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# Appendix to a review article for The Tokyo Foundation for Policy Research T. Kuniya, K. Shibuya, Y. Tokuda, H. Nakamura, T. Moromizato

## 1. Simulation results

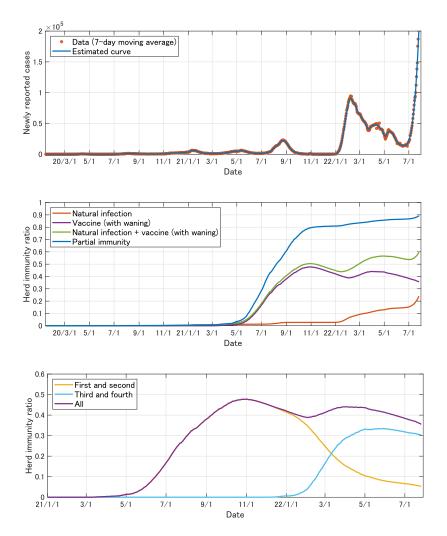


Figure 1: Time variation of newly reported cases (top), estimated herd immunity ratio (middle) and estimated vaccine-induced herd immunity ratio (bottom) for COVID-19 in Japan (2020/1/14 - 2022/7/25).

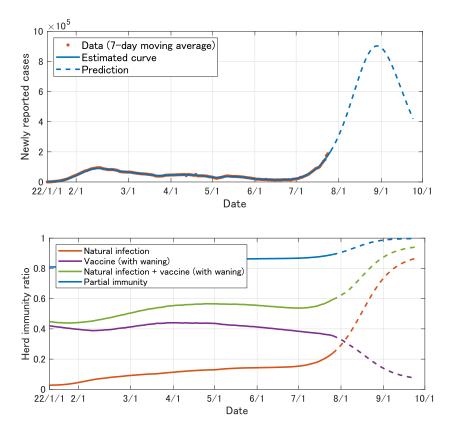


Figure 2: Prediction of newly reported cases (top) and herd immunity ratio (bottom) for COVID-19 in Japan (2022/1/1 - 2022/10/1).

## 2. Parameters

Parameter	Description	Value
S	Susceptible population (unvaccinated)	-
E	Exposed population (unvaccinated)	-
Ι	Infectious population (unvaccinated)	-
R	Removed population (unvaccinated)	-
$S_1$	Susceptible population (vaccinated once)	-
$E_1$	Exposed population (vaccinated once)	-
$I_1$	Infectious population (vaccinated once)	-
$R_1$	Removed population (vaccinated once)	-
$S_2$	Susceptible population (vaccinated twice)	-
$E_2$	Exposed population (vaccinated twice)	-
$I_2$	Infectious population (vaccinated twice)	-
$R_2$	Removed population (vaccinated twice)	-
$S_3$	Susceptible population (vaccinated more than 3 times)	-
$E_3$	Exposed population (vaccinated more than 3 times)	-
$I_3$	Infectious population (vaccinated more than 3 times)	-
$R_3$	Removed population (vaccinated more than 3 times)	-
t	Time	-
a	Class age (time elapsed since the vaccination)	-
$\beta$	Infection rate	Estimated using data in [8]
ε	Onset rate	0.2 (incubation period $1/\varepsilon = 5$ days) [3]
$\gamma$	Removal rate	0.1 (infection period $1/\gamma = 10$ days) [1]
$\lambda$	Force of infection	Equation $(1)$
$1 - \sigma$	Efficacy of one time vaccination	0.46~[5]
$v_n$	Vaccination rate (for $n$ -th)	Estimated using data in [6]
T	Duration between the vaccination	180 days
1 - p(a)	Efficacy of full vaccination at class age $a$	$0.8e^{-0.003a}$ (estimated using data in [5])
δ	Detection rate	0.5 (estimated using data in [4])
N	Total population in Japan	$1.26 \times 10^8$ [7]

See [2] for the details of how to estimate each parameter.

## 3. Model

Before vaccination policy (January 14, 2020 - February 16, 2021).

$$S'(t) = -\beta S(t)I(t),$$
  

$$E'(t) = \beta S(t)I(t) - \varepsilon E(t),$$
  

$$I'(t) = \varepsilon E(t) - \gamma I(t),$$
  

$$R'(t) = \gamma I(t).$$

Under vaccination policy (February 17, 2021 - July 25, 2022).

• Unvaccinated population:

$$S'(t) = -\lambda(t)S(t) - v_1S(t),$$
  

$$E'(t) = \lambda(t)S(t) - (\varepsilon + v_1)E(t),$$
  

$$I'(t) = \varepsilon E(t) - (\gamma + v_1)I(t),$$
  

$$R'(t) = \gamma I(t) - v_1R(t).$$

• Vaccinated once:

$$S_1'(t) = v_1 S(t) - \sigma \lambda(t) S_1(t) - v_2 S_1(t),$$
  

$$E_1'(t) = v_1 E(t) + \sigma \lambda(t) S_1(t) - (\varepsilon + v_2) E_1(t),$$
  

$$I_1'(t) = v_1 I(t) + \varepsilon E_1(t) - (\gamma + v_2) I_1(t),$$
  

$$R_1'(t) = v_1 R(t) + \gamma I_1(t) - v_2 R_1(t).$$

• Vaccinated more than twice (n = 2, 3):

$$\begin{split} S_n(t,0) &= \begin{cases} v_2 S_1(t), & n = 2, \\ v_3 \int_T^{\infty} S_2(t,a) da + v_4 \int_T^{\infty} S_3(t,a) da, & n = 3, \end{cases} \\ E_n(t,0) &= \begin{cases} v_2 E_1(t), & n = 2, \\ v_3 \int_T^{\infty} E_2(t,a) da + v_4 \int_T^{\infty} E_3(t,a) da, & n = 3, \end{cases} \\ I_n(t,0) &= \begin{cases} v_2 I_1(t), & n = 2, \\ v_3 \int_T^{\infty} I_2(t,a) da + v_4 \int_T^{\infty} I_3(t,a) da, & n = 3, \end{cases} \\ R_n(t,0) &= \begin{cases} v_2 R_1(t), & n = 2, \\ v_3 \int_T^{\infty} R_2(t,a) da + v_4 \int_T^{\infty} R_3(t,a) da, & n = 3, \end{cases} \\ \left(\frac{\partial}{\partial t} + \frac{\partial}{\partial a}\right) S_n(t,a) &= -p(a)\lambda(t)S_n(t,a) - q_n(a)S_n(t,a), \end{cases} \\ \left(\frac{\partial}{\partial t} + \frac{\partial}{\partial a}\right) E_n(t,a) &= p(a)\lambda(t)S_n(t,a) - [\varepsilon + q_n(a)]E_n(t,a), \end{cases} \\ \left(\frac{\partial}{\partial t} + \frac{\partial}{\partial a}\right) I_n(t,a) &= \varepsilon E_n(t,a) - [\gamma + q_n(a)]I_n(t,a), \\ \left(\frac{\partial}{\partial t} + \frac{\partial}{\partial a}\right) R_n(t,a) &= \gamma I_n(t,a) - q_n(a)R_n(t,a), \end{split}$$

where

$$q_n(a) = \begin{cases} 0, & a < T, \\ v_{n+1}, & \text{otherwise.} \end{cases}$$

• Force of infection:

$$\lambda(t) = \beta \left[ I(t) + I_1(t) + \sum_{n=2}^3 \int_0^\infty I_n(t, a) da \right].$$
 (1)

• Efficacy of full vaccination at class age  $a: 1 - p(a) = 0.8e^{-0.003a}$ , which is fitted to the data in [5] as shown in Figure 3.

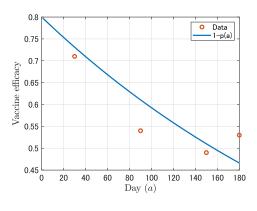


Figure 3:

• Let

$$\begin{split} M_0(t) &:= E(t) + I(t) + R(t), \quad M_1(t) := E_1(t) + I_1(t) + R_1(t), \\ M_n(t) &:= \int_0^\infty [E_n(t,a) + I_n(t,a) + R_n(t,a)] da, \quad n \ge 2. \end{split}$$

Description for each curve in Figure 1:

- Natural infection:  $\sum_{n=0}^{3} M_n(t)$ .
- Vaccine (with waning):  $(1-\sigma)S_1(t) + \sum_{n=2}^3 \int_0^\infty [1-p(a)]S_n(t,a)da$ .
- Natural infection + vaccine (with waning):  $\sum_{n=0}^{3} M_n(t) + (1-\sigma)S_1(t) + \sum_{n=2}^{3} \int_0^\infty [1-p(a)]S_n(t,a)da.$
- Partial immunity: 1 S(t).
- First and second:  $(1-\sigma)S_1(t) + \int_0^\infty [1-p(a)]S_2(t,a)da$
- Third and fourth:  $\int_0^\infty [1-p(a)] S_3(t,a) da$
- All:  $(1 \sigma)S_1(t) + \sum_{n=2}^3 \int_0^\infty [1 p(a)]S_n(t, a)da.$

How to estimate  $\beta = \beta(t)$  and  $\delta$ See [2].

### How to estimate the vaccination rates

Note that  $v_1 \times [S(t) + E(t) + I(t) + R(t)] \times N$  is the number of the first vaccination at time t. Hence, we estimate  $v_1 = v_1(t)$  as

$$v_1(t) = \frac{(number \ of \ the \ first \ vaccination \ at \ time \ t)}{[S(t) + E(t) + I(t) + R(t)] \times N}.$$

In a similar manner, we estimate  $v_n = v_n(t)$   $(n \ge 2)$  as

$$v_{n}(t) = \begin{cases} \frac{(number \ of \ the \ second \ vaccination \ at \ time \ t)}{[S_{1}(t) + E_{1}(t) + I_{1}(t) + R_{1}(t)] \times N}, & n = 2, \\ \frac{(number \ of \ the \ n-th \ vaccination \ at \ time \ t)}{\int_{T}^{\infty} [S_{n-1}(t,a) + E_{n-1}(t,a) + I_{n-1}(t,a) + R_{n-1}(t,a)] da \times N}, & n \ge 3. \end{cases}$$

#### How to predict

We fixed the infection rate and vaccination rates using the latest 1 week data.

#### References

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