



Wikno, 16th–20th September 2025

A CONTINUOUS-TIME *SIS* CRISS-CROSS MODEL OF CO-INFECTION IN A HETEROGENEOUS POPULATION

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ABSTRACT

In a population we indicate two subpopulations, a low-risk (*LS*) and a high-risk (*HS*) subpopulation, that relate to the risk of getting infected. *LS* and *HS* have accordingly lower and higher susceptibility to each disease. For every variable and parameter we assign a subscript i equal to 1 and 2 for *LS* and *HS*, respectively. If i has no assigned value, then $i \in \{1, 2\}$. By S_1 and S_2 we denote a density of healthy people in *LS* and *HS*, respectively. The variables I_i mean a density of individuals from the given subpopulation that are infected by a pathogen of disease which we call disease *A* (*DA*). Similarly, we define J_i as a density of individuals suffering from disease *B* (*DB*). A density of a group infected by pathogens from both diseases is denoted by K_i .

Migrating and newborn individuals join each subpopulation through S_i class with a recruitment rate C_i . A natural death rate for each subpopulation is equal to μ_i . For *DA* we introduces transmission rates: $\beta_{11}, \beta_{22}, \beta_{12}, \beta_{21}$ reflecting transmission: among *LS*, among *HS*, from *HS* to *LS* and from *LS* to *HS*, respectively. Indicating four different rates means that *DA* differs in spreading and contracting a pathogen. To get a preliminary insight on co-infection dynamics for the heterogeneous population, for *DB* we assume that individuals differs only in contracting a pathogen. For this reason we take only two transmission coefficients for *DB*: σ_1 for *LS* and σ_2 for *HS*. By γ_i and g_i we denote a recovery rate for *DA* and *DB*, respectively. The disease-mortality rate for *DA* and *DB* is depicted by α_i and a_i .

The proposed model of co-infection reads

$$\dot{S}_1 = C_1 - \beta_{11}S_1I_1 - \beta_{12}S_1I_2 + \gamma_1I_1 - \mu_1S_1 - \sigma_1S_1(J_1 + J_2) + g_1J_1, \quad (1a)$$

$$\dot{I}_1 = \beta_{11}S_1I_1 + \beta_{12}S_1I_2 - (\gamma_1 + \alpha_1 + \mu_1)I_1 - \sigma_1I_1(J_1 + J_2) + g_1K_1, \quad (1b)$$

$$\dot{J}_1 = \sigma_1S_1(J_1 + J_2) - (g_1 + a_1 + \mu_1)J_1 - \beta_{11}J_1I_1 - \beta_{12}J_1I_2 + \gamma_1K_1, \quad (1c)$$

$$\dot{K}_1 = \sigma_1I_1(J_1 + J_2) + \beta_{11}J_1I_1 + \beta_{12}J_1I_2 - (g_1 + a_1 + \gamma_1 + \alpha_1 + \mu_1)K_1, \quad (1d)$$

$$\dot{S}_2 = C_2 - \beta_{22}S_2I_2 - \beta_{21}S_2I_1 + \gamma_2I_2 - \mu_2S_2 - \sigma_2S_2(J_1 + J_2) + g_2J_2, \quad (1e)$$

$$\dot{I}_2 = \beta_{22}S_2I_2 + \beta_{21}S_2I_1 - (\gamma_2 + \alpha_2 + \mu_2)I_2 - \sigma_2I_2(J_1 + J_2) + g_2K_2, \quad (1f)$$

$$\dot{J}_2 = \sigma_2S_2(J_1 + J_2) - (g_2 + a_2 + \mu_2)J_2 - \beta_{22}J_2I_2 - \beta_{21}J_2I_1 + \gamma_2K_2, \quad (1g)$$

$$\dot{K}_2 = \sigma_2I_2(J_1 + J_2) + \beta_{22}J_2I_2 + \beta_{21}J_2I_1 - (g_2 + a_2 + \gamma_2 + \alpha_2 + \mu_2)K_2. \quad (1h)$$

Each parameter is fixed and positive. In particular, every parameter besides C_i is in the range $(0, 1)$. Figure 1 is a schematic drawing of the proposed model.

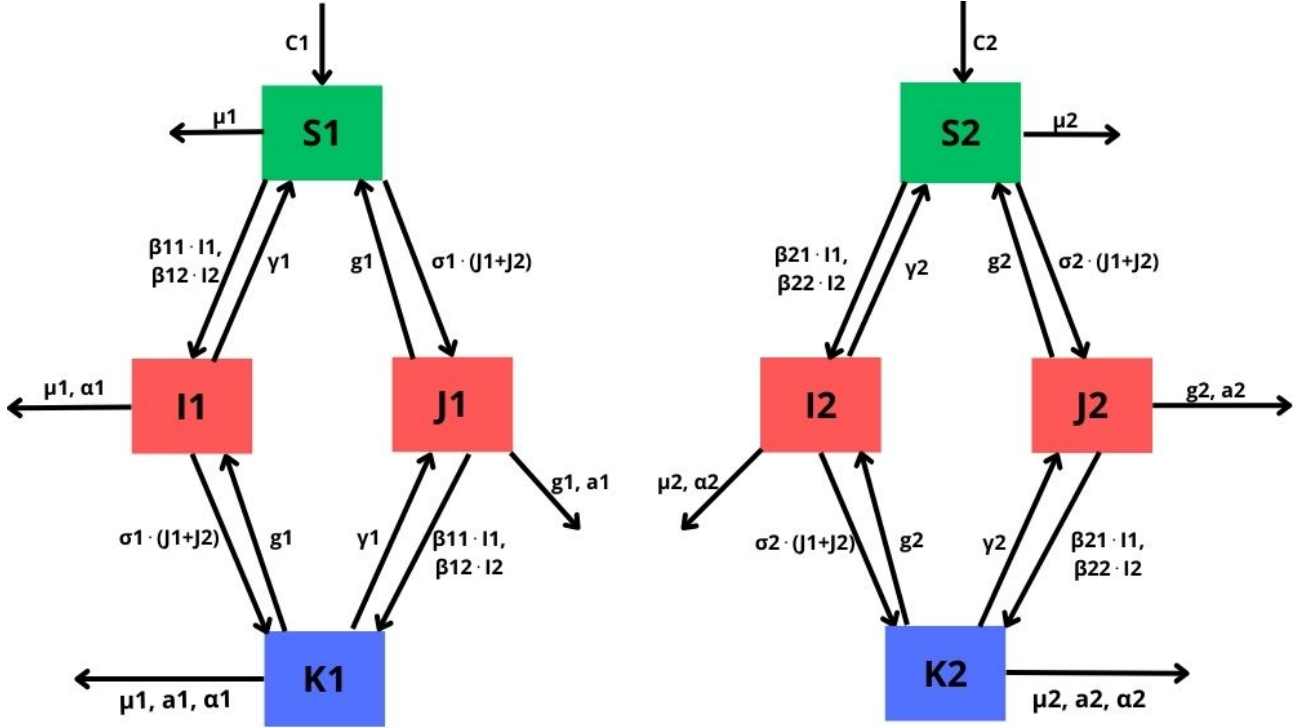


Figure 1. Possible movements between particular classes from system (1).

System (1) has four stationary states: disease-free (E_{df}), with sole DA or DB (E_A and E_B). We also suspect that there exists the endemic state (E_e), with two diseases present, exists, but we did not manage to prove it because of complicated computations. State E_{df} has the form

$$E_{df} = (\widehat{S}_1, 0, 0, 0, \widehat{S}_2, 0, 0, 0), \quad \text{where} \quad \widehat{S}_1 = \frac{C_1}{\mu_1}, \quad \widehat{S}_2 = \frac{C_2}{\mu_2}.$$

It exits unconditionally, while provided conditions determine the existence of E_A and E_B . For state E_e we only gave insight into its existence because of the complexity of the computations. For system (1) we computed the basic reproduction number \mathcal{R}_0 . This number can be written as

$$\mathcal{R}_0 = \max(\lambda_1, \lambda_2),$$

where

$$\lambda_1 = \frac{\sigma_1}{q_1} \widehat{S}_1 + \frac{\sigma_2}{q_2} \widehat{S}_2, \quad q_i := g_i + a_i + \mu_i$$

and

$$\lambda_2 = \frac{1}{2k_1k_2} \left(k_2\beta_{11}\widehat{S}_1 + k_1\beta_{22}\widehat{S}_2 + \sqrt{(k_2\beta_{11}\widehat{S}_1 - k_1\beta_{22}\widehat{S}_2)^2 + 4k_1k_2\beta_{12}\beta_{21}\widehat{S}_1\widehat{S}_2} \right),$$

where $k_i := \gamma_i + \alpha_i + \mu_i$. Later we investigated the local stability of the stationary state. State E_{df} is locally stable if $\mathcal{R}_0 < 1$, what is expected. Analysis of the local stability for E_A and E_B provided the list of conditions. What is important is that the parameters from both diseases affect the local stability of both states.

REFERENCES

- [1] M. Bodzioch, M. Choiński, and U. Foryś: *SIS criss-cross model of tuberculosis in heterogeneous population*, Discrete and Continuous Dynamical Systems – B **24** (2019), 2169–2188.