

Mathematical models in immunology: weekly neutralizing antibodies, antibody dependent enhancement and reinfection

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Outline

- 1 Motivation
- 2 Full model construction
- 3 Stability results
- 4 Beyond stability: dynamics
- 5 Numerical simulations
- 6 Conclusions and future work

Motivation: host-virus-immune system model

- GOAL: : mathematical, numerical and clinical study of the host-virus-immune system interaction
- BASED ON: joint paper with A. Danchin, O. Pagani, & G. Yahiaoui [7] ; further information, slides: available at <https://turinici.com>
- Important ingredients (cf. COVID-19 and not only):
 - ① re-infection,
 - ② variable antibody neutralization capacity
 - ③ antibody disease enhancement (ADE: pediatric cases and not only)

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Full model construction

- compartmental model (cf. [13, 8, 12, 17, 6, 9, 5])
- the viral-host interaction (w/o immune response) = basic model of virus dynamics (see [18, eq (3.1) page 18], [22, eqns. (2.3)-(2.4) page 26], [16, 4, 10])

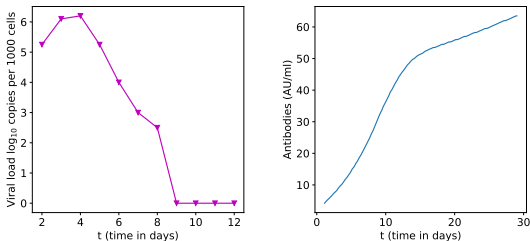
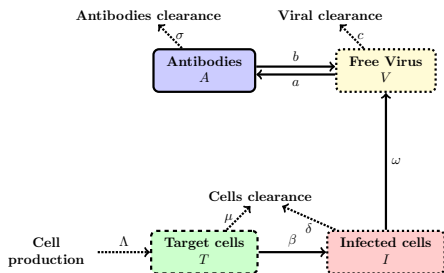


Figure: **Left:** Clinically observed typical variation of SARS-CoV-2 viral load in nasopharyngeal swab normalised using cell quantification. Data taken from [14, figure 3 page 703, patient 4]. **Right:** : Typical time variations for IgG. Data taken from [19, figure 2 page 1085]. **Note that the antibody data is a mean over several days and corresponds to a different patient cohort.**

Full model construction

- T : target cells, I : infected cells, V : the free virus, A : the antibodies, cf. fig 2.
- $\delta > \mu$, cf. [22, 18];
- the immune reaction starts at the threshold $V^t = \sigma/a$; alternative proposals for dA/dt possible (e.g. André and Gandon [1], Pawlek et al. [20])



$$dT/dt = \Lambda - \mu T - \beta(A)V T \quad (1)$$

$$dI/dt = \beta(A)V T - \delta I \quad (2)$$

$$dV/dt = \omega I - cV - bAV \quad (3)$$

$$dA/dt = aVA - \sigma A \quad (4)$$

$$\beta(A) = \beta_0 + \beta_1 A. \quad (5)$$

Figure: Flow in the model (1)-(5).

Building blocks

- No infection, target cells only (see [22, 18]): $dT/dt = \Lambda - \mu T$, cv. to stable equilibrium $T^* = \Lambda/\mu$.
- Model with virus but no immune response: basic model of virus dynamics, see [18, eq (3.1) page 18],[22, eqns. (2.3)-(2.4) page 26]

$$dT/dt = \Lambda - \mu T - \beta_0 VT, \quad (6)$$

$$dI/dt = \beta_0 VT - \delta I, \quad (7)$$

$$dV/dt = \omega I - cV. \quad (8)$$

$$\text{Initial conditions : } T(0) = T^* = \Lambda/\mu, \quad I(0) = 0, \quad V(0) > 0, \quad (9)$$

$$\text{Assumption 1: } \delta > \mu. \quad (10)$$

$$\text{Infection only if : } R_0 = \frac{\beta_0 \omega \Lambda}{c \delta \mu} > 1. \quad (11)$$

$$dT/dt = \Lambda - \mu T - \beta_0 VT, \quad (12)$$

$$dI/dt = \beta_0 VT - \delta I, \quad (13)$$

$$dV/dt = \omega I - cV. \quad (14)$$

Equilibrium no. 1: trivial equilibrium: $T = T^* = \Lambda/\mu$, $V = I = 0$.

Jacobian $\begin{pmatrix} -\mu & 0 & -\beta_0 T^* \\ 0 & -\delta & \beta_0 T^* \\ 0 & \omega & -c \end{pmatrix}$. The eigenvalues are all real but not all

negative: one of them is $\lambda_1 = -\mu$ but the product of the other two is $\delta c - \omega \beta_0 T^* \leq 0$ thus at least one is positive. Therefore, **under assumption (11), ($R_0 > 1$) this critical point is not a stable equilibrium.**

Equilibrium no. 2 : the "immunosuppression" equilibrium (15)

$$T = T^{is} = \frac{\delta c}{\beta_0 \mu}, \quad V = V^{is} := (R_0 - 1) \frac{\mu}{\beta_0}, \quad I = I^{is} := (R_0 - 1) \frac{c \mu}{\omega \beta_0}. \quad (15)$$

the characteristic polynomial of the Jacobian

$P(X) = (X + \delta)(X + c)(X + \mu + \beta_0 V^{is}) - \delta c(X + \mu)$ has all roots with negative real part. Therefore the equilibrium is stable.

V^{is} = viral load in a completely immunodeficient individual. We expect V^{is} to be very high, in particular higher than the threshold $V^t = \sigma/a$ for the initiation of the immune response. This motivates

$$\text{Assumption 2:} \quad (R_0 - 1) \frac{\mu}{\beta_0} > \frac{\sigma}{a}. \quad (16)$$

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Model: virus and immune response but no enhancement

$$\beta_1 = 0$$

Proposition (GT et al 2022)

Under assumptions 1 et 2, the model w/o ADE ($\beta(A) = \beta_0$)

$$dT/dt = \Lambda - \mu T - \beta(A)VT \quad (17)$$

$$dI/dt = \beta(A)VT - \delta I \quad (18)$$

$$dV/dt = \omega I - cV - bAV \quad (19)$$

$$dA/dt = aVA - \sigma A \quad (20)$$

has a single stable equilibrium given by:

$$T = T^{is} := \frac{\Lambda}{\mu + \beta_0 V^t}, \quad I = \frac{\beta_0 \Lambda V^t}{\delta(\mu + \beta_0 V^t)}, \quad V = V^t, \quad A = \frac{c(V^{is} - V^t)}{\beta_0 b(\mu + \beta_0 V^t)}. \quad (21)$$

Proof of the stability w/o ADE ($\beta_1 = 0$)

- In general this is **very difficult (sometimes impossible)** to check (highly nonlinear); very few general approaches exist to study theoretically stability w/r to parameters for a 4-dim system ... existing works by M. Li and J. Muldowney [15] based on geometric approaches like Poincare-Bendixon (non- \exists of chaotic solutions), also Ballyk et al. [2], Bruono Buonomo & Deborah Lacitignola [3], etc.
- there are 3 equilibria, two of them with $A = 0$ (like in the model w/o immune system) which are unstable, as before
- only remaining equilibrium, the one in the proposition.
- Stability = eigenvalues of the Jacobian have negative real parts = $P_0(X) = \det(X \cdot I - J)$ is stable
- Denote $P_0(X) = \gamma_4 X^4 + \gamma_3 X^3 + \gamma_2 X^2 + \gamma_1 X^1 + \gamma_0$
- Routh-Hurwitz criterion [21], [11, p. 1076], stability when

$$\gamma_k > 0, k = 0, 1, 2, 3, 4 \quad (22)$$

$$\gamma_1 \gamma_2 \gamma_3 > \gamma_4 \gamma_1^2 + \gamma_3^2 \gamma_0. \quad (23)$$

Proof of the stability w/o ADE ($\beta_1 = 0$)

- Idea: change of variables: $\zeta = \delta - \mu > 0$, $w = R_0 - 1 - \frac{\sigma\beta_0}{\mu a} > 0$.

Then $\gamma_0 = c\delta\mu\sigma w$, $\gamma_1 = \frac{c\sigma(a^2\delta\mu w + a^2\mu^2 w + a\beta_0\delta\mu w + a\beta_0\delta\mu + a\beta_0\mu\sigma w + \beta_0^2\delta\sigma)}{a(a\mu + \beta_0\sigma)}$,

$$\gamma_2 = \frac{a^2c\mu^2 w + a^2c\mu^2 + a^2c\mu\sigma w + a^2\delta\mu^2 + a\beta_0c\mu\sigma w + 2a\beta_0c\mu\sigma + 2a\beta_0\delta\mu\sigma + \beta_0^2c\sigma^2 + \beta_0^2\delta\sigma^2}{a(a\mu + \beta_0\sigma)}$$

$$\gamma_3 = \frac{a^2\delta\mu + a^2\mu^2 + a\beta_0\delta\sigma + 2a\beta_0\mu\sigma + ac(a\mu(w+1) + \beta_0\sigma) + \beta_0^2\sigma^2}{a(a\mu + \beta_0\sigma)}, \quad \gamma_4 = 1.$$

condition (22) ($\gamma_i > 0$, $i = 0, \dots, 4$) = OK

condition (23) : $\gamma_1\gamma_2\gamma_3 > \gamma_4\gamma_1^2 + \gamma_3^2\gamma_0$

$$\gamma_1\gamma_2\gamma_3 - (\gamma_4\gamma_1^2 + \gamma_3^2\gamma_0) = \frac{ac}{a^3(a\mu + \beta_0\sigma)^3} Q_0(w, a, c, \beta_0, \mu, \sigma, \zeta), \quad (24)$$

$Q_0(w, a, c, \beta_0, \mu, \sigma, \zeta) = w^3 a^6 c^2 \mu^5 + 2w^3 a^6 c^2 \mu^4 \sigma + w^3 a^6 c^2 \mu^3 \sigma \zeta + w^3 a^5 c^2 \beta_0 \mu^5 + 3w^3 a^5 c^2 \beta_0 \mu^4 \sigma + w^3 a^5 c^2 \beta_0 \mu^4 \zeta + \dots$ (cca 170 terms !)

Negative coefficients monomials are combined with other e.g.

$$w^3 a^6 c^2 \mu^3 \sigma \zeta / 2 - w^2 a^5 c \beta_0 \mu^3 \sigma \zeta^2 + w a^4 \beta_0^2 \mu^3 \sigma \zeta^3 / 2 = \frac{w a \mu^3 \sigma \zeta}{2} (w a c - \beta_0 \zeta)^2 \geq 0, \text{ etc. } \dots, \text{ which concludes the proof.}$$

Proposition (GT et al. 2021)

The full model (1)-(4) has three equilibria:

- 1 the trivial equilibrium $T = T^* = \Lambda/\mu$, $V = I = A = 0$ which is unstable;
- 2 the immunosuppression equilibrium, also unstable :
$$T = T^{is} = \frac{\delta c}{\beta_0 \mu}, \quad V = V^{is} := (R_0 - 1) \frac{\mu}{\beta_0}, \quad I = I^{is} := (R_0 - 1) \frac{c\mu}{\omega \beta_0}.$$
- 3 and a third equilibrium: $A^f =$ unique positive solution of
$$\omega \beta(A) \Lambda = \delta(c + bA)(\mu + \beta(A)V^t), \quad T^f = \frac{\delta(c + bA^f)}{\omega \beta(A^f)}, \quad I^f = \frac{V(c + bA^f)}{\omega},$$

$$V = V^t = \frac{\sigma}{a}. \quad \text{Then :}$$
 - 1 when β_1 is small enough this equilibrium is stable;
 - 2 when β_1 is large enough this equilibrium is stable;
 - 3 however there exist choices of parameters (in particular values of β_1) for which this equilibrium is unstable.

Proof idea:

- 1 for β_1 small enough we are near the situation $\beta_1 = 0$ treated previously
- 2 we can find numerically some counter-examples for intermediate values of β_1 : for $a = \sigma = c = b = \omega = 1$, $\mu = 1.e - 3$, $\delta = 2$, $\Lambda = 4$, $\beta_0 = 0.0011$ and $\beta_1 = 0.01188$ all hypotheses are satisfied, equilibrium values are $T = 333.33$, $I = 1.83$, $V = 1$, $A = 0.83$, eigenvalues are -3.45 , 0.50 , 0.01 and -0.90 : unstable (at least one is positive).
- 3 for $\beta_1 \rightarrow \infty$ (others kept fixed) we can use similar techniques as in the previous proof (cf. [7] for details).

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Dynamical aspects

- The equilibrium does not yet tell the full story of the evolution of the system (1)-(4).
- possible behavior: initially A will increase in reaction to $V > V^t$; this drives both I and V to small values : stochastic effects lead to complete remission (virus clearance) $T \rightarrow \Lambda/\mu$, $A \rightarrow 0$;
- If however during the slow decay of A a challenge is presented in the form of a virus load $V > V^t$ a new infection will start and V and I will rise again.
- a distinct possibility: before reaching the stable equilibrium T goes to very low values (severe respiratory difficulties, other organ destruction) or A to high values (autoimmune presentations, organ destruction): infection wins
- numerical simulations are needed in particular situations to explore the limits of the model; the precise dynamics depends crucially on the parameters b (antibody neutralization capacity) and β_1 (ADE)

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Empirical results: initial infection

Numerical procedure to fit the model parameters to reproduce at best the viral load data in fig. 1: resulting numerical simulation in fig. 3.

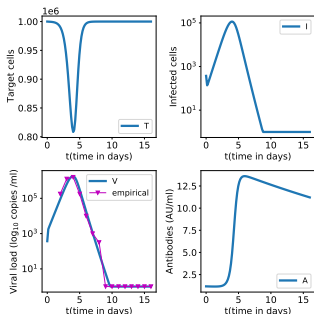


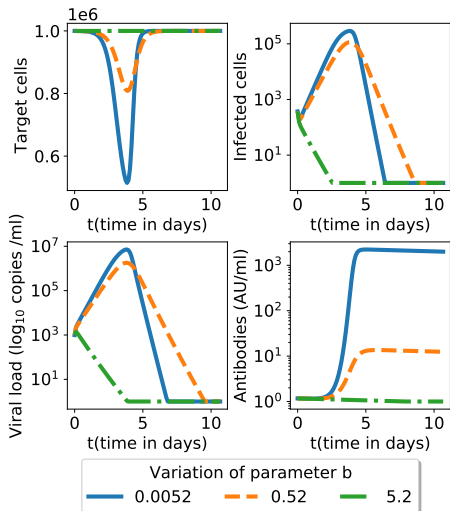
Figure: Numerical simulation of the first infection (no ADE) (1)-(4) for $\mu = 9.66$, $\Lambda = 9.66 \times 10^6$, $\omega = 59.74$, $\beta_0 = 1.28 \times 10^{-6}$, $\beta_1 = 0$, $\delta = 16.22$, $c = 1.45$, $b = 0.52$, $a = 9.15 \times 10^{-7}$, $\sigma = 0.02$, $I(0) = 372.11$, $V(0) = 994.84$, $A(0) = 1.17$.

- good fit $V(t)$, not so for $A(t)$ (joint $V(t)/A(t)$ data not available).
- 20% fall of target cells; $V(t)$ peaks around 4-5 days;
- equilibrium state (21) when $\beta_1 = 0$ (no ADE present) is reached after 2 years BUT $V(t)$ and $I(t)$ reach a minimum within several weeks then increase and oscillate toward the equilibrium state (21) (simulations not shown here): in practice the equilibrium state (21) is stable but not reached, the patient is cured before.

Empirical results: secondary infection, variants, vaccination

Scenario : the immuno-kinetic parameters (b , β_1) change (e.g., a primary infection with a different variant, vaccination, or some immune evolution, aging, ...).

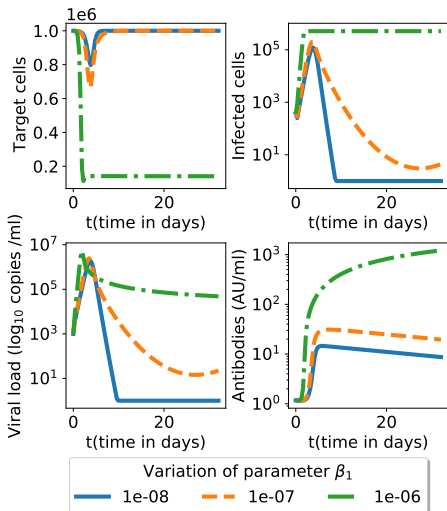
- Variation of the neutralizing capacity b changed around the nominal value $b = 0.52$ (no ADE):
 $b \downarrow$: $V \uparrow$, $A \uparrow$, massive T destruction, largely pejorative.



Empirical results: secondary infection, variants, vaccination

Scenario : the immuno-kinetic parameters (b , β_1) change (e.g., a primary infection with a different variant, vaccination, or some immune evolution, aging, ...).

• Presence of ADE ($\beta_1 > 0$) : then $\beta_1 \uparrow$ lead to $T \downarrow$ (significant, possibly total !), $I \uparrow$, $V \uparrow$, $A \uparrow$ (but limited); higher β_1 cannot be compensated by more antibodies, may lead to respiratory function disruption.



Longer time span dynamics

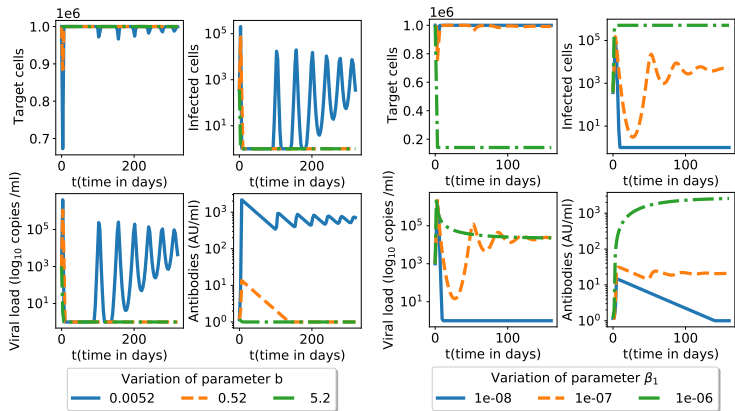


Figure: Left: Simulation in fig. 4 (b variation) for a longer time span. For the smallest value of b (blue line) the theoretic equilibrium is not reached in practice.

Right Simulation in fig. 5 (β_1 variation) for a longer time span.

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Conclusions and future work

Further questions:

- a technique that allows to analyze stability for a nonlinear system of 4 dimensions
- dynamical aspects are important (limits of the model)
- further work : finer description of the $A(t)$ part, add distinction between short term / long term immunity (e.g., IgM vs. IgG)
- test more real data

