

1 **Test-and-treat approach to HIV/AIDS: A primer for**
2 **mathematical modeling**

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15

16 **Abstract**

17 The public benefit of treatment-as-prevention has induced a need to justify goodness
18 for the public, and mathematical modeling studies played a key role in designing and
19 evaluating the test-and-treat strategy for controlling HIV/AIDS. Here we briefly and
20 comprehensively review the essence of contemporary understanding of treatment-as-
21 prevention policy through mathematical modeling approaches and identify key pitfalls
22 that have been identified to date. While the decrease in HIV incidence is achieved
23 with certain coverages of diagnosis, care and continued treatment, HIV prevalence is
24 not necessarily decreased and sometimes the test-and-treat is accompanied by
25 increased long-term cost of antiretroviral therapy (ART). To confront with the
26 complexity of assessment for this policy, the elimination threshold or the effective
27 reproduction number has been proposed for its use in determining the overall success
28 to anticipate eventual elimination. Since the publication of original model in 2009,
29 key issues of test-and-treat modeling studies, including theoretical problems
30 surrounding the sexual partnership network, detailed transmission dynamics and
31 heterogeneous risk groups, have been identified. To explicitly design country-specific
32 control policy, quantitative modeling approaches to each single setting with differing
33 epidemiological context area required through collaboration among clinicians, public
34 health practitioners, laboratory technologists, epidemiologists and mathematical
35 modelers.

36

37 **Background**

38 Whereas the treatment of diseases has been conducted to expect individual benefit,
39 e.g. aiming for eventual cure, in medical facilities, its use for directly transmitted
40 infectious diseases can sometimes offer public benefits. Such treatment for the public
41 interest is represented by the so-called “test and treat” approaches to HIV/AIDS [1]
42 and another well-known radical approach may be the eradication therapy of
43 *Helicobacter pylori* infection in the stomach. The very first test-and-treat model by
44 Granich and his colleagues has excellently resulted in forming a landmark of global
45 health policy [2], assisting the world to be motivated to universally or at least
46 radically screen HIV infected individuals in the population and promote their
47 treatment, not only for their suppression from progression of HIV infection but also
48 for the public benefit. Nevertheless, the public benefit has also induced a need to
49 justify goodness for the public, because “treatment as prevention” is no longer an
50 individual interest but something to be ensured by the public or governmental
51 organizations for its preventive performance [1].

52 The very first model of test-and-treat [2] has been repeatedly criticized for its
53 practical utility, controversies and oversimplified model structure, and a number of
54 alternative mathematical approaches have been proposed to assess the population
55 impact of test-and-treat strategy in both quantitative and qualitative manners. It is
56 valuable to overview mathematical approaches to test-and-treat strategy of HIV/AIDS
57 for both general and expert readers as a primer. The present short review aims to
58 briefly share the essence of contemporary understanding of the treatment-as-
59 prevention.

60 **What is test-and-treat?**

61 In the simplest manner, the test-and-treat strategy is mathematically captured by a
62 four-compartmental model system (Figure 1). While HIV infected individuals are at
63 risk of developing AIDS in a matter of some 10 years since infection, diagnosis of
64 HIV in advance of AIDS could bring infected individuals under antiretroviral therapy
65 (ART). Effective ART in preventing infected individuals from their
66 pathophysiological progression to AIDS has been established and continuously
67 improved over time [3]. In theoretical sense, ART at the population level is
68 considered to offer three different types of impact, i.e., (i) reduced opportunity of
69 secondary transmission [4,5], (ii) reduced infectiousness per contact [6,7], and (iii)
70 individual impact including extended life expectancy [8], and reduced risks of AIDS
71 and AIDS death [3,9]. Considering these benefits, Granich et al. [2] have shown that
72 substantial herd immunity (or to be more precise “indirect population effect” of mass
73 treatment; hereafter we use “herd immunity” for simplicity) could be attained by a
74 combination of universal testing and expanded ART among all infected individuals,
75 helping to curb the HIV epidemic, assuming that a high adherence level is maintained
76 for decades.

77 To achieve such indirect effect and individual treatment series by HIV
78 screening and treatment at a population level, it is essential to ensure that three key
79 tasks are achieved, i.e., (i) finding HIV infected individuals, (ii) maintaining HIV care
80 and monitoring CD4-positive T cell count and (iii) ensuring adherence and successful
81 ART to suppress viral load. The Joint United Nations Programme on HIV/AIDS
82 (UNAIDS) has introduced the concept of an HIV treatment cascade to identify and fill
83 gaps in the continuum of services for testing, care and effective treatment. Following
84 the 21st International AIDS Conference in Durban, South Africa, the UNAIDS report

85 has led to a global slogan of “90-90-90” by 2020 that aims to achieve targets, which
86 are that 90% of people living with HIV know their HIV infection, 90% of people who
87 know their HIV infection are accessing treatment and 90% of people on treatment has
88 enjoyed suppressed viral loads [10]. By the year 2030, UNAIDS is even aiming to
89 achieve 95-95-95 at a global level. From a variety of countries, care cascade of the
90 HIV/AIDS has been estimated and evaluated (e.g. Figure 2 [11]), helping the country
91 to point out the ongoing weakness of interventions. For instance, the case study of the
92 United States in 2011 indicates that the diagnostic coverage is close to reach 90%,
93 while more than half of diagnosed individuals are not continuously engaged in care,
94 and thus, their viral level is not brought under control by ART (Figure 2). The critical
95 point of the USA cascade in 2011 would thus be a need to ensure continued provision
96 of care for diagnosed HIV infected individuals.

97 To date, a part of published empirical evidence indicated that widespread
98 ART has led to reductions in nearly all aspects of HIV/AIDS. For instance, expanded
99 ART in Canada has been shown to be associated with decreased morbidity, mortality
100 and HIV transmission, demonstrating that the combination of HIV testing and ART
101 programs in Canada has had a promising and profound population impact [12]. On the
102 other hand, while the reduced infectiousness has been shown to decrease HIV
103 incidence, the ART certainly increases the life expectancy of people living with
104 HIV/AIDS (PLWHA) and can sometimes increase the prevalence of HIV over time
105 [13]. A more recent study has indicated that even the reduction in HIV incidence is
106 not necessarily promised by test-and-treat program, especially if a part of 90-90-90
107 goal is not satisfied [14]. The importance of comprehensively understanding the pros
108 and cons of treatment-as-prevention strategy is increasingly recognized. Here we
109 introduce a simple mathematical model, based on Figure 1, to understand such
110 controversy in the next section.

111 **Transmission dynamics of HIV under treatment-as-** 112 **prevention**

113 Here we consider a simple mathematical model to understand how test-and-treat
114 influences the population dynamics of HIV/AIDS. First, we divide the population into
115 susceptible individuals, infected individuals without AIDS (H) and those who have
116 been diagnosed as AIDS (A). Population H and A are further divided into undiagnosed
117 (H_u and A_u) and diagnosed (H_d and A_d) groups. Four compartments of HIV infected
118 individuals have been schematically illustrated in Figure 1. At least in this model, we
119 assume that all diagnosed individuals are brought to be under ART.

120 Susceptible individuals experience infection with a rate $\lambda(t)$ which is a
121 function of infectious individuals H_u , A_u , H_d and A_d . We assume that ART reduces
122 one’s infectiousness on a whole from β to $\varepsilon\beta$ where parameter ε takes a value
123 between zero and one, and the value $1 - \varepsilon$ represents the relative reduction in the
124 transmissibility. Such reduction may not only be attributed to direct effectiveness of
125 treatment, but also caused by awareness of infection status and reduced frequency of
126 risky sexual intercourse. Without treatment, infected individuals are assumed to
127 develop AIDS with a progression rate ρ . HIV infected individuals under ART
128 progresses to AIDS with a far smaller rate $\gamma\rho$ where the value of $1 - \gamma$ would be
129 between zero and one and $\frac{1}{\gamma\rho} - \frac{1}{\rho}$ scales the average gain of the extended time without
130 AIDS. In addition to the natural death rate, μ , AIDS patients experience a higher
131 mortality rate than HIV infected individuals, because of disease induced death rate δ .

132 Parameter α represents the rate of diagnosis among HIV infected individuals, and $1/\alpha$
133 gives the average waiting time for diagnosis.

134 The model is written as the system of ordinary differential equations.

$$135 \frac{dH_u}{dt} = \lambda(t)(1 - H_u(t) - A_u(t) - H_d(t) - A_d(t)) - (\alpha + \rho + \mu)H_u(t),$$

$$136 \frac{dA_u}{dt} = \rho H_u(t) - (\mu + \delta)A_u(t),$$

$$137 \frac{dH_d}{dt} = \alpha H_u(t) - (\gamma\rho + \mu)H_d(t),$$

$$138 \frac{dA_d}{dt} = \gamma\rho H_d(t) - (\mu + \delta)A_d(t),$$

139 where the force of infection $\lambda(t)$ is given by

$$140 \lambda(t) = \beta H_u(t) + \varepsilon\beta H_d(t).$$

141 It should be noted that the transmission rate β reflects not only the infectiousness per
142 contact but also the rate of sexual contact per unit time. To understand the concept of
143 treatment-as-prevention in the simplest manner, the model presented here has ignored
144 gender and details of sexual partnership. Since AIDS patients are aware of their own
145 infection status, we do not account for the infectiousness of AIDS patients for
146 simplicity.

147 In the absence of diagnosis and treatment, the basic reproduction number, R_0 ,
148 the average number of secondary cases generated by a single primary case in a fully
149 susceptible population, is given by linearizing the abovementioned system nearby the
150 disease-free equilibrium, and we get

$$151 R_0 = \frac{\beta}{\rho + \mu}.$$

152 In the presence of diagnosis and treatment, the effective reproduction number, R_c , the
153 average number of secondary cases generated by a single primary case under test-and-
154 treat policy is similarly derived as

$$155 R_c = \frac{\beta}{\alpha + \rho + \mu} + \frac{\varepsilon\beta}{\gamma\rho + \mu} \frac{\alpha}{\alpha + \rho + \mu}.$$

156 To assess the test-and-treat strategy, a number of important and different
157 epidemiological metrics have been quantified, e.g. common indicators include (i) the
158 effective reproduction number, (ii) the incidence and prevalence given as the solution
159 of the above mentioned system and (iii) the cost-effectiveness ratio as informed by the
160 model outcome.

161 Different screening approaches would lead to different population outcomes.
162 Such differing patterns of screening could arise in many ways, e.g. different
163 frequency of HIV testing in the population, the use of advanced molecular techniques
164 to detect those in the window period, targeted testing of high risk groups and different
165 HIV infection stage (e.g. time since infection) to start treatment. Granich et al. [2]
166 compared the cost of the so-called “opt-in” and “opt-out” strategies of testing. Opt-in
167 strategy assumes that every infected individual presents to health services and starts
168 ART at CD4+ count 350 cell/mL. Opt-out strategy assumes yearly universal
169 voluntary testing of all individuals in the population, which is followed by immediate
170 ART upon diagnosis of HIV infection. The study has shown that the cost of opt-in
171 strategy will continue to increase whereas the cost of opt-out strategy would
172 eventually decrease with a success of controlling HIV/AIDS at the population level.

173 The suggested opt-out strategy is expected to eliminate HIV within 10 years
174 and the reality on that point has been subject to debate. Granich et al. [2] and
175 Kretzschmar et al. [15] mathematically derived the elimination threshold and studied
176 the conditions of treatment which makes the elimination of HIV feasible, such as the
177 frequency of testing, test coverage or an initiation time of the ART. Figure 3 shows a
178 simulation result of epidemic scenarios using the abovementioned equation system.

179 Sensitivity of the effective reproduction number and PLWHA as a function of the rate
180 of diagnosis α is examined. Given that the rate of diagnosis is greater than a certain
181 threshold to lead to $R_c < 1$, the test-and-treat is proven to successfully control the HIV
182 epidemic. The successful control endorses the global slogan of 90-90-90 strategy,
183 targeting high enough diagnosis and treatment coverage to ensure substantial public
184 benefit of HIV/AIDS.

185 Important pitfalls of test-and-treat are mainly seen in its long-term effects.
186 For instance, the prevalence of HIV infection is not necessarily promised to decrease.
187 Shafer et al. [16] estimated the population impact of ART in the future accounting for
188 the change in the turnover rate of sexual partnership under ART. The model expected
189 that ART will reduce the HIV incidence, while the HIV prevalence may be increased.
190 Figure 4 compares two simple scenarios, i.e., long term dynamics with and without
191 test-and-treat policy, comparing HIV incidence and prevalence. Meeting certain
192 mathematical conditions (especially, with large α and ϵ), both HIV incidence and
193 prevalence would decrease with time. Nevertheless, HIV prevalence in the presence
194 of test-and-treat could exceed that without any control if the relative transmissibility
195 of infected individuals under treatment is not sufficiently small. With the increased
196 HIV prevalence, it follows that testing every year and immediate treatment upon
197 diagnosis is not necessarily the most cost-efficient strategy and could even increase
198 long-term ART costs [17]. Theoretically, such controversial increase can be avoided
199 by reducing the transmissibility for those who are diagnosed, for example, by
200 ensuring high effectiveness of treatment, or by reducing the frequency of risky sexual
201 intercourse after awareness of the infection state. Increase in HIV prevalence also
202 indicates that the impact of test-and-treat should not be assessed by only a single
203 epidemiological indicator, and multiple aspects of the epidemiology have to be
204 carefully examined, especially using the effective reproduction number or elimination
205 threshold.

206 In relation to the population impact, the HIV infection stage at the start of
207 treatment has attracted researchers' attentions [18], because the population impact of
208 ART would be maximize if infected individuals are diagnosed at the very early stage
209 of infection. In addition, at a late infection-age of HIV, the frequency of sexual
210 contact is smaller than those in earlier stages [19].

211 **Future considerations**

212 While many mathematical modeling studies exist, all have certainly agreed that
213 increased diagnostic testing coupled with ART would induce a certain level of herd
214 immunity to the population. Mathematical modeling studies have found that model
215 assumptions, especially many properties of the sex partner network, would have a
216 profound impact on the incidence and prevalence, and incorporating local behavioral
217 data is considered to be critical [17].

218 Due to the need to satisfy high diagnostic coverage and treatment, it is
219 essential to first uncover the care cascade at each country level and locality.
220 Depending on risk populations, the diagnostic coverage may greatly differ due to
221 different awareness of risky behavior. Understanding the transmission dynamics in
222 the present day including the proportions of diagnosed, those followed-up and those
223 adhered to HIV, the topical question to answer may be to see if the effective
224 reproduction number is achieved to be the less than the value of one and if the
225 elimination threshold was met. Country-specific case studies have to be conducted to
226 confront with this task and understand the pros and cons with varying transmission
227 dynamics by country. Depending on the epidemiological context and the coverages of

228 cascade achieved, the optimal frequency of HIV testing is known to vary: opt-out
229 strategy with HIV testing every year is not always optimal [17].

230 Second, long-term epidemiological impact has yet to be explored in detail,
231 preferably along with empirical datasets. In the presence of continued effort of test-
232 and-treat approaches, HIV prevalence (or the number of PLWHA) and their life
233 expectancy are expected to increase. These observations are likely to lead to ageing of
234 infected individuals. Moreover, the aged infected individuals are more and more
235 likely to experience chronic diseases. Nevertheless, the failure to maintain high
236 coverage of care and adherence to ART could lead to dramatic resurgence of the
237 incidence and the surge of ART costs. Another critical issue in the context of long-
238 term impact is the emergence of drug resistant HIV, especially in resource-limited
239 countries with a struggle to maintaining adherence. Dose adherence remains to be the
240 key issue in such settings and continued monitoring of drug sensitivity would be
241 critical [20].

242 Not only leveraging the infrastructure and capacity for scaling up ART in
243 resource limited settings, but the scale-up of diagnostic and treatment coverages of
244 heterogeneous risk populations that are hard to reach are likely to be key issues at
245 practical settings [21]. Depending on epidemiological contexts of sexual mixing,
246 transmission dynamics (incidence/prevalence) and heterogeneous risk groups, realistic
247 quantitative approaches need to be sought supported by collaborations among
248 clinicians, public health practitioners, epidemiologists and mathematical modelers.

249 **Competing interests**

250 The authors declare that they have no competing interests.

251 **Authors' contributions**

252 HN conceived of the short review. KN constructed the model, performed simulations
253 and drafted the basis of manuscript as bullet points. HN wrote the original version of
254 the manuscript. XS and AI gave comments on earlier version of the manuscript. All
255 authors reviewed, revised and approved the final version of the manuscript.

256 **Acknowledgements**

257 HN and AI would like to thank the Health and Labour Sciences Research Grant (H28-
258 AIDS-General-001 and H26-AIDS-YoungInvestigator-004) for supporting this study.
259 HN received funding support from the Japan Agency for Medical Research and
260 Development and the Japan Science and Technology Agency (JST) CREST program
261 and RISTEX program for Science of Science, Technology and Innovation Policy. XS
262 and HN acknowledge the Program for Advancing Strategic International Networks to
263 Accelerate the Circulation of Talented Researchers, supported by the Japan Society
264 for the Promotion of Science. The funders had no role in study design, data collection
265 and analysis, decision to publish, or preparation of the manuscript.
266

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340 **Figures**

341 **Figure 1 - Flow chart of a simple compartmental model**

342 Variable H_u [H_d] is a fraction of undiagnosed [diagnosed] HIV-infected individuals
 343 without AIDS, A_u [A_d] is a fraction of previously undiagnosed [diagnosed] AIDS
 344 cases.

345 **Figure 2 - HIV care continuum in the United States, 2011**

346 Estimated percentages of persons living with HIV infection are shown [11]. In 2011,
 347 an estimated 1.2 million persons were living with HIV infection in the United States.

348 **Figure 3 - Test-and-treat with high screening rate may lead to the elimination** 349 **of HIV**

350 When the rate of diagnosis is greater than a certain threshold value, test-and-treat can
 351 successfully control HIV epidemic. Parameter values are $\mu=1/60$, $\rho=1/10$, $\gamma=1/3$,
 352 $\beta=0.15$, $\delta=1/2$ and $\varepsilon=0.3$.

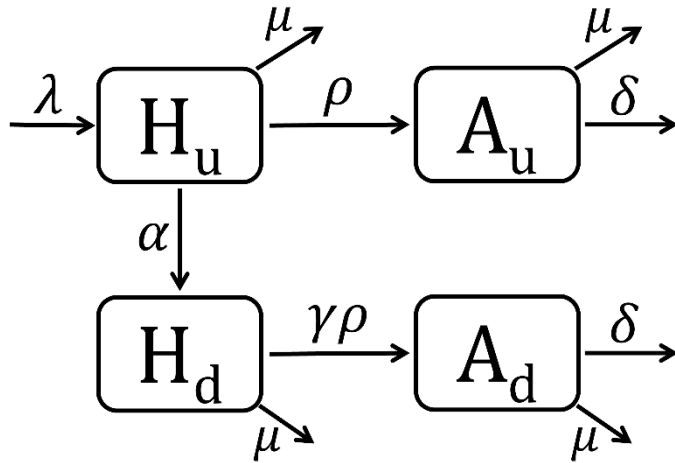
353 **Figure 4 - Test-and-treat could increase HIV prevalence**

354 (a, c) The rate of change in HIV incidence, (b, d) the proportion of the PLWHA
 355 (people living with HIV/AIDS). Without test-and-treat policy, the rate of diagnosis
 356 was set as $\alpha=0$. Under the test-and-treat policy, $\alpha=0.3$ was adopted. Parameter values
 357 are $\mu=1/60$, $\rho=1/10$, $\gamma=1/3$, $\beta=0.15$, $\delta=1/2$ and $\varepsilon=0.3$. The test-and-treat reduces both
 358 the incidence and the prevalence in (a) and (b). For panel (c) and (d), $\varepsilon=0.5$ was used
 359 instead of $\varepsilon=0.3$ as the relative transmissibility for those who are diagnosed. In this
 360 scenario, test-and-treat increases HIV prevalence. Initial values are $H_u=0.15$, $A_u=0.01$,
 361 $H_d=0$ and $A_d=0$.

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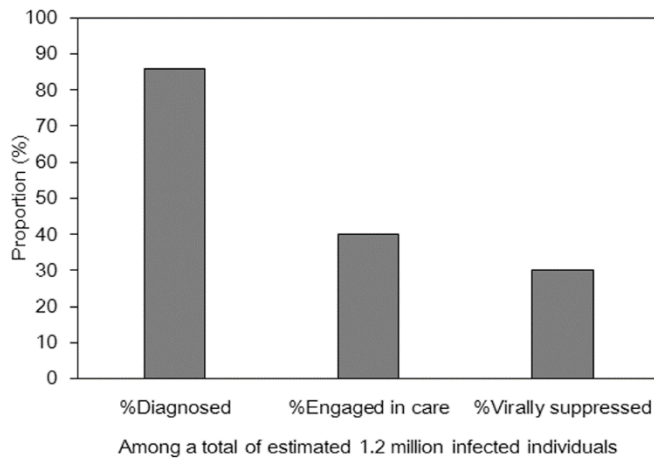
363 Embedded figures

364 Figure 1



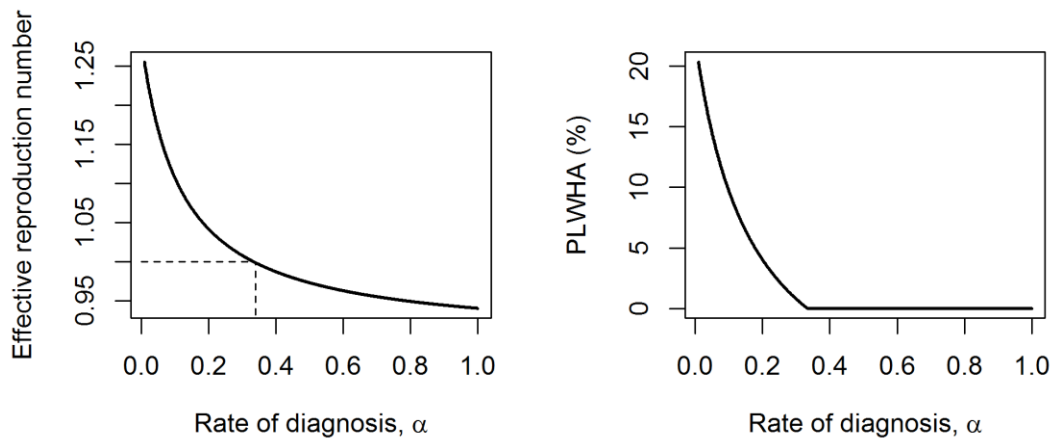
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Figure 2



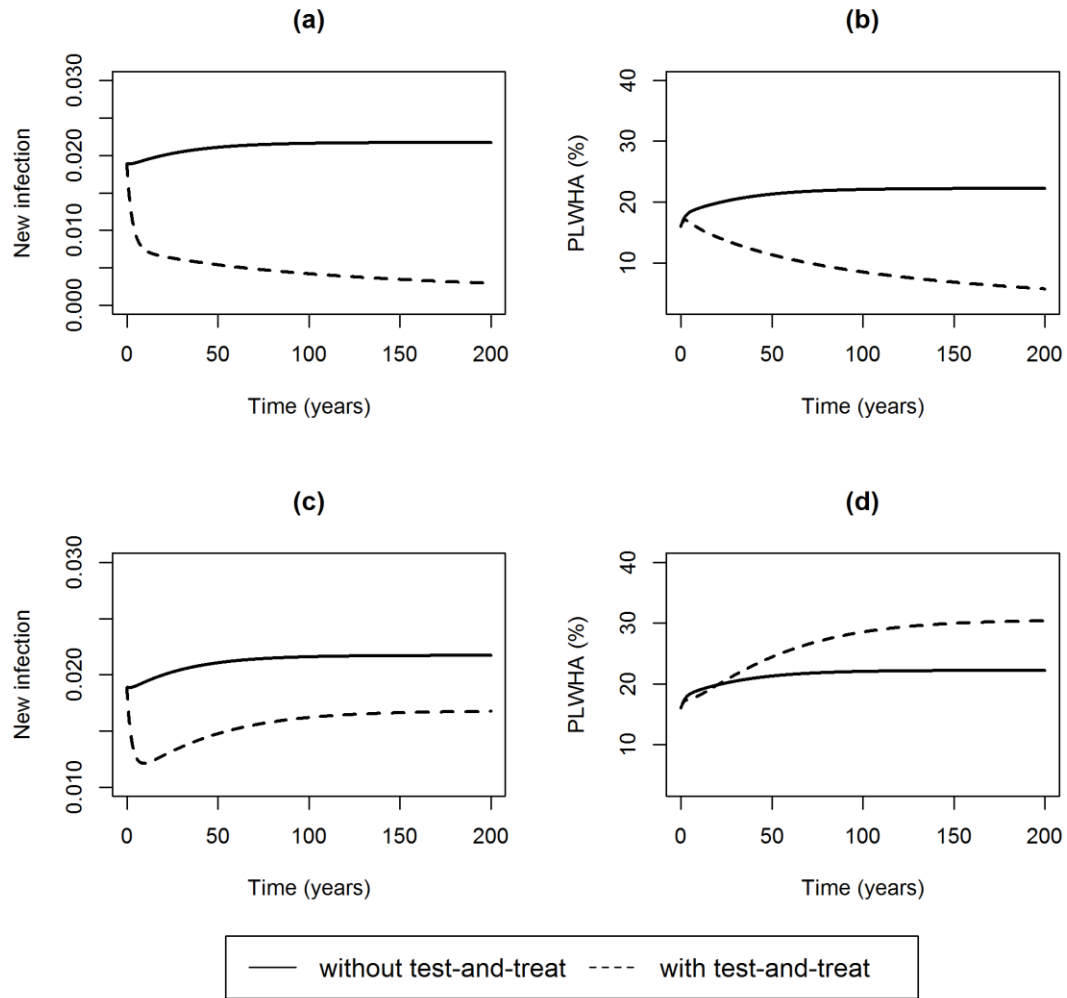
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Figure 3



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Figure 4



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