65 (5.77, 16.70)
Base	log10	Number of cases averted in U.S. (million)	1.7	80%	-0.58 (-2.73, 1.52)	18.39 (10.54, 24.41)	18.15 (10.51, 24.09)
Base	log10	Number of cases averted in U.S. (million)	1.7	100%	-0.57 (-2.73, 1.32)	23.87 (16.47, 30.11)	23.50 (15.85, 29.98)
Base	log10	Number of cases averted in U.S. (million)	2	20%	-0.10 (-1.65, 1.55)	2.93 (-0.23, 6.26)	2.86 (-0.26, 6.16)
Base	log10	Number of cases averted in U.S. (million)	2	50%	-0.39 (-1.94, 1.05)	7.14 (3.03, 11.66)	6.96 (3.00, 11.73)
Base	log10	Number of cases averted in U.S. (million)	2	80%	-0.71 (-2.77, 0.82)	12.01 (7.15, 18.62)	11.63 (6.89, 18.48)
Base	log10	Number of cases averted in U.S. (million)	2	100%	-0.86 (-3.00, 1.19)	15.55 (11.27, 21.38)	15.19 (10.81, 20.99)
Base	log10	Number of cases averted in U.S. (million)	3	20%	-0.05 (-0.49, 0.26)	0.71 (-0.07, 1.45)	0.67 (-0.13, 1.45)
Base	log10	Number of cases averted in U.S. (million)	3	50%	-0.13 (-0.76, 0.40)	1.74 (0.79, 2.83)	1.68 (0.79, 2.77)
Base	log10	Number of cases averted in U.S. (million)	3	80%	-0.21 (-0.82, 0.43)	2.92 (1.94, 4.18)	2.80 (1.78, 4.02)
Base	log10	Number of cases averted in U.S. (million)	3	100%	-0.24 (-0.79, 0.46)	3.77 (2.67, 5.07)	3.67 (2.67, 5.01)
Base	log10	Number of cases averted in U.S. (million)	5	20%	0.00 (-0.07, 0.03)	0.03 (-0.10, 0.13)	0.03 (-0.07, 0.13)
Base	log10	Number of cases averted in U.S. (million)	5	50%	0.00 (-0.10, 0.10)	0.08 (-0.07, 0.20)	0.07 (-0.07, 0.20)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
Base	log10	Number of cases averted in U.S. (million)	5	80%	-0.01 (-0.10, 0.07)	0.11 (-0.07, 0.33)	0.11 (-0.07, 0.30)
Base	log10	Number of cases averted in U.S. (million)	5	100%	-0.02 (-0.13, 0.10)	0.14 (-0.07, 0.33)	0.12 (-0.07, 0.33)
Base	log10	Deaths reduced (thousand)	1.2	20%	16.47 (-19.47, 48.11)	19.46 (-14.14, 58.52)	33.85 (1.69, 71.15)
Base	log10	Deaths reduced (thousand)	1.2	50%	43.36 (1.58, 91.31)	48.14 (-8.82, 117.86)	79.11 (35.78, 146.51)
Base	log10	Deaths reduced (thousand)	1.2	80%	65.25 (17.88, 121.49)	76.95 (18.23, 148.21)	113.96 (53.68, 166.28)
Base	log10	Deaths reduced (thousand)	1.2	100%	81.27 (17.71, 143.13)	95.19 (20.40, 165.92)	133.31 (60.98, 185.92)
Base	log10	Deaths reduced (thousand)	1.5	20%	36.87 (-13.91, 86.89)	13.07 (-49.99, 78.46)	50.23 (-1.93, 114.14)
Base	log10	Deaths reduced (thousand)	1.5	50%	94.77 (5.56, 164.45)	35.75 (-48.24, 121.17)	123.79 (39.24, 201.48)
Base	log10	Deaths reduced (thousand)	1.5	80%	150.07 (66.01, 238.88)	61.77 (-71.70, 162.12)	188.19 (89.04, 277.98)
Base	log10	Deaths reduced (thousand)	1.5	100%	186.98 (83.38, 276.20)	82.50 (-48.31, 178.02)	231.40 (124.93, 335.46)
Base	log10	Deaths reduced (thousand)	1.7	20%	48.49 (7.54, 104.72)	11.93 (-44.68, 78.58)	59.43 (9.13, 129.86)
Base	log10	Deaths reduced (thousand)	1.7	50%	120.77 (28.96, 197.79)	35.10 (-62.19, 134.72)	145.44 (45.60, 221.34)
Base	log10	Deaths reduced (thousand)	1.7	80%	190.28 (94.44, 297.07)	52.69 (-43.96, 174.65)	221.79 (115.52, 315.25)
Base	log10	Deaths reduced (thousand)	1.7	100%	238.48 (136.21, 349.11)	70.01 (-40.05, 211.69)	272.44 (180.06, 392.08)
Base	log10	Deaths reduced (thousand)	2	20%	61.73 (5.45, 117.52)	9.02 (-78.52, 74.64)	70.75 (4.05, 144.58)
Base	log10	Deaths reduced (thousand)	2	50%	153.74 (64.08, 240.41)	21.92 (-76.91, 129.39)	174.36 (83.69, 271.33)
Base	log10	Deaths reduced (thousand)	2	80%	244.56 (108.42, 341.85)	44.70 (-75.05, 166.76)	275.73 (158.47, 377.52)
Base	log10	Deaths reduced (thousand)	2	100%	303.84 (158.23, 408.09)	57.57 (-94.86, 173.98)	338.93 (209.58, 456.58)
Base	log10	Deaths reduced (thousand)	3	20%	103.03 (41.24, 174.82)	7.76 (-109.69, 76.28)	109.67 (35.95, 179.83)
Base	log10	Deaths reduced (thousand)	3	50%	257.11 (148.62, 362.85)	16.97 (-91.08, 123.79)	266.69 (156.71, 362.77)
Base	log10	Deaths reduced (thousand)	3	80%	404.90 (266.90, 515.39)	31.25 (-113.81, 158.07)	425.24 (277.20, 552.68)
Base	log10	Deaths reduced (thousand)	3	100%	507.09 (359.15, 646.19)	41.88 (-109.69, 182.87)	525.51 (384.94, 663.72)
Base	log10	Deaths reduced (thousand)	5	20%	154.32 (66.30, 259.38)	0.90 (-120.85, 116.60)	161.81 (56.01, 261.83)
Base	log10	Deaths reduced (thousand)	5	50%	387.76 (231.08, 547.49)	-0.16 (-175.10, 154.01)	399.03 (246.21, 562.19)
Base	log10	Deaths reduced (thousand)	5	80%	620.23 (444.13, 801.75)	2.05 (-263.34, 205.53)	632.56 (447.52, 812.49)
Base	log10	Deaths reduced (thousand)	5	100%	768.62 (569.70, 997.59)	-6.88 (-310.29, 245.90)	782.03 (586.95, 1002.60)
Base	log10	NMB (\$ billion)	1.2	20%	25.35 (-35.19, 84.22)	31.17 (-32.77, 103.74)	56.95 (2.62, 122.63)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
Base	log10	NMB (\$ billion)	1.2	50%	68.03 (-6.92, 151.96)	80.36 (-24.60, 205.73)	135.60 (62.52, 261.32)
Base	log10	NMB (\$ billion)	1.2	80%	102.12 (13.07, 189.75)	130.70 (26.22, 254.08)	197.15 (97.62, 293.82)
Base	log10	NMB (\$ billion)	1.2	100%	126.96 (16.21, 228.25)	163.44 (22.57, 288.36)	232.26 (107.17, 332.85)
Base	log10	NMB (\$ billion)	1.5	20%	58.18 (-19.35, 127.21)	15.55 (-90.17, 134.55)	81.07 (-10.22, 194.06)
Base	log10	NMB (\$ billion)	1.5	50%	149.22 (-1.33, 276.58)	45.58 (-99.51, 193.46)	201.38 (50.16, 327.12)
Base	log10	NMB (\$ billion)	1.5	80%	238.03 (85.13, 389.40)	83.46 (-140.08, 282.27)	307.42 (125.44, 449.21)
Base	log10	NMB (\$ billion)	1.5	100%	295.70 (111.03, 447.24)	115.04 (-99.49, 299.60)	378.72 (198.82, 560.47)
Base	log10	NMB (\$ billion)	1.7	20%	76.14 (5.94, 157.33)	12.67 (-82.30, 133.89)	95.66 (8.54, 196.23)
Base	log10	NMB (\$ billion)	1.7	50%	190.12 (32.17, 332.79)	39.00 (-130.16, 170.20)	232.35 (80.45, 379.51)
Base	log10	NMB (\$ billion)	1.7	80%	300.90 (109.99, 451.75)	60.66 (-112.13, 257.44)	356.40 (176.38, 499.36)
Base	log10	NMB (\$ billion)	1.7	100%	377.20 (196.17, 547.91)	86.49 (-102.48, 328.03)	439.82 (266.32, 610.84)
Base	log10	NMB (\$ billion)	2	20%	96.74 (3.28, 194.97)	4.57 (-122.98, 126.50)	111.39 (4.57, 246.24)
Base	log10	NMB (\$ billion)	2	50%	240.68 (78.62, 396.23)	14.18 (-169.44, 191.31)	276.52 (97.96, 459.36)
Base	log10	NMB (\$ billion)	2	80%	385.48 (139.63, 559.60)	41.73 (-184.27, 232.06)	439.08 (208.27, 618.77)
Base	log10	NMB (\$ billion)	2	100%	478.73 (205.19, 687.67)	54.86 (-194.98, 282.01)	539.73 (309.83, 764.58)
Base	log10	NMB (\$ billion)	3	20%	159.56 (62.97, 284.89)	-2.66 (-172.88, 125.58)	170.17 (60.49, 286.14)
Base	log10	NMB (\$ billion)	3	50%	401.32 (219.88, 578.78)	-4.31 (-179.49, 193.58)	417.18 (208.34, 580.13)
Base	log10	NMB (\$ billion)	3	80%	633.73 (384.92, 820.60)	1.64 (-239.00, 234.46)	665.92 (414.82, 878.32)
Base	log10	NMB (\$ billion)	3	100%	789.52 (541.89, 982.78)	5.43 (-272.37, 270.19)	821.12 (595.37, 1059.33)
Base	log10	NMB (\$ billion)	5	20%	234.19 (101.47, 417.79)	-24.98 (-212.90, 142.87)	247.08 (92.89, 409.43)
Base	log10	NMB (\$ billion)	5	50%	596.98 (351.54, 838.32)	-53.96 (-348.55, 188.46)	616.81 (366.77, 863.68)
Base	log10	NMB (\$ billion)	5	80%	960.35 (646.44, 1267.53)	-77.94 (-500.47, 261.01)	982.74 (685.66, 1298.68)
Base	log10	NMB (\$ billion)	5	100%	1189.80 (860.17, 1567.85)	-112.37 (-612.92, 327.29)	1214.09 (885.29, 1605.70)
Base	log10	Treatment courses used (million)	1.2	20%	6.45 (4.78, 7.68)	5.77 (4.38, 7.08)	5.77 (4.38, 7.15)
Base	log10	Treatment courses used (million)	1.2	50%	16.24 (12.55, 18.42)	12.08 (7.31, 14.89)	12.13 (8.86, 14.89)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
Base	log10	Treatment courses used (million)	1.2	80%	26.07 (21.35, 29.42)	14.90 (5.93, 20.92)	15.04 (6.79, 21.02)
Base	log10	Treatment courses used (million)	1.2	100%	32.68 (26.66, 36.80)	15.15 (4.94, 21.78)	15.30 (5.93, 21.91)
Base	log10	Treatment courses used (million)	1.5	20%	11.92 (10.84, 12.88)	11.64 (10.31, 12.88)	11.64 (10.35, 12.92)
Base	log10	Treatment courses used (million)	1.5	50%	29.74 (27.51, 31.73)	27.54 (25.60, 29.39)	27.57 (25.57, 29.36)
Base	log10	Treatment courses used (million)	1.5	80%	47.60 (44.88, 50.28)	41.50 (38.68, 44.88)	41.58 (38.78, 45.04)
Base	log10	Treatment courses used (million)	1.5	100%	59.62 (56.41, 63.26)	49.52 (45.17, 53.01)	49.69 (45.34, 52.98)
Base	log10	Treatment courses used (million)	1.7	20%	13.88 (12.62, 15.42)	13.58 (12.42, 15.09)	13.57 (12.42, 15.12)
Base	log10	Treatment courses used (million)	1.7	50%	34.60 (32.55, 36.70)	32.81 (31.04, 34.73)	32.85 (30.87, 34.76)
Base	log10	Treatment courses used (million)	1.7	80%	55.38 (52.09, 58.38)	50.70 (47.61, 53.54)	50.74 (47.61, 53.90)
Base	log10	Treatment courses used (million)	1.7	100%	69.26 (65.30, 73.05)	61.49 (58.02, 63.95)	61.61 (58.22, 64.28)
Base	log10	Treatment courses used (million)	2	20%	16.54 (14.83, 18.09)	16.32 (14.73, 17.96)	16.32 (14.73, 18.02)
Base	log10	Treatment courses used (million)	2	50%	41.40 (39.11, 44.35)	40.03 (37.96, 42.40)	40.07 (37.86, 42.47)
Base	log10	Treatment courses used (million)	2	80%	66.37 (62.57, 70.02)	62.78 (58.68, 66.36)	62.89 (58.58, 66.19)
Base	log10	Treatment courses used (million)	2	100%	83.17 (78.45, 87.68)	77.32 (72.98, 81.94)	77.43 (72.98, 81.81)
Base	log10	Treatment courses used (million)	3	20%	24.58 (22.50, 26.72)	24.42 (22.31, 26.62)	24.41 (22.34, 26.56)
Base	log10	Treatment courses used (million)	3	50%	61.43 (58.45, 64.41)	60.16 (57.00, 63.13)	60.21 (57.07, 63.16)
Base	log10	Treatment courses used (million)	3	80%	98.13 (94.92, 101.78)	94.96 (91.56, 98.71)	95.06 (91.70, 98.68)
Base	log10	Treatment courses used (million)	3	100%	122.84 (119.41, 126.75)	118.06 (114.50, 121.78)	118.21 (114.89, 122.17)
Base	log10	Treatment courses used (million)	5	20%	38.57 (36.08, 41.15)	38.42 (35.98, 41.09)	38.43 (35.98, 41.19)
Base	log10	Treatment courses used (million)	5	50%	96.23 (92.29, 100.16)	95.25 (91.20, 99.27)	95.34 (91.50, 99.34)
Base	log10	Treatment courses used (million)	5	80%	154.18 (150.34, 159.50)	151.64 (146.42, 156.47)	151.84 (146.62, 156.83)
Base	log10	Treatment courses used (million)	5	100%	193.11 (188.47, 198.65)	189.18 (184.81, 194.49)	189.44 (185.27, 195.02)
Base	log10	Hospitalizations reduced (million)	1.2	20%	0.14 (-0.13, 0.40)	0.16 (-0.13, 0.53)	0.28 (0.03, 0.59)
Base	log10	Hospitalizations reduced (million)	1.2	50%	0.36 (0.00, 0.76)	0.41 (-0.07, 0.99)	0.67 (0.33, 1.25)
Base	log10	Hospitalizations reduced (million)	1.2	80%	0.55 (0.10, 0.96)	0.65 (0.16, 1.22)	0.96 (0.49, 1.42)
Base	log10	Hospitalizations reduced (million)	1.2	100%	0.68 (0.20, 1.15)	0.80 (0.16, 1.35)	1.13 (0.56, 1.58)
Base	log10	Hospitalizations reduced (million)	1.5	20%	0.30 (-0.03, 0.63)	0.10 (-0.40, 0.66)	0.41 (-0.03, 0.92)
Base	log10	Hospitalizations reduced (million)	1.5	50%	0.77 (0.07, 1.38)	0.28 (-0.40, 0.96)	1.01 (0.30, 1.58)
Base	log10	Hospitalizations reduced (million)	1.5	80%	1.23 (0.49, 1.91)	0.48 (-0.43, 1.38)	1.53 (0.66, 2.14)
Base	log10	Hospitalizations reduced (million)	1.5	100%	1.53 (0.66, 2.24)	0.65 (-0.36, 1.52)	1.88 (0.99, 2.67)
Base	log10	Hospitalizations reduced (million)	1.7	20%	0.39 (0.07, 0.79)	0.09 (-0.33, 0.66)	0.48 (0.07, 0.92)
Base	log10	Hospitalizations reduced (million)	1.7	50%	0.97 (0.26, 1.68)	0.26 (-0.49, 0.86)	1.16 (0.49, 1.85)

Scenarios	Infectiousness	Measures	$R_t$	Treatment %	Direct	Indirect	All (Direct + Indirect)
Base	log10	Hospitalizations reduced (million)	1.7	80%	1.54 (0.69, 2.27)	0.40 (-0.43, 1.32)	1.78 (0.89, 2.44)
Base	log10	Hospitalizations reduced (million)	1.7	100%	1.93 (1.15, 2.73)	0.55 (-0.36, 1.61)	2.19 (1.38, 2.93)
Base	log10	Hospitalizations reduced (million)	2	20%	0.49 (0.03, 0.92)	0.06 (-0.46, 0.59)	0.55 (0.10, 1.19)
Base	log10	Hospitalizations reduced (million)	2	50%	1.22 (0.46, 1.94)	0.16 (-0.72, 1.02)	1.38 (0.59, 2.24)
Base	log10	Hospitalizations reduced (million)	2	80%	1.95 (0.86, 2.80)	0.34 (-0.76, 1.19)	2.19 (1.12, 3.00)
Base	log10	Hospitalizations reduced (million)	2	100%	2.43 (1.19, 3.43)	0.44 (-0.69, 1.42)	2.69 (1.65, 3.72)
Base	log10	Hospitalizations reduced (million)	3	20%	0.80 (0.40, 1.32)	0.05 (-0.69, 0.59)	0.85 (0.36, 1.38)
Base	log10	Hospitalizations reduced (million)	3	50%	2.01 (1.15, 2.83)	0.13 (-0.63, 1.12)	2.08 (1.12, 2.83)
Base	log10	Hospitalizations reduced (million)	3	80%	3.17 (2.08, 4.05)	0.24 (-0.82, 1.35)	3.30 (2.14, 4.22)
Base	log10	Hospitalizations reduced (million)	3	100%	3.95 (2.90, 4.84)	0.31 (-0.89, 1.48)	4.08 (3.03, 5.11)
Base	log10	Hospitalizations reduced (million)	5	20%	1.17 (0.59, 2.04)	-0.03 (-0.89, 0.72)	1.23 (0.49, 1.94)
Base	log10	Hospitalizations reduced (million)	5	50%	2.99 (1.88, 3.99)	-0.02 (-1.32, 1.05)	3.08 (1.94, 4.25)
Base	log10	Hospitalizations reduced (million)	5	80%	4.81 (3.26, 6.19)	0.00 (-1.91, 1.61)	4.91 (3.59, 6.36)
Base	log10	Hospitalizations reduced (million)	5	100%	5.97 (4.48, 7.74)	-0.06 (-2.14, 1.98)	6.07 (4.68, 7.78)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.2	20%	-0.23 (-3.29, 3.26)	9.68 (0.33, 23.16)	9.58 (0.49, 23.16)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.2	50%	-0.51 (-6.52, 3.53)	26.14 (11.83, 47.61)	25.91 (11.80, 47.61)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.2	80%	-0.80 (-6.66, 3.89)	43.60 (25.80, 71.83)	43.31 (25.77, 71.63)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.2	100%	-1.23 (-5.67, 2.50)	54.28 (34.96, 81.22)	54.10 (34.93, 81.19)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.5	20%	-0.33 (-3.10, 1.55)	4.77 (-0.43, 10.74)	4.73 (-0.43, 10.74)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.5	50%	-0.60 (-4.12, 2.37)	13.72 (5.83, 21.42)	13.61 (5.77, 22.08)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.5	80%	-1.09 (-5.34, 3.06)	23.27 (14.04, 32.75)	23.18 (13.38, 32.62)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.5	100%	-1.19 (-5.21, 3.49)	30.51 (19.24, 46.92)	30.32 (18.81, 44.22)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.7	20%	-0.03 (-2.90, 2.27)	4.31 (-0.40, 10.48)	4.26 (-0.66, 10.44)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.7	50%	-0.18 (-3.49, 3.99)	10.84 (4.94, 17.76)	10.84 (4.32, 17.73)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.7	80%	-0.44 (-3.99, 3.95)	18.14 (10.38, 25.63)	17.86 (9.56, 25.21)
V1	log10-proportional	Number of cases averted in U.S. (million)	1.7	100%	-0.56 (-4.58, 4.28)	23.47 (14.99, 33.08)	23.25 (14.96, 32.68)
V1	log10-proportional	Number of cases averted in U.S. (million)	2	20%	-0.21 (-2.90, 2.01)	2.97 (-3.16, 9.13)	2.88 (-4.68, 8.04)



Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V1	log10-proportional	Number of cases averted in U.S. (million)	2	50%	-0.55 (-5.11, 2.24)	7.50 (1.94, 14.00)	7.48 (1.71, 13.97)
V1	log10-proportional	Number of cases averted in U.S. (million)	2	80%	-0.68 (-5.21, 3.23)	12.97 (6.03, 20.53)	12.82 (5.60, 20.53)
V1	log10-proportional	Number of cases averted in U.S. (million)	2	100%	-0.78 (-5.11, 4.15)	16.57 (8.80, 22.73)	16.42 (8.11, 22.50)
V1	log10-proportional	Number of cases averted in U.S. (million)	3	20%	-0.03 (-0.53, 0.63)	0.66 (-0.16, 1.52)	0.65 (-0.16, 1.52)
V1	log10-proportional	Number of cases averted in U.S. (million)	3	50%	-0.11 (-0.82, 0.63)	1.77 (0.63, 2.93)	1.73 (0.63, 3.13)
V1	log10-proportional	Number of cases averted in U.S. (million)	3	80%	-0.23 (-1.12, 0.63)	3.11 (1.94, 4.71)	3.07 (1.71, 4.58)
V1	log10-proportional	Number of cases averted in U.S. (million)	3	100%	-0.27 (-1.12, 0.49)	4.07 (2.77, 5.30)	3.94 (2.50, 5.47)
V1	log10-proportional	Number of cases averted in U.S. (million)	5	20%	0.00 (-0.10, 0.10)	0.04 (-0.10, 0.23)	0.04 (-0.10, 0.23)
V1	log10-proportional	Number of cases averted in U.S. (million)	5	50%	0.00 (-0.13, 0.13)	0.10 (-0.07, 0.30)	0.09 (-0.07, 0.30)
V1	log10-proportional	Number of cases averted in U.S. (million)	5	80%	-0.01 (-0.16, 0.13)	0.15 (-0.07, 0.36)	0.14 (-0.10, 0.36)
V1	log10-proportional	Number of cases averted in U.S. (million)	5	100%	-0.01 (-0.13, 0.16)	0.19 (-0.03, 0.53)	0.17 (-0.03, 0.46)
V1	log10-proportional	Deaths reduced (thousand)	1.2	20%	16.15 (-7.13, 46.43)	19.76 (-33.68, 71.44)	34.03 (-14.27, 81.90)
V1	log10-proportional	Deaths reduced (thousand)	1.2	50%	44.11 (-3.92, 89.28)	51.05 (-26.43, 138.79)	82.35 (7.31, 163.00)
V1	log10-proportional	Deaths reduced (thousand)	1.2	80%	67.64 (8.59, 138.97)	81.04 (-0.60, 160.94)	119.05 (40.86, 189.58)
V1	log10-proportional	Deaths reduced (thousand)	1.2	100%	83.65 (16.44, 151.47)	100.01 (20.42, 171.24)	139.10 (73.01, 212.84)
V1	log10-proportional	Deaths reduced (thousand)	1.5	20%	33.63 (-0.12, 78.28)	12.81 (-42.80, 71.09)	47.99 (6.95, 101.49)
V1	log10-proportional	Deaths reduced (thousand)	1.5	50%	90.69 (36.36, 170.90)	39.94 (-40.40, 130.11)	121.03 (47.64, 208.50)
V1	log10-proportional	Deaths reduced (thousand)	1.5	80%	144.10 (49.36, 230.93)	59.86 (-33.14, 152.94)	180.95 (90.04, 265.45)
V1	log10-proportional	Deaths reduced (thousand)	1.5	100%	182.30 (86.35, 270.41)	79.50 (-26.56, 196.44)	218.48 (123.64, 336.94)
V1	log10-proportional	Deaths reduced (thousand)	1.7	20%	46.15 (-5.03, 89.22)	11.64 (-57.23, 85.24)	57.36 (2.24, 112.79)
V1	log10-proportional	Deaths reduced (thousand)	1.7	50%	114.48 (27.48, 185.66)	31.45 (-43.26, 138.69)	141.60 (70.57, 236.61)
V1	log10-proportional	Deaths reduced (thousand)	1.7	80%	183.45 (76.63, 278.15)	55.14 (-38.76, 176.52)	217.61 (122.43, 320.85)
V1	log10-proportional	Deaths reduced (thousand)	1.7	100%	229.49 (124.71, 338.22)	69.92 (-52.10, 187.19)	263.30 (146.93, 365.55)
V1	log10-proportional	Deaths reduced (thousand)	2	20%	58.54 (9.12, 112.42)	11.68 (-58.97, 88.91)	69.51 (-5.67, 153.17)
V1	log10-proportional	Deaths reduced (thousand)	2	50%	147.20 (59.10, 222.91)	25.51 (-79.17, 119.34)	165.22 (69.98, 249.83)
V1	log10-proportional	Deaths reduced (thousand)	2	80%	235.31 (142.64, 353.27)	50.51 (-68.60, 156.17)	265.32 (167.86, 383.74)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V1	log10-proportional	Deaths reduced (thousand)	2	100%	291.76 (176.58, 422.76)	63.44 (-74.48, 184.51)	324.68 (203.59, 456.88)
V1	log10-proportional	Deaths reduced (thousand)	3	20%	102.22 (33.69, 176.18)	14.81 (-88.66, 110.68)	111.41 (39.11, 199.14)
V1	log10-proportional	Deaths reduced (thousand)	3	50%	252.10 (156.92, 352.38)	24.51 (-104.53, 150.04)	267.64 (160.55, 374.07)
V1	log10-proportional	Deaths reduced (thousand)	3	80%	399.96 (250.34, 527.24)	34.29 (-153.84, 174.53)	418.93 (271.57, 551.23)
V1	log10-proportional	Deaths reduced (thousand)	3	100%	498.05 (369.00, 636.06)	41.63 (-133.15, 202.59)	516.84 (399.85, 663.90)
V1	log10-proportional	Deaths reduced (thousand)	5	20%	143.09 (44.85, 221.24)	1.03 (-102.17, 106.06)	151.24 (28.65, 246.44)
V1	log10-proportional	Deaths reduced (thousand)	5	50%	363.16 (205.05, 492.13)	-1.54 (-176.25, 185.46)	379.90 (240.41, 509.77)
V1	log10-proportional	Deaths reduced (thousand)	5	80%	591.48 (414.50, 772.95)	-0.61 (-259.31, 225.62)	607.46 (410.58, 780.11)
V1	log10-proportional	Deaths reduced (thousand)	5	100%	744.21 (555.56, 948.46)	2.30 (-204.90, 238.01)	762.27 (578.26, 948.34)
V1	log10-proportional	NMB (\$ billion)	1.2	20%	26.38 (-17.97, 77.89)	32.56 (-53.16, 132.75)	58.42 (-22.99, 147.37)
V1	log10-proportional	NMB (\$ billion)	1.2	50%	71.10 (-21.67, 144.25)	85.19 (-58.30, 226.41)	141.49 (11.22, 272.68)
V1	log10-proportional	NMB (\$ billion)	1.2	80%	108.00 (-5.39, 211.17)	137.99 (-16.62, 268.51)	206.76 (68.35, 340.23)
V1	log10-proportional	NMB (\$ billion)	1.2	100%	133.23 (11.65, 249.42)	172.64 (34.64, 295.28)	244.01 (123.06, 388.28)
V1	log10-proportional	NMB (\$ billion)	1.5	20%	54.78 (-10.38, 134.97)	18.11 (-78.60, 116.73)	79.85 (-2.13, 175.76)
V1	log10-proportional	NMB (\$ billion)	1.5	50%	145.43 (54.07, 266.83)	56.00 (-94.11, 208.78)	198.98 (70.70, 341.65)
V1	log10-proportional	NMB (\$ billion)	1.5	80%	232.33 (88.94, 391.27)	86.58 (-63.72, 256.86)	299.60 (163.65, 448.54)
V1	log10-proportional	NMB (\$ billion)	1.5	100%	293.67 (140.23, 451.41)	116.19 (-61.40, 324.23)	361.50 (213.47, 546.30)
V1	log10-proportional	NMB (\$ billion)	1.7	20%	74.18 (-6.97, 149.06)	12.39 (-115.95, 143.27)	92.47 (-8.09, 189.80)
V1	log10-proportional	NMB (\$ billion)	1.7	50%	183.32 (49.23, 333.69)	36.96 (-107.80, 223.19)	229.69 (112.26, 409.15)
V1	log10-proportional	NMB (\$ billion)	1.7	80%	293.23 (118.15, 476.10)	68.72 (-100.37, 270.10)	352.17 (187.33, 534.45)
V1	log10-proportional	NMB (\$ billion)	1.7	100%	366.16 (178.21, 548.88)	86.87 (-129.16, 307.14)	426.63 (235.62, 631.03)
V1	log10-proportional	NMB (\$ billion)	2	20%	92.71 (1.97, 189.01)	11.32 (-130.04, 134.33)	111.16 (-23.52, 239.59)
V1	log10-proportional	NMB (\$ billion)	2	50%	235.13 (84.42, 361.11)	24.31 (-148.18, 191.67)	265.87 (116.58, 411.46)
V1	log10-proportional	NMB (\$ billion)	2	80%	375.17 (217.96, 572.52)	56.23 (-155.61, 219.98)	427.06 (264.40, 645.71)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V1	log10-proportional	NMB (\$ billion)	2	100%	463.52 (275.26, 697.98)	69.58 (-172.78, 280.70)	521.94 (320.73, 766.65)
V1	log10-proportional	NMB (\$ billion)	3	20%	158.56 (37.55, 261.16)	9.71 (-164.07, 176.62)	173.43 (54.85, 315.51)
V1	log10-proportional	NMB (\$ billion)	3	50%	394.79 (228.36, 542.75)	12.37 (-207.26, 203.07)	421.14 (246.16, 592.96)
V1	log10-proportional	NMB (\$ billion)	3	80%	625.68 (373.94, 844.12)	12.41 (-292.61, 228.89)	659.71 (409.27, 872.60)
V1	log10-proportional	NMB (\$ billion)	3	100%	777.18 (546.23, 994.81)	12.37 (-265.89, 309.20)	812.41 (620.29, 1015.68)
V1	log10-proportional	NMB (\$ billion)	5	20%	217.41 (59.78, 366.05)	-22.84 (-235.77, 151.93)	232.36 (36.11, 383.17)
V1	log10-proportional	NMB (\$ billion)	5	50%	557.85 (316.19, 751.54)	-55.98 (-363.93, 206.17)	586.89 (336.42, 818.34)
V1	log10-proportional	NMB (\$ billion)	5	80%	912.76 (627.53, 1197.26)	-81.56 (-497.01, 293.76)	942.06 (623.17, 1215.65)
V1	log10-proportional	NMB (\$ billion)	5	100%	1147.82 (789.06, 1496.34)	-97.63 (-447.29, 279.14)	1181.88 (858.51, 1509.11)
V1	log10-proportional	Treatment courses used (million)	1.2	20%	6.54 (5.21, 7.91)	5.93 (4.35, 7.15)	5.94 (4.35, 7.15)
V1	log10-proportional	Treatment courses used (million)	1.2	50%	16.47 (13.38, 19.01)	12.36 (8.86, 15.49)	12.39 (8.86, 15.49)
V1	log10-proportional	Treatment courses used (million)	1.2	80%	26.37 (21.61, 29.95)	15.13 (7.55, 20.56)	15.20 (7.55, 20.69)
V1	log10-proportional	Treatment courses used (million)	1.2	100%	33.17 (27.45, 37.63)	15.41 (6.69, 21.58)	15.47 (6.72, 21.65)
V1	log10-proportional	Treatment courses used (million)	1.5	20%	11.27 (9.88, 12.65)	10.92 (9.75, 12.36)	10.92 (9.79, 12.36)
V1	log10-proportional	Treatment courses used (million)	1.5	50%	28.15 (25.77, 30.71)	25.97 (23.76, 28.43)	25.98 (23.76, 28.43)
V1	log10-proportional	Treatment courses used (million)	1.5	80%	45.11 (42.40, 48.93)	39.19 (36.54, 42.57)	39.22 (36.38, 42.60)
V1	log10-proportional	Treatment courses used (million)	1.5	100%	56.46 (53.08, 60.56)	46.80 (43.00, 50.58)	46.88 (43.03, 50.58)
V1	log10-proportional	Treatment courses used (million)	1.7	20%	13.33 (11.83, 14.63)	13.06 (11.80, 14.50)	13.06 (11.80, 14.56)
V1	log10-proportional	Treatment courses used (million)	1.7	50%	33.29 (29.88, 35.58)	31.62 (29.09, 34.33)	31.61 (29.16, 34.10)
V1	log10-proportional	Treatment courses used (million)	1.7	80%	53.37 (49.62, 56.64)	48.76 (45.57, 51.73)	48.83 (45.73, 51.89)
V1	log10-proportional	Treatment courses used (million)	1.7	100%	66.72 (63.13, 70.11)	59.43 (55.88, 63.36)	59.50 (55.81, 63.39)
V1	log10-proportional	Treatment courses used (million)	2	20%	15.74 (14.23, 17.27)	15.53 (14.17, 17.00)	15.54 (14.17, 17.17)
V1	log10-proportional	Treatment courses used (million)	2	50%	39.47 (36.77, 43.13)	38.05 (35.45, 40.69)	38.04 (35.42, 40.79)
V1	log10-proportional	Treatment courses used (million)	2	80%	63.19 (58.81, 67.18)	59.51 (56.47, 62.47)	59.56 (56.37, 62.47)
V1	log10-proportional	Treatment courses used (million)	2	100%	78.97 (74.00, 84.32)	73.24 (69.36, 77.36)	73.29 (69.72, 77.13)
V1	log10-proportional	Treatment courses used (million)	3	20%	24.12 (22.21, 25.73)	23.90 (22.14, 25.67)	23.91 (22.21, 25.67)
V1	log10-proportional	Treatment courses used (million)	3	50%	60.41 (57.53, 63.56)	59.31 (56.34, 62.31)	59.33 (56.34, 62.37)
V1	log10-proportional	Treatment courses used (million)	3	80%	96.70 (92.78, 100.00)	93.67 (90.51, 96.74)	93.71 (90.58, 96.84)
V1	log10-proportional	Treatment courses used (million)	3	100%	121.11 (117.43, 125.24)	116.26 (112.88, 120.33)	116.41 (112.91, 120.43)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V1	log10-proportional	Treatment courses used (million)	5	20%	37.34 (35.06, 40.33)	37.15 (35.16, 39.83)	37.19 (35.32, 39.90)
V1	log10-proportional	Treatment courses used (million)	5	50%	93.34 (89.85, 97.10)	92.18 (88.73, 96.08)	92.29 (88.73, 96.08)
V1	log10-proportional	Treatment courses used (million)	5	80%	149.39 (145.50, 153.57)	146.60 (142.83, 150.25)	146.78 (143.36, 150.48)
V1	log10-proportional	Treatment courses used (million)	5	100%	186.99 (182.30, 191.86)	182.61 (177.86, 186.36)	182.85 (177.89, 186.82)
V1	log10-proportional	Hospitalizations reduced (million)	1.2	20%	0.14 (-0.07, 0.40)	0.16 (-0.26, 0.63)	0.29 (-0.10, 0.69)
V1	log10-proportional	Hospitalizations reduced (million)	1.2	50%	0.38 (-0.07, 0.76)	0.43 (-0.30, 1.05)	0.69 (0.10, 1.28)
V1	log10-proportional	Hospitalizations reduced (million)	1.2	80%	0.58 (0.00, 1.05)	0.68 (-0.03, 1.28)	1.00 (0.33, 1.61)
V1	log10-proportional	Hospitalizations reduced (million)	1.2	100%	0.71 (0.16, 1.28)	0.84 (0.20, 1.38)	1.18 (0.63, 1.88)
V1	log10-proportional	Hospitalizations reduced (million)	1.5	20%	0.29 (0.00, 0.63)	0.11 (-0.33, 0.56)	0.40 (0.00, 0.86)
V1	log10-proportional	Hospitalizations reduced (million)	1.5	50%	0.75 (0.33, 1.32)	0.32 (-0.36, 1.02)	0.99 (0.36, 1.65)
V1	log10-proportional	Hospitalizations reduced (million)	1.5	80%	1.20 (0.59, 1.91)	0.50 (-0.20, 1.32)	1.50 (0.86, 2.21)
V1	log10-proportional	Hospitalizations reduced (million)	1.5	100%	1.51 (0.82, 2.31)	0.66 (-0.20, 1.65)	1.80 (1.09, 2.64)
V1	log10-proportional	Hospitalizations reduced (million)	1.7	20%	0.38 (0.00, 0.72)	0.09 (-0.49, 0.69)	0.46 (0.00, 0.96)
V1	log10-proportional	Hospitalizations reduced (million)	1.7	50%	0.94 (0.30, 1.61)	0.25 (-0.46, 1.09)	1.14 (0.59, 2.04)
V1	log10-proportional	Hospitalizations reduced (million)	1.7	80%	1.50 (0.66, 2.31)	0.43 (-0.40, 1.38)	1.75 (1.05, 2.57)
V1	log10-proportional	Hospitalizations reduced (million)	1.7	100%	1.87 (0.99, 2.77)	0.54 (-0.49, 1.52)	2.12 (1.25, 3.03)
V1	log10-proportional	Hospitalizations reduced (million)	2	20%	0.47 (0.03, 0.89)	0.09 (-0.56, 0.66)	0.55 (-0.03, 1.15)
V1	log10-proportional	Hospitalizations reduced (million)	2	50%	1.19 (0.49, 1.78)	0.21 (-0.56, 0.96)	1.33 (0.66, 2.01)
V1	log10-proportional	Hospitalizations reduced (million)	2	80%	1.90 (1.15, 2.77)	0.41 (-0.63, 1.19)	2.13 (1.38, 3.16)
V1	log10-proportional	Hospitalizations reduced (million)	2	100%	2.35 (1.48, 3.43)	0.49 (-0.59, 1.52)	2.60 (1.65, 3.76)
V1	log10-proportional	Hospitalizations reduced (million)	3	20%	0.79 (0.23, 1.25)	0.10 (-0.69, 0.89)	0.86 (0.30, 1.48)
V1	log10-proportional	Hospitalizations reduced (million)	3	50%	1.98 (1.22, 2.70)	0.21 (-0.86, 1.09)	2.09 (1.28, 2.87)
V1	log10-proportional	Hospitalizations reduced (million)	3	80%	3.13 (1.98, 4.09)	0.29 (-1.12, 1.28)	3.28 (2.14, 4.25)
V1	log10-proportional	Hospitalizations reduced (million)	3	100%	3.89 (2.83, 4.88)	0.35 (-0.92, 1.65)	4.04 (3.10, 5.04)
V1	log10-proportional	Hospitalizations reduced (million)	5	20%	1.09 (0.40, 1.78)	-0.02 (-1.02, 0.72)	1.16 (0.26, 1.88)
V1	log10-proportional	Hospitalizations reduced (million)	5	50%	2.81 (1.65, 3.76)	-0.04 (-1.42, 1.22)	2.94 (1.85, 3.99)
V1	log10-proportional	Hospitalizations reduced (million)	5	80%	4.58 (3.23, 5.90)	-0.03 (-1.91, 1.61)	4.71 (3.26, 5.96)
V1	log10-proportional	Hospitalizations reduced (million)	5	100%	5.76 (4.18, 7.35)	-0.01 (-1.55, 1.68)	5.91 (4.45, 7.41)
V2	threshold	Number of cases averted in U.S. (million)	1.2	20%	-0.30 (-3.06, 1.94)	11.37 (3.66, 26.39)	11.27 (3.66, 26.36)
V2	threshold	Number of cases averted in U.S. (million)	1.2	50%	-0.70 (-5.50, 2.47)	29.90 (17.30, 45.14)	29.48 (17.23, 45.11)
V2	threshold	Number of cases averted in U.S. (million)	1.2	80%	-1.00 (-6.66, 2.60)	47.54 (33.15, 69.62)	47.10 (32.62, 69.62)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V2	threshold	Number of cases averted in U.S. (million)	1.2	100%	-1.31 (-6.66, 3.26)	59.91 (43.89, 81.05)	59.33 (43.13, 78.29)
V2	threshold	Number of cases averted in U.S. (million)	1.5	20%	-0.18 (-2.04, 1.65)	5.84 (0.96, 10.48)	5.73 (0.72, 10.44)
V2	threshold	Number of cases averted in U.S. (million)	1.5	50%	-0.40 (-2.73, 1.78)	16.29 (10.94, 23.99)	16.11 (10.94, 24.22)
V2	threshold	Number of cases averted in U.S. (million)	1.5	80%	-0.63 (-3.03, 2.08)	28.52 (20.82, 36.84)	28.14 (20.46, 36.14)
V2	threshold	Number of cases averted in U.S. (million)	1.5	100%	-0.74 (-3.03, 2.08)	38.71 (30.31, 48.93)	38.28 (30.21, 49.29)
V2	threshold	Number of cases averted in U.S. (million)	1.7	20%	0.02 (-1.09, 1.35)	4.87 (0.53, 8.50)	4.76 (0.49, 8.50)
V2	threshold	Number of cases averted in U.S. (million)	1.7	50%	-0.27 (-2.67, 1.85)	12.75 (8.57, 18.65)	12.44 (7.58, 18.48)
V2	threshold	Number of cases averted in U.S. (million)	1.7	80%	-0.51 (-2.73, 1.75)	21.96 (16.31, 27.15)	21.59 (15.85, 27.05)
V2	threshold	Number of cases averted in U.S. (million)	1.7	100%	-0.69 (-3.06, 1.61)	28.51 (21.55, 34.66)	28.16 (20.56, 34.40)
V2	threshold	Number of cases averted in U.S. (million)	2	20%	-0.14 (-1.75, 1.22)	3.25 (0.20, 6.23)	3.15 (-0.13, 6.00)
V2	threshold	Number of cases averted in U.S. (million)	2	50%	-0.39 (-2.08, 1.68)	8.25 (4.61, 12.32)	8.09 (4.35, 12.03)
V2	threshold	Number of cases averted in U.S. (million)	2	80%	-0.62 (-2.97, 1.55)	14.28 (9.39, 18.58)	13.99 (9.36, 17.89)
V2	threshold	Number of cases averted in U.S. (million)	2	100%	-0.76 (-3.79, 2.77)	18.89 (13.34, 24.38)	18.56 (12.72, 23.66)
V2	threshold	Number of cases averted in U.S. (million)	3	20%	-0.03 (-0.40, 0.23)	0.77 (0.16, 1.45)	0.75 (0.10, 1.35)
V2	threshold	Number of cases averted in U.S. (million)	3	50%	-0.11 (-0.53, 0.26)	2.02 (1.15, 2.97)	1.94 (1.12, 2.87)
V2	threshold	Number of cases averted in U.S. (million)	3	80%	-0.17 (-0.72, 0.43)	3.39 (2.08, 4.45)	3.29 (2.14, 4.45)
V2	threshold	Number of cases averted in U.S. (million)	3	100%	-0.22 (-0.79, 0.30)	4.37 (3.10, 5.67)	4.23 (2.87, 5.63)
V2	threshold	Number of cases averted in U.S. (million)	5	20%	0.00 (-0.07, 0.07)	0.03 (-0.07, 0.13)	0.03 (-0.07, 0.13)
V2	threshold	Number of cases averted in U.S. (million)	5	50%	0.00 (-0.10, 0.07)	0.09 (-0.07, 0.23)	0.08 (-0.07, 0.23)
V2	threshold	Number of cases averted in U.S. (million)	5	80%	-0.01 (-0.16, 0.13)	0.14 (-0.07, 0.36)	0.12 (-0.07, 0.33)
V2	threshold	Number of cases averted in U.S. (million)	5	100%	-0.02 (-0.16, 0.10)	0.17 (-0.03, 0.40)	0.16 (-0.07, 0.40)
V2	threshold	Deaths reduced (thousand)	1.2	20%	17.15 (-5.44, 39.23)	23.64 (-19.82, 67.70)	38.07 (-1.81, 78.46)
V2	threshold	Deaths reduced (thousand)	1.2	50%	43.42 (-1.53, 82.14)	59.43 (0.59, 119.73)	88.79 (32.68, 139.21)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V2	threshold	Deaths reduced (thousand)	1.2	80%	69.08 (14.55, 125.17)	91.36 (21.40, 153.53)	125.88 (60.56, 182.18)
V2	threshold	Deaths reduced (thousand)	1.2	100%	85.50 (28.36, 141.43)	112.59 (39.52, 178.55)	145.22 (82.37, 205.39)
V2	threshold	Deaths reduced (thousand)	1.5	20%	38.90 (-3.80, 89.05)	20.64 (-39.52, 74.91)	55.65 (9.18, 108.35)
V2	threshold	Deaths reduced (thousand)	1.5	50%	97.29 (34.09, 173.00)	45.62 (-40.74, 129.39)	132.70 (69.69, 219.13)
V2	threshold	Deaths reduced (thousand)	1.5	80%	154.81 (71.13, 245.44)	79.46 (-46.08, 176.96)	200.81 (97.89, 285.56)
V2	threshold	Deaths reduced (thousand)	1.5	100%	189.57 (96.72, 283.47)	105.21 (-7.20, 192.23)	243.83 (128.26, 332.42)
V2	threshold	Deaths reduced (thousand)	1.7	20%	48.21 (-0.12, 93.07)	12.45 (-51.97, 65.31)	57.75 (-6.85, 114.48)
V2	threshold	Deaths reduced (thousand)	1.7	50%	119.90 (30.23, 204.81)	34.10 (-80.21, 115.88)	143.57 (40.77, 231.89)
V2	threshold	Deaths reduced (thousand)	1.7	80%	192.31 (103.54, 292.20)	66.60 (-53.11, 158.34)	228.64 (125.36, 318.28)
V2	threshold	Deaths reduced (thousand)	1.7	100%	236.97 (149.69, 321.74)	84.82 (-29.04, 191.11)	278.39 (186.82, 362.48)
V2	threshold	Deaths reduced (thousand)	2	20%	60.91 (1.93, 115.54)	11.41 (-80.14, 81.83)	71.67 (-8.76, 140.83)
V2	threshold	Deaths reduced (thousand)	2	50%	150.46 (46.38, 231.93)	28.97 (-113.92, 112.25)	174.79 (44.59, 261.64)
V2	threshold	Deaths reduced (thousand)	2	80%	241.23 (117.19, 325.13)	54.42 (-91.38, 161.98)	275.81 (152.71, 365.84)
V2	threshold	Deaths reduced (thousand)	2	100%	304.10 (171.89, 402.99)	72.76 (-66.30, 176.92)	342.22 (201.93, 444.01)
V2	threshold	Deaths reduced (thousand)	3	20%	100.07 (32.03, 174.68)	10.84 (-99.57, 119.45)	111.08 (14.27, 205.09)
V2	threshold	Deaths reduced (thousand)	3	50%	249.78 (133.48, 354.23)	23.75 (-100.42, 187.99)	272.99 (161.82, 389.79)
V2	threshold	Deaths reduced (thousand)	3	80%	400.58 (282.98, 533.51)	37.29 (-107.29, 254.93)	427.53 (307.19, 580.04)
V2	threshold	Deaths reduced (thousand)	3	100%	503.61 (361.84, 667.05)	52.91 (-96.46, 262.92)	531.16 (388.16, 675.35)
V2	threshold	Deaths reduced (thousand)	5	20%	154.99 (67.61, 247.44)	-2.63 (-108.52, 97.96)	156.35 (44.86, 243.04)
V2	threshold	Deaths reduced (thousand)	5	50%	382.55 (239.73, 524.66)	-3.43 (-171.40, 144.47)	387.41 (246.14, 530.07)
V2	threshold	Deaths reduced (thousand)	5	80%	609.96 (449.45, 814.46)	5.58 (-194.93, 261.04)	623.02 (465.10, 822.44)
V2	threshold	Deaths reduced (thousand)	5	100%	753.47 (585.00, 982.33)	-0.19 (-219.71, 291.34)	772.75 (604.30, 995.18)
V2	threshold	NMB (\$ billion)	1.2	20%	27.40 (-16.31, 68.82)	39.29 (-41.89, 112.38)	65.31 (0.71, 141.95)
V2	threshold	NMB (\$ billion)	1.2	50%	68.75 (1.74, 141.86)	100.42 (-4.40, 200.97)	153.23 (55.32, 240.08)
V2	threshold	NMB (\$ billion)	1.2	80%	110.58 (23.88, 203.74)	157.30 (29.39, 261.71)	219.81 (102.90, 308.62)
V2	threshold	NMB (\$ billion)	1.2	100%	135.48 (32.55, 244.15)	195.35 (71.31, 299.55)	254.70 (151.88, 358.54)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V2	threshold	NMB (\$ billion)	1.5	20%	62.02 (-24.68, 135.07)	28.64 (-68.19, 123.77)	90.37 (12.18, 177.46)
V2	threshold	NMB (\$ billion)	1.5	50%	154.30 (39.73, 283.09)	63.29 (-82.45, 206.15)	216.02 (106.80, 373.11)
V2	threshold	NMB (\$ billion)	1.5	80%	246.14 (94.47, 398.27)	116.48 (-98.95, 294.35)	329.44 (149.48, 505.84)
V2	threshold	NMB (\$ billion)	1.5	100%	300.88 (117.18, 463.67)	156.02 (-35.21, 324.45)	401.00 (209.15, 563.02)
V2	threshold	NMB (\$ billion)	1.7	20%	77.10 (-12.75, 163.34)	13.47 (-92.40, 97.95)	92.79 (-15.12, 191.20)
V2	threshold	NMB (\$ billion)	1.7	50%	189.32 (24.86, 338.37)	39.88 (-153.00, 193.40)	230.76 (45.40, 401.40)
V2	threshold	NMB (\$ billion)	1.7	80%	305.03 (136.52, 475.63)	87.23 (-107.73, 246.16)	369.74 (197.26, 543.65)
V2	threshold	NMB (\$ billion)	1.7	100%	375.34 (218.51, 537.96)	110.57 (-92.29, 295.91)	449.87 (260.43, 602.46)
V2	threshold	NMB (\$ billion)	2	20%	96.51 (-3.59, 199.44)	12.27 (-144.80, 149.40)	115.82 (-12.85, 241.25)
V2	threshold	NMB (\$ billion)	2	50%	237.29 (38.48, 361.34)	27.86 (-224.24, 175.03)	278.55 (75.74, 417.55)
V2	threshold	NMB (\$ billion)	2	80%	380.58 (189.93, 538.44)	59.57 (-195.98, 243.22)	441.14 (214.89, 603.82)
V2	threshold	NMB (\$ billion)	2	100%	480.20 (251.39, 649.18)	82.60 (-163.77, 274.87)	547.65 (332.82, 702.43)
V2	threshold	NMB (\$ billion)	3	20%	155.09 (48.09, 287.88)	4.20 (-164.87, 186.12)	173.53 (29.24, 323.30)
V2	threshold	NMB (\$ billion)	3	50%	389.94 (183.64, 567.09)	10.28 (-247.85, 271.98)	430.46 (242.14, 621.07)
V2	threshold	NMB (\$ billion)	3	80%	627.47 (413.89, 865.42)	16.61 (-257.12, 336.24)	673.68 (468.40, 958.89)
V2	threshold	NMB (\$ billion)	3	100%	787.30 (556.13, 1068.43)	29.82 (-251.45, 364.75)	836.36 (601.25, 1070.03)
V2	threshold	NMB (\$ billion)	5	20%	237.27 (79.70, 389.26)	-28.97 (-194.99, 122.52)	240.32 (67.46, 384.58)
V2	threshold	NMB (\$ billion)	5	50%	589.59 (366.06, 828.63)	-56.53 (-363.41, 178.08)	599.76 (377.93, 858.01)
V2	threshold	NMB (\$ billion)	5	80%	945.02 (670.93, 1275.73)	-69.84 (-402.22, 308.64)	968.79 (728.58, 1313.21)
V2	threshold	NMB (\$ billion)	5	100%	1165.59 (892.37, 1477.99)	-98.65 (-442.35, 355.92)	1200.79 (904.45, 1531.25)
V2	threshold	Treatment courses used (million)	1.2	20%	6.54 (5.27, 7.74)	5.86 (4.28, 7.51)	5.86 (4.28, 7.51)
V2	threshold	Treatment courses used (million)	1.2	50%	16.54 (12.92, 19.21)	11.70 (7.68, 14.20)	11.78 (8.04, 14.20)
V2	threshold	Treatment courses used (million)	1.2	80%	26.53 (20.82, 30.38)	14.26 (8.27, 18.35)	14.39 (8.27, 18.48)
V2	threshold	Treatment courses used (million)	1.2	100%	33.34 (26.36, 38.02)	13.92 (7.15, 19.67)	14.12 (7.28, 19.90)
V2	threshold	Treatment courses used (million)	1.5	20%	11.85 (10.68, 13.25)	11.44 (10.31, 12.75)	11.45 (10.31, 12.75)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V2	threshold	Treatment courses used (million)	1.5	50%	29.60 (27.45, 32.52)	26.97 (24.91, 29.39)	27.01 (24.91, 29.65)
V2	threshold	Treatment courses used (million)	1.5	80%	47.43 (44.45, 50.94)	40.16 (36.44, 43.62)	40.27 (36.84, 43.72)
V2	threshold	Treatment courses used (million)	1.5	100%	59.45 (55.98, 63.39)	47.02 (42.44, 50.71)	47.17 (42.50, 51.10)
V2	threshold	Treatment courses used (million)	1.7	20%	13.92 (12.55, 15.45)	13.62 (12.09, 14.70)	13.63 (12.09, 14.70)
V2	threshold	Treatment courses used (million)	1.7	50%	34.81 (32.65, 37.66)	32.74 (30.48, 35.39)	32.80 (30.54, 35.39)
V2	threshold	Treatment courses used (million)	1.7	80%	55.73 (52.62, 58.88)	49.92 (46.92, 52.92)	50.02 (47.22, 53.01)
V2	threshold	Treatment courses used (million)	1.7	100%	69.79 (66.16, 73.24)	60.34 (57.30, 64.02)	60.40 (57.43, 64.22)
V2	threshold	Treatment courses used (million)	2	20%	16.45 (14.99, 18.42)	16.18 (14.86, 18.02)	16.19 (14.89, 18.02)
V2	threshold	Treatment courses used (million)	2	50%	41.30 (38.78, 43.95)	39.74 (37.53, 41.81)	39.76 (37.56, 41.94)
V2	threshold	Treatment courses used (million)	2	80%	66.08 (62.67, 69.13)	61.92 (58.58, 65.04)	62.01 (58.58, 65.14)
V2	threshold	Treatment courses used (million)	2	100%	82.73 (79.24, 87.15)	75.77 (72.26, 78.88)	75.87 (72.42, 79.14)
V2	threshold	Treatment courses used (million)	3	20%	24.45 (22.57, 26.85)	24.26 (22.40, 26.59)	24.25 (22.40, 26.72)
V2	threshold	Treatment courses used (million)	3	50%	61.22 (58.62, 64.05)	59.91 (56.97, 62.93)	59.95 (57.23, 62.90)
V2	threshold	Treatment courses used (million)	3	80%	97.90 (94.33, 100.92)	94.65 (91.40, 97.63)	94.77 (91.53, 97.99)
V2	threshold	Treatment courses used (million)	3	100%	122.75 (119.01, 126.23)	117.38 (113.87, 121.35)	117.55 (114.27, 121.42)
V2	threshold	Treatment courses used (million)	5	20%	37.66 (35.49, 39.90)	37.42 (35.29, 39.54)	37.45 (35.35, 39.54)
V2	threshold	Treatment courses used (million)	5	50%	94.27 (89.69, 98.19)	93.21 (88.80, 97.36)	93.29 (88.80, 97.43)
V2	threshold	Treatment courses used (million)	5	80%	150.84 (145.76, 155.22)	148.10 (142.96, 152.42)	148.31 (143.16, 152.49)
V2	threshold	Treatment courses used (million)	5	100%	188.93 (184.08, 193.51)	184.47 (180.10, 189.26)	184.75 (180.43, 189.32)
V2	threshold	Hospitalizations reduced (million)	1.2	20%	0.15 (-0.07, 0.36)	0.20 (-0.20, 0.56)	0.32 (0.03, 0.69)
V2	threshold	Hospitalizations reduced (million)	1.2	50%	0.37 (0.07, 0.69)	0.49 (0.00, 0.96)	0.74 (0.30, 1.15)
V2	threshold	Hospitalizations reduced (million)	1.2	80%	0.59 (0.20, 0.99)	0.77 (0.16, 1.25)	1.06 (0.49, 1.48)
V2	threshold	Hospitalizations reduced (million)	1.2	100%	0.72 (0.26, 1.25)	0.94 (0.36, 1.45)	1.22 (0.76, 1.75)
V2	threshold	Hospitalizations reduced (million)	1.5	20%	0.32 (-0.10, 0.66)	0.16 (-0.26, 0.63)	0.45 (0.10, 0.86)
V2	threshold	Hospitalizations reduced (million)	1.5	50%	0.79 (0.26, 1.35)	0.35 (-0.30, 0.99)	1.07 (0.56, 1.81)
V2	threshold	Hospitalizations reduced (million)	1.5	80%	1.26 (0.59, 1.98)	0.63 (-0.36, 1.55)	1.63 (0.76, 2.44)
V2	threshold	Hospitalizations reduced (million)	1.5	100%	1.55 (0.69, 2.34)	0.83 (-0.07, 1.65)	1.97 (1.05, 2.77)
V2	threshold	Hospitalizations reduced (million)	1.7	20%	0.40 (-0.03, 0.79)	0.10 (-0.43, 0.49)	0.47 (-0.03, 0.92)
V2	threshold	Hospitalizations reduced (million)	1.7	50%	0.97 (0.20, 1.65)	0.26 (-0.59, 0.96)	1.15 (0.33, 1.91)
V2	threshold	Hospitalizations reduced (million)	1.7	80%	1.56 (0.79, 2.31)	0.52 (-0.33, 1.25)	1.84 (1.05, 2.67)
V2	threshold	Hospitalizations reduced (million)	1.7	100%	1.92 (1.19, 2.70)	0.65 (-0.33, 1.45)	2.23 (1.35, 2.97)
V2	threshold	Hospitalizations reduced (million)	2	20%	0.49 (0.03, 0.92)	0.10 (-0.59, 0.69)	0.58 (0.00, 1.19)



Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V2	threshold	Hospitalizations reduced (million)	2	50%	1.20 (0.30, 1.81)	0.23 (-0.89, 0.89)	1.39 (0.43, 2.01)
V2	threshold	Hospitalizations reduced (million)	2	80%	1.93 (0.96, 2.67)	0.43 (-0.76, 1.28)	2.20 (1.12, 2.97)
V2	threshold	Hospitalizations reduced (million)	2	100%	2.43 (1.38, 3.23)	0.56 (-0.59, 1.42)	2.72 (1.75, 3.46)
V2	threshold	Hospitalizations reduced (million)	3	20%	0.78 (0.30, 1.38)	0.08 (-0.69, 0.92)	0.86 (0.23, 1.58)
V2	threshold	Hospitalizations reduced (million)	3	50%	1.95 (0.99, 2.73)	0.20 (-1.02, 1.38)	2.14 (1.22, 3.06)
V2	threshold	Hospitalizations reduced (million)	3	80%	3.14 (2.11, 4.18)	0.31 (-0.96, 1.78)	3.34 (2.31, 4.65)
V2	threshold	Hospitalizations reduced (million)	3	100%	3.94 (2.83, 5.27)	0.42 (-0.89, 1.91)	4.15 (3.00, 5.21)
V2	threshold	Hospitalizations reduced (million)	5	20%	1.18 (0.43, 1.91)	-0.06 (-0.79, 0.59)	1.20 (0.46, 1.88)
V2	threshold	Hospitalizations reduced (million)	5	50%	2.95 (1.94, 4.12)	-0.04 (-1.45, 1.09)	3.00 (1.98, 4.25)
V2	threshold	Hospitalizations reduced (million)	5	80%	4.73 (3.53, 6.26)	0.03 (-1.45, 1.71)	4.84 (3.59, 6.39)
V2	threshold	Hospitalizations reduced (million)	5	100%	5.85 (4.61, 7.45)	-0.01 (-1.58, 2.01)	6.00 (4.61, 7.58)
V3	sigmoid	Number of cases averted in U.S. (million)	1.2	20%	-0.12 (-3.00, 2.31)	10.82 (1.35, 22.47)	10.77 (1.35, 21.58)
V3	sigmoid	Number of cases averted in U.S. (million)	1.2	50%	-0.38 (-4.12, 2.90)	31.47 (15.75, 47.81)	31.14 (15.75, 47.08)
V3	sigmoid	Number of cases averted in U.S. (million)	1.2	80%	-0.59 (-5.47, 3.66)	50.12 (31.00, 70.18)	49.81 (30.94, 70.18)
V3	sigmoid	Number of cases averted in U.S. (million)	1.2	100%	-0.84 (-5.60, 3.39)	61.94 (44.68, 81.42)	61.51 (41.61, 81.38)
V3	sigmoid	Number of cases averted in U.S. (million)	1.5	20%	-0.17 (-2.08, 2.04)	5.91 (0.16, 12.03)	5.82 (0.16, 12.13)
V3	sigmoid	Number of cases averted in U.S. (million)	1.5	50%	-0.45 (-3.92, 2.08)	16.33 (7.41, 24.97)	16.11 (8.60, 24.91)
V3	sigmoid	Number of cases averted in U.S. (million)	1.5	80%	-0.73 (-4.09, 2.67)	27.39 (19.41, 36.94)	27.01 (18.29, 36.14)
V3	sigmoid	Number of cases averted in U.S. (million)	1.5	100%	-0.93 (-4.22, 2.21)	37.09 (29.09, 47.35)	36.61 (27.58, 46.39)
V3	sigmoid	Number of cases averted in U.S. (million)	1.7	20%	-0.19 (-1.98, 1.61)	5.19 (0.23, 9.59)	5.12 (0.13, 9.42)
V3	sigmoid	Number of cases averted in U.S. (million)	1.7	50%	-0.34 (-2.50, 2.21)	12.52 (6.56, 19.27)	12.30 (6.49, 18.81)
V3	sigmoid	Number of cases averted in U.S. (million)	1.7	80%	-0.65 (-3.43, 2.14)	21.64 (15.58, 28.01)	21.49 (15.29, 27.74)
V3	sigmoid	Number of cases averted in U.S. (million)	1.7	100%	-0.93 (-4.28, 2.04)	28.69 (22.90, 36.61)	28.28 (22.34, 36.08)
V3	sigmoid	Number of cases averted in U.S. (million)	2	20%	-0.17 (-2.57, 1.78)	3.56 (0.23, 6.46)	3.48 (0.03, 6.42)
V3	sigmoid	Number of cases averted in U.S. (million)	2	50%	-0.45 (-2.83, 1.68)	8.61 (4.48, 13.15)	8.48 (2.93, 12.98)
V3	sigmoid	Number of cases averted in U.S. (million)	2	80%	-0.75 (-3.39, 2.17)	15.37 (10.08, 21.22)	15.12 (9.82, 20.72)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V3	sigmoid	Number of cases averted in U.S. (million)	2	100%	-0.88 (-3.39, 2.21)	20.10 (14.20, 26.52)	19.77 (13.64, 26.13)
V3	sigmoid	Number of cases averted in U.S. (million)	3	20%	-0.07 (-0.43, 0.26)	0.76 (-0.07, 1.55)	0.75 (-0.07, 1.71)
V3	sigmoid	Number of cases averted in U.S. (million)	3	50%	-0.13 (-0.56, 0.49)	2.06 (1.15, 3.13)	2.00 (1.05, 3.10)
V3	sigmoid	Number of cases averted in U.S. (million)	3	80%	-0.16 (-0.69, 0.46)	3.41 (2.04, 4.68)	3.31 (1.98, 4.58)
V3	sigmoid	Number of cases averted in U.S. (million)	3	100%	-0.22 (-0.79, 0.43)	4.54 (2.90, 5.80)	4.40 (2.77, 5.86)
V3	sigmoid	Number of cases averted in U.S. (million)	5	20%	0.00 (-0.07, 0.07)	0.03 (-0.10, 0.13)	0.03 (-0.10, 0.13)
V3	sigmoid	Number of cases averted in U.S. (million)	5	50%	-0.01 (-0.10, 0.10)	0.08 (-0.10, 0.26)	0.08 (-0.10, 0.26)
V3	sigmoid	Number of cases averted in U.S. (million)	5	80%	-0.01 (-0.10, 0.10)	0.15 (-0.03, 0.36)	0.13 (-0.03, 0.36)
V3	sigmoid	Number of cases averted in U.S. (million)	5	100%	-0.01 (-0.13, 0.13)	0.20 (-0.03, 0.43)	0.18 (0.00, 0.40)
V3	sigmoid	Deaths reduced (thousand)	1.2	20%	13.83 (-10.76, 37.77)	16.68 (-25.08, 57.35)	31.17 (-10.81, 74.90)
V3	sigmoid	Deaths reduced (thousand)	1.2	50%	38.52 (-3.62, 89.16)	53.82 (-5.08, 116.02)	81.77 (34.03, 148.24)
V3	sigmoid	Deaths reduced (thousand)	1.2	80%	60.34 (5.43, 108.93)	86.51 (7.13, 157.01)	117.50 (55.24, 175.39)
V3	sigmoid	Deaths reduced (thousand)	1.2	100%	75.94 (11.17, 133.66)	107.09 (29.00, 174.19)	136.52 (69.56, 196.79)
V3	sigmoid	Deaths reduced (thousand)	1.5	20%	39.17 (-0.12, 81.56)	18.94 (-30.17, 85.48)	54.96 (5.33, 110.45)
V3	sigmoid	Deaths reduced (thousand)	1.5	50%	98.13 (12.68, 181.36)	47.02 (-53.91, 150.73)	133.24 (39.94, 210.32)
V3	sigmoid	Deaths reduced (thousand)	1.5	80%	157.20 (53.38, 243.62)	68.36 (-27.07, 170.86)	196.67 (106.07, 293.55)
V3	sigmoid	Deaths reduced (thousand)	1.5	100%	191.94 (113.41, 274.38)	97.48 (-12.74, 213.41)	241.77 (166.98, 332.90)
V3	sigmoid	Deaths reduced (thousand)	1.7	20%	46.11 (-8.66, 94.07)	12.65 (-62.72, 74.95)	57.53 (-14.22, 132.01)
V3	sigmoid	Deaths reduced (thousand)	1.7	50%	113.71 (21.24, 201.09)	36.38 (-69.50, 139.28)	145.77 (44.69, 242.28)
V3	sigmoid	Deaths reduced (thousand)	1.7	80%	184.69 (78.06, 275.95)	60.59 (-52.71, 186.03)	223.39 (123.57, 335.84)
V3	sigmoid	Deaths reduced (thousand)	1.7	100%	231.30 (112.47, 348.62)	84.64 (-37.65, 212.72)	274.23 (165.21, 388.33)
V3	sigmoid	Deaths reduced (thousand)	2	20%	55.95 (-12.04, 112.31)	10.74 (-87.13, 80.44)	69.07 (-12.40, 144.24)
V3	sigmoid	Deaths reduced (thousand)	2	50%	145.65 (22.98, 253.46)	28.23 (-80.26, 142.57)	172.26 (55.27, 269.35)
V3	sigmoid	Deaths reduced (thousand)	2	80%	236.46 (78.41, 355.13)	55.79 (-58.47, 154.98)	271.54 (152.20, 367.46)
V3	sigmoid	Deaths reduced (thousand)	2	100%	297.70 (125.48, 429.83)	72.75 (-61.56, 184.70)	334.87 (189.30, 437.30)
V3	sigmoid	Deaths reduced (thousand)	3	20%	99.55 (37.83, 177.92)	11.38 (-88.92, 121.61)	108.96 (21.69, 200.82)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V3	sigmoid	Deaths reduced (thousand)	3	50%	253.66 (121.44, 374.01)	28.16 (-129.14, 135.85)	271.04 (149.96, 389.81)
V3	sigmoid	Deaths reduced (thousand)	3	80%	401.91 (276.73, 511.36)	41.95 (-143.92, 174.49)	425.44 (301.54, 559.92)
V3	sigmoid	Deaths reduced (thousand)	3	100%	504.77 (371.43, 640.87)	50.28 (-128.64, 185.57)	525.93 (395.51, 672.69)
V3	sigmoid	Deaths reduced (thousand)	5	20%	151.95 (63.42, 229.30)	0.90 (-99.91, 97.17)	156.10 (55.43, 233.34)
V3	sigmoid	Deaths reduced (thousand)	5	50%	377.19 (252.90, 510.31)	-1.17 (-157.28, 149.66)	383.18 (234.69, 513.12)
V3	sigmoid	Deaths reduced (thousand)	5	80%	597.26 (423.69, 769.79)	5.57 (-205.87, 226.22)	613.42 (449.77, 791.40)
V3	sigmoid	Deaths reduced (thousand)	5	100%	743.34 (578.10, 990.40)	1.95 (-219.56, 271.67)	760.42 (581.84, 998.20)
V3	sigmoid	NMB (\$ billion)	1.2	20%	21.52 (-23.43, 67.77)	27.51 (-49.79, 103.60)	53.78 (-24.03, 125.87)
V3	sigmoid	NMB (\$ billion)	1.2	50%	60.33 (-13.57, 136.17)	91.21 (-21.58, 210.19)	141.62 (47.05, 268.21)
V3	sigmoid	NMB (\$ billion)	1.2	80%	95.11 (-2.17, 171.22)	149.38 (23.41, 277.38)	205.69 (94.70, 307.31)
V3	sigmoid	NMB (\$ billion)	1.2	100%	119.68 (-0.80, 230.59)	187.07 (54.82, 300.88)	241.02 (125.68, 348.97)
V3	sigmoid	NMB (\$ billion)	1.5	20%	62.30 (-11.80, 136.89)	25.88 (-57.61, 129.27)	89.06 (6.69, 177.41)
V3	sigmoid	NMB (\$ billion)	1.5	50%	154.44 (11.93, 277.73)	63.05 (-113.31, 218.76)	215.29 (59.40, 337.65)
V3	sigmoid	NMB (\$ billion)	1.5	80%	248.48 (64.16, 379.89)	93.93 (-81.19, 235.40)	320.12 (172.36, 469.64)
V3	sigmoid	NMB (\$ billion)	1.5	100%	303.61 (136.79, 437.05)	141.46 (-63.87, 319.62)	396.05 (242.23, 544.14)
V3	sigmoid	NMB (\$ billion)	1.7	20%	73.64 (-10.68, 153.90)	13.57 (-133.00, 128.02)	92.01 (-20.11, 207.94)
V3	sigmoid	NMB (\$ billion)	1.7	50%	178.30 (30.45, 300.87)	40.38 (-161.57, 225.27)	231.93 (41.43, 396.19)
V3	sigmoid	NMB (\$ billion)	1.7	80%	291.57 (83.02, 451.51)	74.23 (-131.66, 283.05)	358.74 (172.34, 556.19)
V3	sigmoid	NMB (\$ billion)	1.7	100%	365.30 (172.03, 544.61)	111.40 (-107.46, 360.48)	443.11 (245.86, 624.27)
V3	sigmoid	NMB (\$ billion)	2	20%	87.12 (-16.38, 182.30)	9.39 (-138.67, 143.49)	110.38 (-23.31, 229.11)
V3	sigmoid	NMB (\$ billion)	2	50%	227.99 (21.34, 413.35)	25.92 (-172.69, 216.54)	273.66 (75.75, 444.19)
V3	sigmoid	NMB (\$ billion)	2	80%	373.70 (112.23, 580.56)	62.86 (-135.78, 245.16)	434.97 (236.08, 600.06)
V3	sigmoid	NMB (\$ billion)	2	100%	471.37 (177.97, 702.54)	84.40 (-179.88, 292.65)	537.26 (287.70, 727.58)
V3	sigmoid	NMB (\$ billion)	3	20%	154.17 (51.52, 273.51)	3.51 (-158.82, 185.08)	168.75 (32.52, 321.46)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V3	sigmoid	NMB (\$ billion)	3	50%	396.63 (177.90, 586.99)	16.70 (-206.54, 207.29)	426.45 (230.79, 603.19)
V3	sigmoid	NMB (\$ billion)	3	80%	628.62 (425.56, 807.12)	23.58 (-254.36, 267.38)	668.95 (437.00, 863.34)
V3	sigmoid	NMB (\$ billion)	3	100%	787.60 (534.26, 1020.35)	25.19 (-258.68, 298.63)	826.38 (584.43, 1049.06)
V3	sigmoid	NMB (\$ billion)	5	20%	231.36 (82.93, 359.40)	-23.49 (-198.21, 135.26)	239.20 (77.17, 368.28)
V3	sigmoid	NMB (\$ billion)	5	50%	579.66 (347.04, 803.20)	-53.14 (-342.45, 207.84)	592.66 (358.26, 802.81)
V3	sigmoid	NMB (\$ billion)	5	80%	922.76 (653.13, 1223.77)	-72.10 (-450.78, 317.79)	951.38 (660.49, 1266.58)
V3	sigmoid	NMB (\$ billion)	5	100%	1147.35 (891.51, 1540.52)	-97.25 (-450.15, 352.45)	1179.06 (893.80, 1582.14)
V3	sigmoid	Treatment courses used (million)	1.2	20%	6.38 (4.91, 7.61)	5.68 (4.15, 6.99)	5.68 (4.15, 6.99)
V3	sigmoid	Treatment courses used (million)	1.2	50%	16.08 (12.22, 18.71)	11.07 (7.28, 14.30)	11.12 (7.35, 14.30)
V3	sigmoid	Treatment courses used (million)	1.2	80%	25.77 (20.69, 29.95)	12.93 (7.25, 17.66)	13.00 (7.25, 17.69)
V3	sigmoid	Treatment courses used (million)	1.2	100%	32.36 (26.00, 36.87)	12.37 (6.66, 17.59)	12.51 (6.66, 18.22)
V3	sigmoid	Treatment courses used (million)	1.5	20%	11.93 (10.68, 13.31)	11.55 (10.38, 12.75)	11.56 (10.38, 12.75)
V3	sigmoid	Treatment courses used (million)	1.5	50%	29.79 (27.31, 32.32)	27.33 (24.78, 29.59)	27.37 (24.78, 29.65)
V3	sigmoid	Treatment courses used (million)	1.5	80%	47.75 (44.61, 50.77)	40.73 (37.59, 44.02)	40.85 (37.76, 44.15)
V3	sigmoid	Treatment courses used (million)	1.5	100%	59.79 (56.31, 63.52)	47.94 (42.80, 52.22)	48.08 (43.62, 52.52)
V3	sigmoid	Treatment courses used (million)	1.7	20%	13.68 (12.45, 15.29)	13.36 (12.06, 14.76)	13.36 (12.06, 14.76)
V3	sigmoid	Treatment courses used (million)	1.7	50%	33.99 (31.66, 35.91)	32.00 (29.13, 34.17)	32.02 (29.06, 34.17)
V3	sigmoid	Treatment courses used (million)	1.7	80%	54.53 (51.04, 57.03)	48.92 (45.67, 51.53)	48.95 (45.70, 51.47)
V3	sigmoid	Treatment courses used (million)	1.7	100%	68.34 (64.61, 70.94)	58.98 (55.91, 62.17)	59.09 (56.01, 62.01)
V3	sigmoid	Treatment courses used (million)	2	20%	16.09 (14.53, 17.66)	15.86 (14.53, 17.40)	15.87 (14.53, 17.36)
V3	sigmoid	Treatment courses used (million)	2	50%	40.42 (38.19, 42.50)	38.81 (36.47, 41.35)	38.83 (36.70, 41.45)
V3	sigmoid	Treatment courses used (million)	2	80%	64.80 (61.78, 68.14)	60.24 (56.94, 63.69)	60.33 (57.26, 63.72)
V3	sigmoid	Treatment courses used (million)	2	100%	81.08 (77.26, 84.71)	73.87 (70.18, 77.43)	73.95 (70.35, 77.96)
V3	sigmoid	Treatment courses used (million)	3	20%	24.34 (22.27, 26.36)	24.14 (22.21, 26.52)	24.14 (22.21, 26.56)
V3	sigmoid	Treatment courses used (million)	3	50%	60.99 (58.15, 63.85)	59.64 (56.61, 62.73)	59.67 (56.80, 62.60)
V3	sigmoid	Treatment courses used (million)	3	80%	97.51 (94.33, 101.12)	94.16 (91.20, 97.03)	94.21 (91.07, 97.07)
V3	sigmoid	Treatment courses used (million)	3	100%	121.98 (119.11, 125.50)	116.64 (113.31, 120.10)	116.79 (113.24, 120.10)
V3	sigmoid	Treatment courses used (million)	5	20%	37.56 (34.99, 39.77)	37.36 (34.86, 39.74)	37.38 (34.96, 39.83)
V3	sigmoid	Treatment courses used (million)	5	50%	93.85 (89.78, 97.89)	92.61 (88.20, 96.87)	92.72 (88.50, 96.84)

Scenarios	Infectiousness	Measures	R <sub>t</sub>	Treatment %	Direct	Indirect	All (Direct + Indirect)
V3	sigmoid	Treatment courses used (million)	5	80%	150.43 (145.30, 154.89)	147.35 (142.17, 151.96)	147.54 (142.83, 152.26)
V3	sigmoid	Treatment courses used (million)	5	100%	188.39 (183.56, 192.95)	183.51 (178.38, 187.81)	183.76 (178.88, 188.10)
V3	sigmoid	Hospitalizations reduced (million)	1.2	20%	0.12 (-0.10, 0.33)	0.14 (-0.23, 0.49)	0.27 (-0.07, 0.59)
V3	sigmoid	Hospitalizations reduced (million)	1.2	50%	0.33 (-0.03, 0.72)	0.46 (-0.07, 1.02)	0.69 (0.26, 1.25)
V3	sigmoid	Hospitalizations reduced (million)	1.2	80%	0.52 (0.03, 0.89)	0.73 (0.16, 1.32)	0.99 (0.46, 1.45)
V3	sigmoid	Hospitalizations reduced (million)	1.2	100%	0.65 (0.07, 1.19)	0.90 (0.30, 1.48)	1.16 (0.59, 1.71)
V3	sigmoid	Hospitalizations reduced (million)	1.5	20%	0.32 (-0.03, 0.66)	0.15 (-0.26, 0.63)	0.44 (0.03, 0.86)
V3	sigmoid	Hospitalizations reduced (million)	1.5	50%	0.79 (0.16, 1.32)	0.35 (-0.43, 1.05)	1.06 (0.36, 1.61)
V3	sigmoid	Hospitalizations reduced (million)	1.5	80%	1.27 (0.40, 1.88)	0.52 (-0.30, 1.22)	1.58 (0.89, 2.31)
V3	sigmoid	Hospitalizations reduced (million)	1.5	100%	1.56 (0.76, 2.24)	0.76 (-0.23, 1.55)	1.95 (1.19, 2.70)
V3	sigmoid	Hospitalizations reduced (million)	1.7	20%	0.38 (0.00, 0.76)	0.09 (-0.63, 0.63)	0.46 (-0.07, 1.02)
V3	sigmoid	Hospitalizations reduced (million)	1.7	50%	0.91 (0.26, 1.48)	0.26 (-0.66, 1.09)	1.15 (0.30, 1.94)
V3	sigmoid	Hospitalizations reduced (million)	1.7	80%	1.49 (0.49, 2.17)	0.46 (-0.49, 1.38)	1.78 (0.89, 2.64)
V3	sigmoid	Hospitalizations reduced (million)	1.7	100%	1.86 (0.99, 2.67)	0.65 (-0.36, 1.75)	2.20 (1.22, 3.00)
V3	sigmoid	Hospitalizations reduced (million)	2	20%	0.45 (-0.03, 0.89)	0.09 (-0.59, 0.72)	0.55 (-0.07, 1.09)
V3	sigmoid	Hospitalizations reduced (million)	2	50%	1.16 (0.23, 2.04)	0.21 (-0.72, 1.02)	1.36 (0.43, 2.14)
V3	sigmoid	Hospitalizations reduced (million)	2	80%	1.89 (0.76, 2.87)	0.44 (-0.46, 1.25)	2.16 (1.22, 2.90)
V3	sigmoid	Hospitalizations reduced (million)	2	100%	2.39 (1.09, 3.46)	0.57 (-0.72, 1.52)	2.67 (1.55, 3.59)
V3	sigmoid	Hospitalizations reduced (million)	3	20%	0.77 (0.33, 1.28)	0.07 (-0.66, 0.92)	0.84 (0.26, 1.48)
V3	sigmoid	Hospitalizations reduced (million)	3	50%	1.98 (0.99, 2.83)	0.23 (-0.86, 1.09)	2.12 (1.22, 2.93)
V3	sigmoid	Hospitalizations reduced (million)	3	80%	3.15 (2.24, 4.02)	0.35 (-0.96, 1.28)	3.32 (2.31, 4.22)
V3	sigmoid	Hospitalizations reduced (million)	3	100%	3.94 (2.67, 5.01)	0.40 (-0.86, 1.58)	4.10 (2.97, 5.14)
V3	sigmoid	Hospitalizations reduced (million)	5	20%	1.15 (0.40, 1.78)	-0.03 (-0.86, 0.72)	1.19 (0.46, 1.85)
V3	sigmoid	Hospitalizations reduced (million)	5	50%	2.91 (1.91, 3.92)	-0.02 (-1.32, 1.22)	2.96 (1.88, 3.95)
V3	sigmoid	Hospitalizations reduced (million)	5	80%	4.63 (3.43, 6.10)	0.02 (-1.78, 1.88)	4.75 (3.43, 6.19)
V3	sigmoid	Hospitalizations reduced (million)	5	100%	5.76 (4.51, 7.64)	0.00 (-1.68, 1.98)	5.90 (4.55, 7.71)

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