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Abstract

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18 Introduction

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38 Models and Methods

39 **Data.** We use daily confirmed case data $\mathbf{X} = (X_1, \dots, X_T)$ from several different countries. These
 40 incidence data summarize the number of individuals X_t tested positive for SARS-CoV-2 RNA (using
 41 RT-qPCR) at each day t . Data was retrieved from **REFS** for the following regions: Wuhan, China;
 42 Austria; ...

43 **SEIR model.** We model SARS-CoV-2 infection dynamics by following the number of susceptible
 44 S , exposed E , reported infected I_r , and unreported infected I_u individuals in a population of size N .
 45 This model distinguishes between reported and unreported infected individuals: the reported infected
 46 are those that have enough symptoms to eventually be tested and thus appear in daily case reports, to
 47 which we fit the model.

48 Susceptible (S) individuals become exposed due to contact with reported or unreported infected
 49 individuals (I_r or I_u) at a rate β_t or $\mu\beta_t$. The parameter $0 < \mu < 1$ represents the decreased transmission
 50 rate from unreported infected individuals, who are often subclinical or even asymptomatic. The
 51 transmission rate $\beta_t \geq 0$ may change over time t due to behavioral changes of both susceptible and
 52 infected individuals. Exposed individuals, after an average incubation period of Z days, become
 53 reported infected with probability α_t or unreported infected with probability $(1 - \alpha_t)$. The reporting
 54 rate $0 < \alpha_t < 1$ may also change over time due to changes in human behavior. Infected individuals
 55 remain infectious for an average period of D days, after which they either recover, or becomes ill
 56 enough to be quarantined. They therefore no longer infect other individuals, and the model does not
 57 track their frequency. The model is described by the following equations:

$$\begin{aligned}
 \frac{dS}{dt} &= -\beta_t S \frac{I_p}{N} - \mu\beta_t S \frac{I_u}{N} \\
 \frac{dE}{dt} &= \beta_t S \frac{I_p}{N} + \mu\beta_t S \frac{I_u}{N} - \frac{E}{Z} \\
 \frac{dI_r}{dt} &= \alpha_t \frac{E}{Z} - \frac{I_r}{D} \\
 \frac{dI_u}{dt} &= (1 - \alpha_t) \frac{E}{Z} - \frac{I_u}{D}.
 \end{aligned}
 \tag{1}$$

59 The initial numbers of exposed $E(0)$ and unreported infected $I_u(0)$ are considered model parameters,
 60 whereas the initial number of reported infected is assumed to be zero $I_r(0) = 0$, and the number of
 61 susceptible is $S(0) = N - E(0) - I_u(0)$. The vector θ of model parameters is

$$\theta = \left(Z, D, \mu, \{\beta_t\}, \{\alpha_t\}, \{p_t\}, E(0), I_u(0) \right).
 \tag{2}$$

63 This model is inspired by Li et al. (2020) and Pei & Shaman (2020), who used a similar model with
 64 multiple regions and constant transmission β and reporting rate α to infer COVID-19 dynamics in
 65 China and the continental US, respectively.

66 **Likelihood function.** The *expected* cumulative number of reported infected individuals until day t
 67 is

$$Y_t = \int_0^t \alpha_s \frac{E(s)}{Z} ds, \quad Y_0 = 0.
 \tag{3}$$

We assume that reported infected individuals are confirmed and therefore observed in the daily case
 report of day t with probability p_t (note that an individual can only be observed once, and that p_t may
 change over time, but t is a specific date rather than the time elapsed since the individual was infected).
 Hence, we assume that the number of confirmed cases in day t is binomially distributed,

$$X_t \sim \text{Bin}(n_t, p_t),$$

where n_t is the *realized* (rather than expected) number of reported infected individuals yet to appear in daily reports by day t . The cumulative number of confirmed cases until day t is

$$\tilde{X}_t = \sum_{i=1}^t X_i, \quad X_0 = 0.$$

Given \tilde{X}_{t-1} , we assume n_t is Poisson distributed,

$$(n_t | \tilde{X}_{t-1}) \sim \text{Poi}(Y_t - \tilde{X}_{t-1}), \quad n_1 \sim \text{Poi}(Y_1).$$

Therefore, $(X_t | \tilde{X}_{t-1})$ is a binomial conditioned on a Poisson, which reduces to a Poisson with

$$(X_t | \tilde{X}_{t-1}) \sim \text{Poi}((Y_t - \tilde{X}_{t-1}) \cdot p_t), \quad X_1 \sim \text{Poi}(Y_1 \cdot p_1). \quad (4)$$

For given vector θ of model parameters (Eq. (2)), we compute the expected cumulative number of reported infected individuals $\{Y_t\}_{t=1}^T$ for each day (Eq. (3)). Then, since \tilde{X}_{t-1} is a function of X_1, \dots, X_{t-1} , we can use Eq. (4) to write the probability to observe the confirmed case data $\mathbf{X} = (X_1, \dots, X_T)$ as

$$\mathbb{L}(\theta | \mathbf{X}) = P(\mathbf{X} | \theta) = P(X_1 | \theta)P(X_2 | \tilde{X}_1, \theta) \cdots P(X_T | \tilde{X}_{T-1}, \theta). \quad (5)$$

This defines a *likelihood function* $\mathbb{L}(\theta | \mathbf{X})$ for the parameter vector θ given the data \mathbf{X} .

NPI model. To model non-pharmaceutical interventions (NPIs), we set the beginning of the NPIs to day τ and define

$$\beta_t = \begin{cases} \beta, & t < \tau \\ \beta\lambda, & t \geq \tau \end{cases}, \quad \alpha_t = \begin{cases} \alpha_1, & t < \tau \\ \alpha_2, & t \geq \tau \end{cases}, \quad p_t = \begin{cases} 1/9, & t < \tau \\ 1/6, & t \geq \tau \end{cases},$$

where $0 < \lambda < 1$. The values for p_t follow Li et al. (2020), who estimated the average time between infection and reporting in Wuhan, China, at 9 days before the start of NPIs (Jan 23, 2020) and 6 days after start of NPIs. The parameter τ is then added to the parameter vector θ (Eq. (2)).

Model fitting. To fit our model (Eq. (1)) to the data \mathbf{X} and estimate the model parameters θ , we apply a Bayesian inference approach. We define the following flat priors on the model parameters $P(\theta)$:

$$\begin{aligned} Z &\sim \text{Uniform}(2, 5) \\ D &\sim \text{Uniform}(2, 5) \\ \mu &\sim \text{Uniform}(0.2, 1) \\ \beta &\sim \text{Uniform}(0.8, 1.5) \\ \lambda &\sim \text{Uniform}(0, 1) \\ \alpha_1, \alpha_2 &\sim \text{Uniform}(0.02, 1) \\ E(0) &\sim \text{Uniform}(0, 3000) \\ I_u(0) &\sim \text{Uniform}(0, 3000) \\ \tau &\sim \text{Uniform}(1, T - 1), \end{aligned} \quad (6)$$

where T is the number of days in the data \mathbf{X} . Most priors follow Li et al. (2020), except λ , which is used to enforce that the transmission rates are lower after the start of the NPIs ($\lambda < 1$). The likelihood function is defined in Eq. (5). The posterior distribution on the model parameters $P(\theta | \mathbf{X})$ is then estimated using an affine-invariant ensemble sampler for Markov chain Monte Carlo (MCMC) implemented in the `emcee` Python package (Foreman-Mackey et al. 2013).

89 Discussion

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108 As several countries (e.g. Austria, Israel) begin to relieve lockdowns and ease restrictions, we
109 expect similar delays and advances to occur: in some countries people will begin to behave as if
110 restrictions were eased before the official date, and in some countries people will continue to self-
111 restrict even after restrictions are officially removed. Such delays and advances could confuse analyses
112 and lead to wrong conclusions about the effects of NPI removals.

113 **Acknowledgements**

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115 References

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