

Effect of Asymptomatic Transmission and Emergence Time on Multi-strain Viral Disease Severity

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RESEARCH ARTICLE

Effect of asymptomatic transmission and emergence time on multi-strain viral disease severity

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Abstract

Background: In a viral epidemic, the emergence of a novel strain with increased transmissibility (larger value of basic reproduction number R_0) sparks the fear that the increase in transmissibility is likely to lead to an increase in disease severity. It is required to investigate if a new, more contagious strain will be necessarily dominant in the population and resulting in more disease severity.

Methods: The impact of the asymptomatic transmission and the emergence time of a more transmissible variant of a multi-strain viral disease on the disease prevalence, disease severity, and the dominant variant in an epidemic was investigated by a proposed 2-strain epidemic model.

Results: The simulation results showed that considering only R_0 , is insufficient to predict the outcome of a new, more contagious strain in the population. A more transmissible strain with a high fraction of asymptomatic cases can substantially reduce the mortality rate. If the emergence time of the new strain is closer to the start of the epidemic, the new, more contagious variant has more chance to win the viral competition and be the dominant strain; otherwise, despite being more contagious, it cannot dominate previous strains.

Conclusions: Three factors of R_0 , the fraction of asymptomatic transmission, and the emergence time of the new strain are required to correctly determine the prevalence, disease severity, and the winner of the viral competition.

Keywords: asymptomatic; basic reproduction number; COVID-19; disease severity; dominant strain; emergence time; multi-strain virus

¹Background

² The emergence of the novel coronavirus strain in the UK, called SARS-CoV-2 VOC
³ 202012/01 or B.1.1.7, was shocking because this novel variant could be up to about
⁴ 70% more transmissible than pre-existing variants of SARS-CoV-2 [1]. This in-
⁵ creased transmissibility can add between 0.4 and 0.7 to the basic reproduction
⁶ number R_0 . This news sparked the fear that the increase in transmissibility is likely
⁷ to lead to a large increase in hospitalization, intensive care unit (ICU) admission
⁸ rate, and mortality. However, the studies about previous variants of SARS-CoV-2
⁹ showed that despite the rise of the lab-confirmed cases, the COVID-19 case fatal-
¹⁰ ity rate (CFR) declined, i.e., more transmissibility did not necessarily cause more
¹¹ severity [2, 3]. Other studies in the UK and England showed that besides increasing
¹² the COVID-19 cases, the hospitalization rate, the ICU admission rate, and the CFR
¹³ declined [4, 5]. The preliminary explanation was the predominant shift towards pos-
¹⁴ itivity in younger age groups who have a better outcome. However, the analysis of
¹⁵ German COVID-19 data [6], which was reported by age categories, showed that the
¹⁶ COVID-19 CFR declined across all age groups [7]. Interestingly, the older groups
¹⁷ drove the overall reduction in CFR.
¹⁸

¹⁹ The public health authorities need to pinpoint the cause of this decline in the fatal-
²⁰ ity rate in order to decide how to react against the newly emerged viral strains. The
²¹ decision to fight blindly against a novel strain because of its increased transmissibil-
²² ity is not necessarily the most comprehensive and effective solution. We need to take
²³ other factors along with the transmissibility into account in our decision-making.

²⁴ The spread of COVID-19 is an iceberg with the invisible part of being the asymp-
²⁵ tomatic transmission. The percent of asymptomatic cases who never experience
²⁶ COVID-19 symptoms remains uncertain. From about 20% to 50% of infected peo-
²⁷ ple are reported to be asymptomatic [8, 9, 10]. In a study, 39% of children aged
²⁸ 6-13 years tested positive for COVID-19 with no symptoms [11]. Different stud-
²⁹ ies reported an insignificant difference in the upper respiratory viral load between
³⁰ symptomatic and asymptomatic cases [8, 12]. Even a new study found that asymp-
³¹ tomatic patients had higher SARS-CoV-2 viral loads than symptomatic cases [13].
³² Consequently, the asymptomatic infected people could play a significant driver role
³³ in the community spread of COVID-19. The results of a study demonstrated that

¹both R_0 and the proportion of asymptomatic transmissions were the main factors¹
²in controlling an infectious disease outbreak [14].²
³³
⁴⁴
⁵ In this study, we investigate the effect of the emergent viral strain on the totals
⁶number of infected people and the illness severity by using epidemiological modeling.⁶
⁷We will show that in an epidemic situation, the emergence time of the new strain⁷
⁸and the relative R_0 of the primary and the emergent strains determine the winners⁸
⁹of the competition between two viral strains. Moreover, we will see that the disease⁹
¹⁰severity and the mortality rate can be significantly influenced by the emergence¹⁰
¹¹time and the fraction of asymptomatic infectious cases of the emergent strain.¹¹

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¹⁹Methods¹⁹

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²¹²¹
²²For each viral strain, we use the extended version of the classic SEIRD epidemic²²
²³model, called the SEICARD model, consisting of susceptible (S), exposed (in the²³
²⁴latent period) (E), symptomatic infected (I), critically infected (C), asymptomatic²⁴
²⁵infected (A), recovered (R), and dead (D) people. By paralleling two SEICARD²⁵
²⁶models, we develop a 2-strain model, called 2-SEICARD, that describes the exis-²⁶
²⁷tence and competition of two viral variants in the population (Fig. 1). The index²⁷
²⁸ $s = 1$ or 2 in E_s , I_s , C_s , A_s , R_s , and D_s represents the infectious strain in each²⁸
²⁹group. It is assumed that the emergence time of the second strain is T_E days after²⁹
³⁰the emergence time of the primary one, which is day 0. Moreover, we assume that³⁰
³¹there is no viral superinfection, i.e., the reinfection or co-infection between variants³¹
³²does not occur. In other words, the recovered individuals are cross-immunized and³²
³³are immune to the new variants.³³

¹ The parameters of the 2-SEICARD model are explained in Table 1. The ODE¹
²system of this 2-strain model is given by ²

$$\begin{aligned} \text{3} & \\ \text{4} \quad dS/dt &= - \sum_{s=1}^2 (\beta_s^I I_s + \beta_s^C C_s + \beta_s^A A_s) S/N, & \text{(1a)}^4 \\ \text{5} & \end{aligned}$$

$$\begin{aligned} \text{6} \quad dE_s/dt &= (\beta_s^I I_s + \beta_s^C C_s + \beta_s^A A_s) S/N - \varepsilon_s E_s, & \text{(1b)}_6 \end{aligned}$$

$$\begin{aligned} \text{7} \quad dI_s/dt &= (1 - P_s^A)(1 - P_s^C) \varepsilon_s E_s - \gamma_s^I I_s, & \text{(1c)}_7 \end{aligned}$$

$$\begin{aligned} \text{8} \quad dC_s/dt &= (1 - P_s^A) P_s^C \varepsilon_s E_s - [(1 - P_s^D) \gamma_s^C + P_s^D \gamma_s^D] C_s, & \text{(1d)}_8 \end{aligned}$$

$$\begin{aligned} \text{9} \quad dA_s/dt &= P_s^A \varepsilon_s E_s - \gamma_s^A A_s, & \text{(1e)}_9 \end{aligned}$$

$$\begin{aligned} \text{10} \quad dR_s/dt &= \gamma_s^I I_s + (1 - P_s^D) \gamma_s^C C_s + \gamma_s^A A_s, & \text{(1f)}_{10} \end{aligned}$$

$$\begin{aligned} \text{11} \quad dD_s/dt &= P_s^D \gamma_s^D C_s & \text{(1g)}_{11} \\ \text{12} & \end{aligned}$$

¹³ for $s = 1$ and 2 . The total population is $N = S + \sum_{s=1}^2 (E_s + I_s + C_s + A_s + R_s + D_s)$. ¹³

¹⁴ For simplicity, the natural birth and death rates are ignored in the model. To im- ¹⁴
¹⁵plement the emergence time T_E , we set all parameters of the second strain to zero ¹⁵
¹⁶for $t < T_E$. ¹⁶

¹⁷ For each strain, the infection rates β_s^I , β_s^C , and β_s^A denote the probability of ¹⁷
¹⁸transmitting disease from I_s , C_s , or A_s to S , respectively. On the other hand, as ¹⁸
¹⁹Fig. 1 shows, the outcome of each exposed person E_s could be A_s , I_s , C_s , and ¹⁹
²⁰ D_s with the probabilities of P_s^A , $(1 - P_s^A)(1 - P_s^C)$, $(1 - P_s^A)P_s^C(1 - P_s^D)$, and ²⁰
²¹ $(1 - P_s^A)P_s^C P_s^D$, respectively. By using the method of next-generation matrices [15], ²¹
²²we can obtain the following expression for the R_0 of each strain ²²
²³

$$\begin{aligned} \text{24} \quad R_0^{(s)} &= P_s^A \frac{\beta_s^A}{\gamma_s^A} + (1 - P_s^A)(1 - P_s^C) \frac{\beta_s^I}{\gamma_s^I} + (1 - P_s^A) P_s^C (1 - P_s^D) \frac{\beta_s^C}{\gamma_s^C} & \text{24} \\ \text{25} & \\ \text{26} \quad &= P_s^A R_0^{A(s)} + (1 - P_s^A)(1 - P_s^C) R_0^{I(s)} + (1 - P_s^A) P_s^C (1 - P_s^D) R_0^{C(s)}, & \text{(2)}_{26} \\ \text{27} & \end{aligned}$$

²⁸where $R_0^{A(s)}$, $R_0^{I(s)}$, and $R_0^{C(s)}$ denote the reproduction number of each outcome, ²⁸
²⁹and $R_0^{(s)}$ is obtained by their weighted sum. The weight of each outcome is the ²⁹
³⁰probability of its occurrence. ³⁰

³¹ The values of different parameters used in the 2-SEICARD model are listed in ³¹
³²Table 1. In our modeling, we assume a wild animal population in which there is no ³²
³³isolation and social policy or restriction. Hence, we consider that all I_s , C_s , and A_s ³³

outcomes have the same probability of transmission; i.e., $\beta_s^I = \beta_s^C = \beta_s^A$. According to Eq. (2), by considering $\beta_1^I = \beta_1^C = \beta_1^A = 0.2$ and the values of 0.13, 0.2, and 0.27 for all β values of strain 2, we have $R_0^{(1)} = 2$ and $R_0^{(2)} = 1.3, 2, \text{ and } 2.7$, respectively.

Results and Discussion

In this section, we consider fixed parameters for the first strain in the 2-SEICARD model; i.e., $R_0^{(1)} = 2$ and $P_1^A = 0.1$. Then, the emergent strain with different values of $R_0^{(2)} = 1.3, 2, \text{ and } 2.7$ and $P_2^A = 0.1, 0.2 \text{ and } 0.4$ emerges at day T_E , where $0 \leq T_E \leq 100$. In all the above scenarios, we study the effect of the emergent strain on the total number of infected cases and the mortality rate, as a measure of severity, from the beginning of the epidemic until we reach the endemic steady state. The total number of infected cases is $N - S_\infty$, where S_∞ denotes the number of susceptible cases that have not been infected at all when the disease has gone. The mortality rate is the total proportion of deaths in the population due to infection.

Effect of R_0 and T_E on the total number of infections and the dominant strain

The simulation results show that the total number of infected cases does not vary with P_2^A for the fixed values of $R_0^{(1)}$ and $R_0^{(2)}$. In other words, the values of $R_0^{(1)}$ and $R_0^{(2)}$ determine the total number of infected individuals during the epidemic spread.

Provided that $R_0^{(2)} < R_0^{(1)}$, the emergent strain does not have any chance to compete with the primary strain and would become extinct immediately (see Fig. 2(A1)).

In the case of $R_0^{(2)} = R_0^{(1)}$, the total number of infected cases with two strains remains the same as that in the case of spreading only the primary strain in the population with the same value of the basic reproduction number. Moreover, Fig.

2(A2) demonstrates that the total number of infected cases does not vary with the emergence time of the second strain, T_E . However, the later emergence of strain 2

results in less proportion of infection with this strain in the population. In contrast,

Fig. 2(A3) depicts that the emergence of a more contagious strain ($R_0^{(2)} > R_0^{(1)}$)

increases the total number of infected cases compared to the existence of only the primary strain. Furthermore, Fig. 2(A3) shows that the new, more contagious strain

with larger value of R_0 does not necessarily dominate in the population. In other

words, R_0 alone does not determine which virus wins the viral competition. Indeed,

besides the values of R_0 , the emergence time T_E also determines whether the new

strain with more transmissibility dominates the primary one or not. The sooner

emergence of the new, more contagious variant can make it dominant; otherwise, the primary strain remains dominant in the population.

Effect of emergent strain on mortality rate

Our main question is that in the case of $R_0^{(2)} > R_0^{(1)}$, whether or not more infected cases increase similarly the mortality rate of the disease or not. To answer this question, consider the mortality rate in different circumstances in Figures 2(B1-B3). These figures show concurrently the effect of the fraction of asymptomatic individuals infected with strain 2, P_2^A , and the emergence time of the second strain, T_E , on the mortality rate for $R_0^{(1)} = 2$ and $R_0^{(2)} = 1.3, 2$, and 2.7 . As we have discussed, if $R_0^{(2)} < R_0^{(1)}$, the emergent strain cannot compete with the primary strain, and hence, the mortality rate remains fixed for all values of P_2^A , equal to the mortality rate of the primary strain alone. Hence, in these figures, although blue curves are corresponding to the 2-strain scenario, they also represent the mortality rate of the primary strain alone.

As we expected, provided that $R_0^{(1)} = R_0^{(2)}$ and $P_1^A = P_2^A$, the mortality rate remains the same as that of the primary strain alone. On the other hand, as Fig. 2(B1) depicts, if both strains have a similar proportion of asymptomatic cases, i.e., $P_1^A = P_2^A$, the mortality rate increases with the emergence of a more contagious strain ($R_0^{(2)} > R_0^{(1)}$). In this case, the sooner that the new strain emerges, the more in the mortality rate increases. However, as Figures 2(B2-B3) show, the increase in P_2^A can reduce the mortality rate for $R_0^{(2)} \geq R_0^{(1)}$. For large values of P_2^A , although a more contagious strain emerges, it can make the mortality rate less than that before the emergence (Fig. 2(B3)). Interestingly, for large values of P_2^A , the emergent strain with higher R_0 decreases the mortality rate more. In other words, more transmissibility does not reflect necessarily more severity, and both $R_0^{(s)}$ and P_s^A values should be considered to correctly determine the effect of the more transmissible new strain on the viral disease severity.

Conclusion

In this study, we investigated the impact of the asymptomatic transmission and the emergence time of the new, more contagious viral strain on the disease prevalence, disease severity, and the dominant variant in an epidemic. Our results demonstrated that being an emergent strain with more transmissibility, i.e., having a larger basic

¹reproduction number R_0 , compared to previous variants, does not necessarily lead¹
²to more severity or mean that the new variant will dominate in the population.²
³Indeed, when a new strain with a larger basic reproduction number emerges, it will³
⁴increase the number of infected cases in the population. However, the creation of⁴
⁵more severe outcomes depends on the fraction of asymptomatic transmissions. If a⁵
⁶large proportion of infections, due to the new variant, do not show any symptom,⁶
⁷they can even reduce the mortality rate in the population. Moreover, provided that⁷
⁸the emergence time of the new strain is closer to the start of the epidemic, the⁸
⁹new, more contagious variant has more chance to win the viral competition and be⁹
¹⁰the dominant strain; otherwise, despite being more contagious, it cannot dominate¹⁰
¹¹previous strains. 11

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¹⁴research. 14

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17 Abbreviations 17

¹⁸CFR: Case fatality rate; COVID-19: Coronavirus disease 2019 18

19 Availability of data and materials 19

¹⁹Not applicable. 19

20 20**Ethics approval and consent to participate**

²¹Not required. 21

22 Competing interests 22

²³The authors declare that they have no competing interests. 23

24 Consent for publication 24

Not applicable.

25 25**Authors' contributions**

²⁶A.A. developed, analyzed, and simulated the methods, and wrote the manuscript. M.S. conceived the study and 26

²⁷guided the method development. All authors read and approved the final manuscript. 27

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30 References 30

³¹1. Davies NG, Barnard RC, Jarvis CI, Kucharski AJ, Munday J, Pearson CAB, et al. Estimated transmissibility and
³¹severity of novel SARS-CoV-2 Variant of Concern 202012/01 in England. medRxiv. 2020;. 31

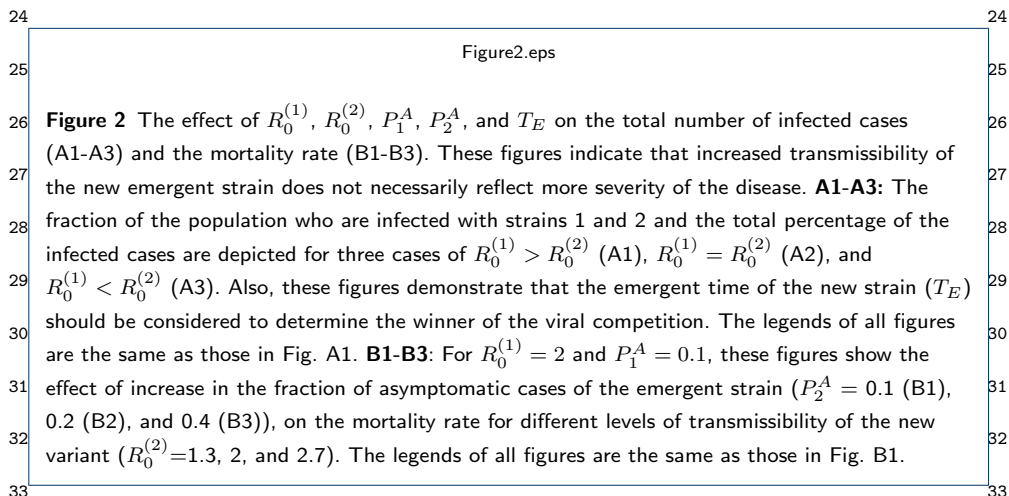
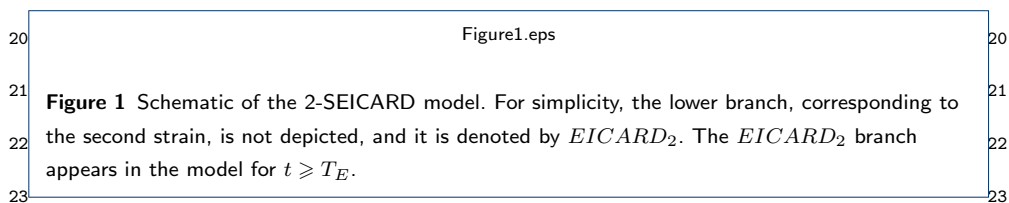
³²2. SARS-CoV-2 Variants;. Available from: www.who.int/csr/don/31-december-2020-sars-cov2-variants/. 32

³³3. Why do COVID death rates seem to be falling?;. Available from: 33

www.nature.com/articles/d41586-020-03132-4.

1 4. Howdon D, Oke J, Heneghan C. COVID-19: Declining Admissions to Intensive Care Units;. Available from: 1
 2 www.cebm.net/covid-19/covid-19-declining-admissions-to-intensive-care-units. 2
 3 5. Howdon D, Heneghan C. The Declining Case Fatality Ratio in England;. Available from: 3
 4 www.cebm.net/covid-19/the-declining-case-fatality-ratio-in-england. 4
 5 6. RKI COVID19 data hub;. Available from: https://npgco-corona-npgeo-de.hub.arcgis.com/datasets/dd4580c810204019a7b8eb3e0b329dd6_0/data. 5
 6 7. Oke J, Howdon D, Heneghan C. Declining COVID-19 Case Fatality Rates across all ages: analysis of German 6
 7 data;. Available from: <https://www.cebm.net/covid-19/declining-covid-19-case-fatality-rates-across-all-ages-analysis-of-german-data>. 7
 8 8. Ra SH, Lim JS, Kim Gu, Kim MJ, Jung J, Kim SH. Upper respiratory viral load in asymptomatic individuals 8
 9 and mildly symptomatic patients with SARS-CoV-2 infection. *Thorax*. 2021;76(1):61–63. 9
 10 9. Gudbjartsson DF, Helgason A, Jonsson H, Magnusson OT, Melsted P, Norddahl GL, et al. Spread of 10
 11 SARS-CoV-2 in the Icelandic Population. *New England Journal of Medicine*. 2020;382(24):2302–2315. 11
 12 10. Byambasuren O, Cardona M, Bell K, Clark J, McLaws ML, Glasziou P. Estimating the extent of asymptomatic 12
 13 COVID-19 and its potential for community transmission: Systematic review and meta-analysis. *Journal of the* 13
 14 *Association of Medical Microbiology and Infectious Disease Canada*. 2020;5:223–234. 14
 15 11. Hurst JH, Heston SM, Chambers HN, Cunningham HM, Price MJ, Suarez L, et al. Severe Acute Respiratory 15
 16 Syndrome Coronavirus 2 Infections Among Children in the Biospecimens from Respiratory Virus-Exposed Kids 16
 17 (BRAVE Kids) Study. *Clinical Infectious Diseases*. 2020 11;. 17
 18 12. Cereda D, Tirani M, Rovida F, Demicheli V, Ajelli M, Poletti P, et al. The early phase of the COVID-19 18
 19 outbreak in Lombardy, Italy. *arXiv*. 2020;Available from: <https://arxiv.org/abs/2003.09320>. 19
 20 13. Hasanoglu I, Korukluoglu G, Asilturk D, Cosgun Y, Kalem AK, Altas AB, et al. Higher viral loads in 20
 21 asymptomatic COVID-19 patients might be the invisible part of the iceberg. *Infection*. 2020 Nov;. 21
 22 14. Fraser C, Riley S, Anderson RM, Ferguson NM. Factors that make an infectious disease outbreak controllable. 22
 23 *Proceedings of the National Academy of Sciences*. 2004;101(16):6146–6151. 23
 24 15. Diekmann O, Heesterbeek JA, Roberts MG. The construction of next-generation matrices for compartmental 24
 25 epidemic models. *J R Soc Interface*. 2010 Jun;7(47):873–885. 25

19 **Figures**



1	Tables		1
2	Table 1	Explanation of the symbols of the 2-SEICARD model.	2
3	Symbol	Explanation	Value
4	s	Strain number	1, 2
5	S	Susceptible individuals	4
6	E_s	Exposed to strain s and still in the latent period	5
7	I_s	Symptomatic individuals infected with strain s	6
8	C_s	Critically infectious individuals infected with strain s	6
9	A_s	Asymptomatic individuals infected with strain s	7
10	R_s	Individuals recovered from strain s infection	8
11	D_s	Dead individuals infected with strain s	8
12	N	Total number of individuals	10000
13	P_s^A	Fraction of asymptomatic individuals infected with strain s	0.1 ($s = 1$) 0.1, 0.2, 0.4 ($s = 2$)
14	P_s^C	Fraction of symptomatic individuals who are critically infected with strain s	0.1
15	P_s^D	Fraction of critically infected individuals who die from infection with strain s	0.05
16	$\beta_s^I, \beta_s^C, \beta_s^A$	Infection rate of different outcomes of strain s	0.2 ($s = 1$) 0.13, 0.2, 0.27 ($s = 2$)
17	$1/\varepsilon_s$	Average incubation period of strain s	5 days
18	$1/\gamma_s^I, 1/\gamma_s^C, 1/\gamma_s^A, 1/\gamma_s^D$	Average infection period of different outcomes of strain s	10 days
19	T_E	Emergence time of strain 2	$0 \leq T_E \leq 100$ day
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Figures

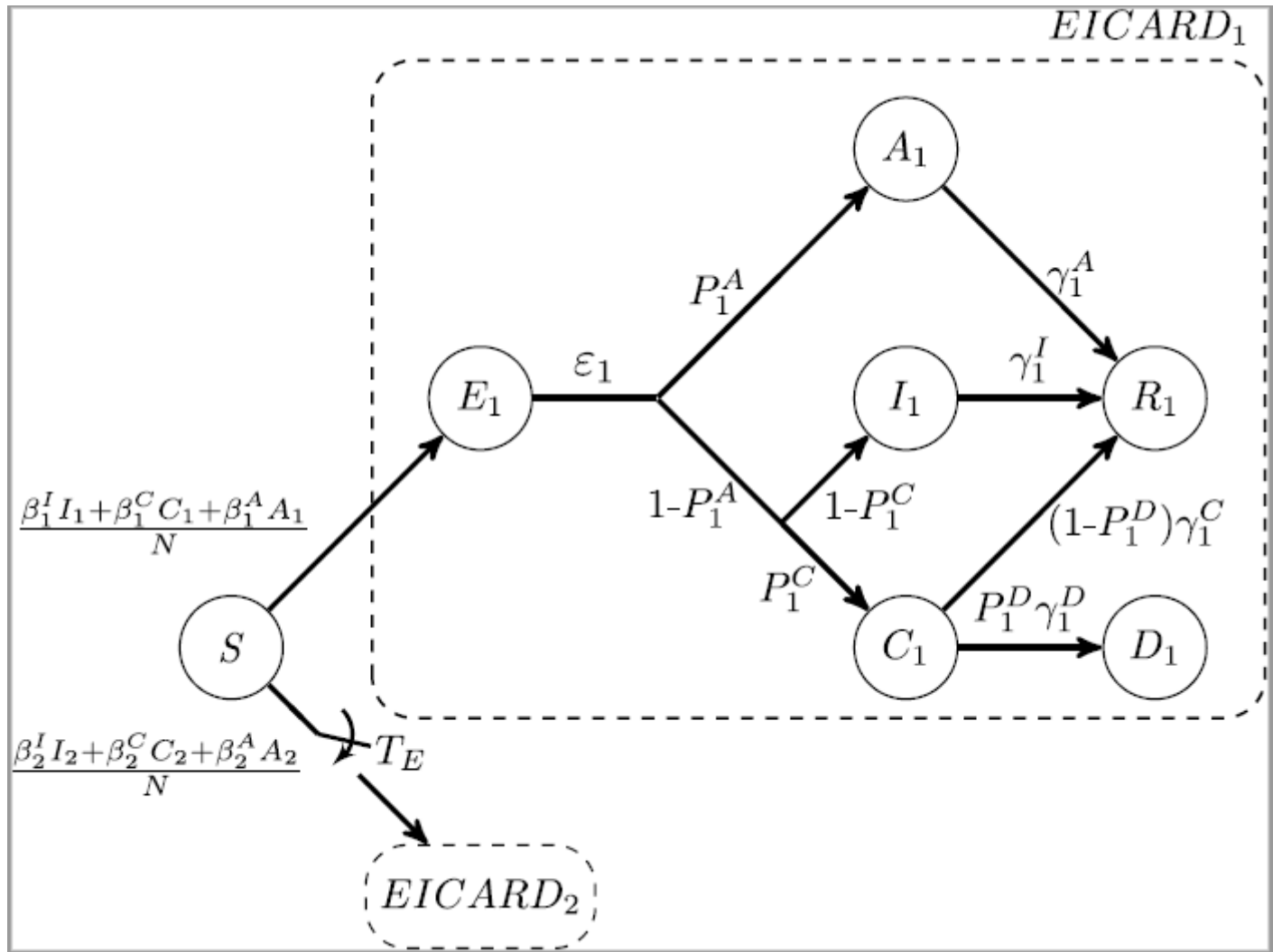


Figure 1

See the Manuscript Files section for the complete figure caption.

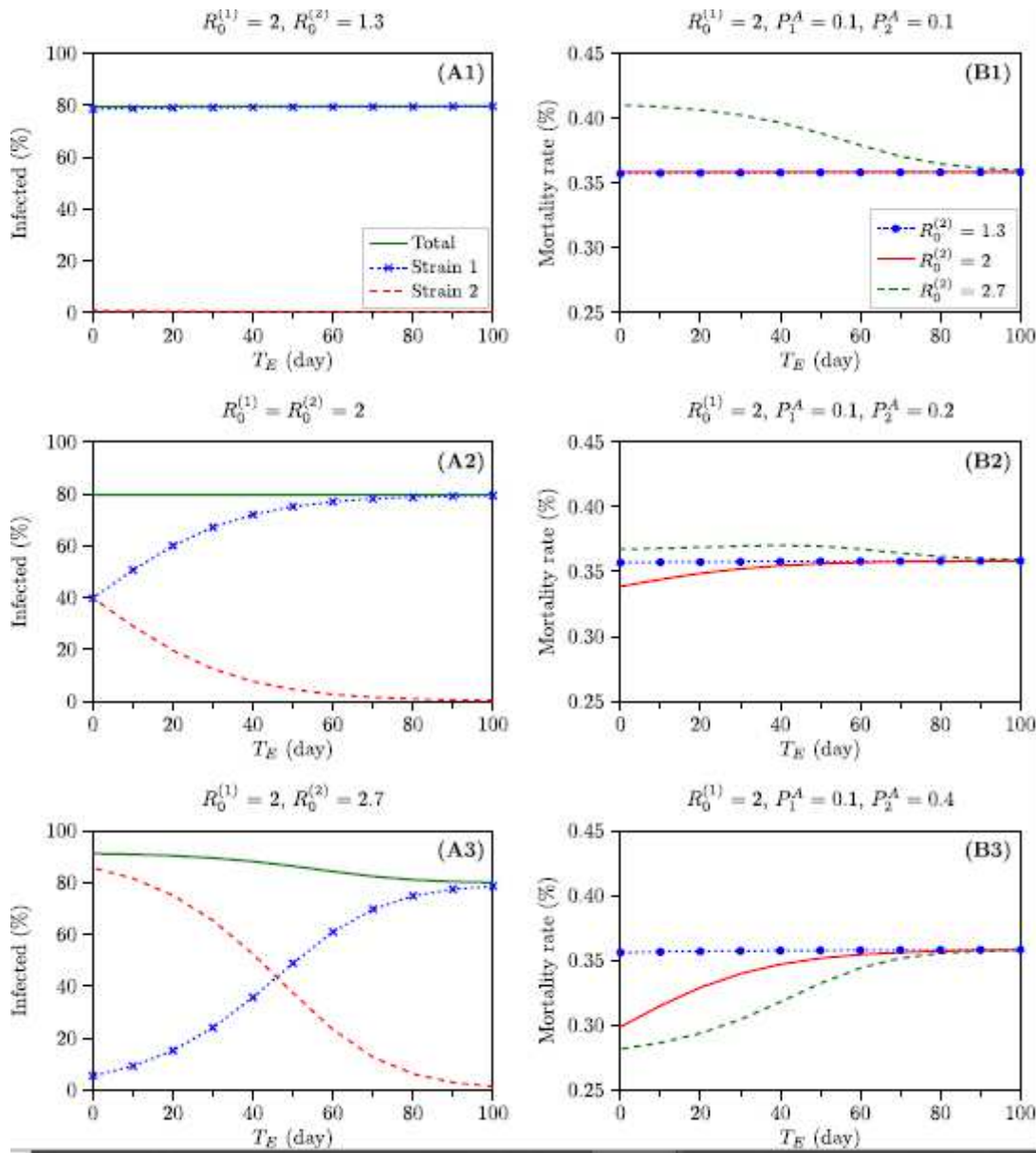


Figure 2

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