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A geometric analysis of the SIR, SIRS and SIRWS epidemiological models

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Abstract

We study fast-slow versions of the SIR, SIRS, and SIRWS epidemiological models. The multiple time scale behavior is introduced to account for large differences between some of the rates of the epidemiological pathways. Our main purpose is to show that the fast-slow models, even though in nonstandard form, can be studied by means of Geometric Singular Perturbation Theory (GSPT). In particular, without using Lyapunov's method, we are able to not only analyze the stability of the endemic equilibria but also to show that in some of the models limit cycles arise. We show that the proposed approach is particularly useful in more complicated (higher dimensional) models such as the SIRWS model, for which we provide a detailed description of its dynamics by combining analytic and numerical techniques.

Keywords: fast-slow system, epidemic model, non-standard form, entry-exit function, bifurcation analysis, numerical continuation.

1 Introduction

Epidemic modelling has grown from the pioneering 1927 article by Kermack and McKendrick [21] into a wide body of theory and applications to several diseases [1, 14, 19, 9, 28], used also for developing appropriate control strategies.

The model by Kermack and McKendrick [21] was of S-I-R type, meaning that individuals are classified as *Susceptibles* (S), *Infected* (I) or *Recovered* (R), and that the only possible transitions are $S \rightarrow I$ (new infection) and $I \rightarrow R$ (recovery with permanent immunity). As that model does not consider new births or deaths (other than because of the disease), it is appropriate for an epidemic that develops on a time-scale much faster than demographic turn-around. The epidemic SIR model was extended by Soper who added [34] (constant) birth and death rates to the model, obtaining the so-called SIR endemic model, that has been extensively analysed in the following decades, especially to investigate how to explain the apparent periodicities in the notifications of childhood diseases [33, 20]. The SIR endemic model can be seen as the basis, over which more complex and realistic models have been built.

31 The difference in time-scales between epidemic spread and demographic turnaround has been
32 observed by several authors. Smith [33] introduced a small parameter ϵ as the ratio between the
33 average lengths of the infection period and of life; he proved that, if the contact rate is a sinusoidal
34 function of period 1 and ϵ is sufficiently small, a subharmonic bifurcation of a 2-periodic stable
35 positive solution can occur. Andreasen [2] showed that, for ϵ small enough, the endemic equilibrium
36 is always stable in a certain class of age-dependent SIR models. Diekmann, Heesterbeek and Britton
37 [9] have exploited the fact that ϵ is a small parameter in an informal argument about the minimum
38 community size in which a measles-like infection can persist. However, to our knowledge very few
39 authors have systematically used geometric singular perturbation theory as a tool to investigate
40 properties of epidemic models. We only know of the paper by Rocha *et al.* [31] that used singular
41 perturbation methods for the analysis of a SIRUV model for a vector-borne epidemic.

42 Our main objective in this paper is to show that under certain assumptions of the system pa-
43 rameters (namely the transition rates between states), tools from Geometric Singular Perturbation
44 Theory (GSPT) are suitable to describe the intricate dynamics that such models exhibit due to the
45 presence of multiple time scales.

46 The first part of the paper is devoted to the classical SIR and SIRS epidemic models, that we
47 analyse in the limiting case of $\epsilon \rightarrow 0$. For such models, it is well known that, when $R_0 > 1$, there
48 exists a unique endemic equilibrium, which is globally asymptotically stable.

49
50 In the second part, we instead consider a model, named SIRWS, introduced for pertussis in [27],
51 and partially analysed in [4]. In the model it is assumed that immunity wanes in two stages: after
52 recovering from infection individuals are totally immune, but then immune memory starts to fade:
53 if they are challenged by the pathogen when they are in the stage of partial immunity, they recover
54 a complete immunity; otherwise, they completely lose immunity, and re-enter the susceptible stage.

55 Our main results can be summarized as follows:

- 56 • For the fast-slow SIR and SIRS models we capture the transient behaviour from an initial intro-
57 duction of the infection, and show that, when $R_0 > 1$, the dynamics leads, in the slow time-scale,
58 to a neighbourhood of the endemic equilibrium, see Sections 3.1 and 3.3. Then convergence to
59 the equilibrium can be established by local methods.
- 60 • For the fast-slow SIRWS model, in particular, we confirm the result obtained numerically in [4]
61 that stable periodic epidemic outburst can exist. Moreover, we give a detailed description of the
62 system parameters for which such behaviour occurs and the corresponding time scales involved,
63 see Section 3.4.

64 Our mathematical analysis is largely based on GSPT, see more details in Section 2.

65 In such a context, it is worth mentioning that the models we study are not immediately, nor
66 globally, in a standard singularly perturbed form, but in each model the fast-slow decomposition
67 appears only in specific regions of the phase space, similarly to what is considered in e.g. [22, 25].
68 As it is usually the case in such biological models, the main difficulty for analysis is due to the
69 loss of normal hyperbolicity of the critical manifold. To overcome this obstacle, we use here the
70 so called entry-exit function, as presented by De Maesschalck and Schecter [6], which gives details
71 regarding the behaviour of an orbit in regions where the critical manifold changes its stability
72 properties. Moreover, for the modified SIRWS system we present a combination of analytical and
73 numerical studies regarding the dependence of the dynamics with respect to some of the parameters,
74 and compare our results with the ones obtained in [4]. In particular, we focus on the interplay

75 between life expectancy (or birth/death rate) and boosting rate, and on how different values of
 76 these parameters can give rise to damped or sustained oscillations. Finally, the novelty of our
 77 analysis is not confined to the usage of GSPT in the context of the well-known SIR model, but we
 78 also show that our techniques can be potentially used in higher dimensional systems (as the SIRWS
 79 model). This is rather important since the well-studied SIR and SIRS models often depend on
 80 Lyapunov’s method to show stability of trajectories [29], and it is known that Lyapunov functions
 81 are difficult to obtain. Our GSPT analysis does not require global Lyapunov functions.

82 The remainder of this paper is arranged as follows: in Section 2 we provide some necessary
 83 mathematical preliminaries which will be later used for the analysis of the models. Afterwards,
 84 we present in Section 3 the mathematical analysis of the SIR, SIRS, and the SIRWS epidemiolog-
 85 ical models. We finish in Section 4 with a summary and an outlook of open-problems regarding
 86 modelling and analysis of epidemiological models with fast-slow dynamics.

87 2 Preliminaries

88 In the main part of this paper we study three compartment models whose dynamics evolve at distinct
 89 time scales. Therefore, we now provide a brief description of Geometric Singular Perturbation
 90 Theory (GSPT), and in particular of the entry-exit function [6], which is fundamental in our
 91 analysis.

92 2.1 Fast-slow systems

93 The term “fast-slow systems” is commonly used to model phenomena which evolve on two (or
 94 more) different time scales [3, 24]. Often such behaviour can be described by a singularly perturbed
 95 ordinary differential equation (ODE), that is

$$\begin{aligned} \epsilon \dot{x} &= f(x, y, \epsilon), \\ \dot{y} &= g(x, y, \epsilon), \end{aligned} \tag{1}$$

96 where $x = x(\tau) \in \mathbb{R}^m$, $y = y(\tau) \in \mathbb{R}^n$, with $m, n \geq 1$, are the fast and slow variables respectively,
 97 f and g are functions of class C^k , with k as large as needed, and $0 < \epsilon \ll 1$ is a small parameter
 98 which gives the ratio of the two time scales. Here the overdot ($\dot{\cdot}$) indicates $\frac{d}{d\tau}$. The system (1) is
 99 formulated on the *slow time* scale τ . When studying fast-slow systems we often define a new *fast*
 100 *time* $t = \tau/\epsilon$ with which (1) can be rewritten as

$$\begin{aligned} x' &= f(x, y, \epsilon), \\ y' &= \epsilon g(x, y, \epsilon), \end{aligned} \tag{2}$$

101 where now the prime ($'$) indicates $\frac{d}{dt}$. Clearly, since we simply rescaled the time variable, sys-
 102 tems (1) and (2) are equivalent for $\epsilon > 0$.

103 Fast-slow systems given by (1)-(2) are said to be *in standard form*. In a more general context,
 104 it is possible to have a fast-slow system given by

$$z' = F(z, \epsilon), \tag{3}$$

105 where the time scale separation is not explicit. In fact, many biological models [22, 25], among
 106 others, and in particular the models we study in this paper are in such non-standard form.

107 The main idea of GSPT is to consider (1)-(2) in the limit $\epsilon \rightarrow 0$ and then use perturbation
 108 arguments to describe the dynamics of the full fast-slow system. The motivation behind this
 109 strategy is that one expects that the analysis of the limit systems ($\epsilon = 0$) is simpler compared to
 110 the analysis of (1)-(2) with $\epsilon > 0$.

111 Taking the limit $\epsilon \rightarrow 0$ in systems (1) and (2) yields, respectively

$$\begin{aligned} 0 &= f(x, y, 0), \\ \dot{y} &= g(x, y, 0), \end{aligned} \tag{4}$$

112 and

$$\begin{aligned} x' &= f(x, y, 0), \\ y' &= 0, \end{aligned} \tag{5}$$

where (4) is called *reduced subsystem* (or *slow subsystem*), and (5) is called *the layer equation* (or *fast subsystem*). We note that the reduced subsystem describes a dynamic evolution constrained to the set

$$\mathcal{C}_0 = \{x \in \mathbb{R}^m, y \in \mathbb{R}^n \mid f(x, y, 0) = 0\},$$

113 which is called the *critical manifold*. On the other hand, we note that \mathcal{C}_0 defines the set of equilib-
 114 rium points of the layer equation.

115 Fenichel's theorems, which are the basis of GSPT, require certain assumptions on \mathcal{C}_0 . Namely, we
 116 suppose there exists an n -dimensional compact submanifold \mathcal{M}_0 , possibly with boundary, contained
 117 in \mathcal{C}_0 . Moreover, the manifold \mathcal{M}_0 is assumed to be *normally hyperbolic* and *locally invariant*, which
 118 mean, respectively, that the eigenvalues of the Jacobian $D_x f(x, y, 0)|_{\mathcal{M}_0}$ are uniformly bounded
 119 away from the imaginary axis, and that the flow can only leave \mathcal{M}_0 through its boundary. In such
 120 a setting, the following can be proved (see [10]):

121 **Theorem 2.1.** *For $\epsilon > 0$ sufficiently small, there exists a manifold \mathcal{M}_ϵ , called slow manifold,*
 122 *which lies $\mathcal{O}(\epsilon)$ close to \mathcal{M}_0 , is diffeomorphic to \mathcal{M}_0 and is locally invariant under the flow of (2).*

123 We note that the manifold \mathcal{M}_ϵ is usually not unique, but all the possible choices lie $\mathcal{O}(\epsilon^{-K/\epsilon})$ -
 124 close to each other, for some $K > 0$. Therefore, in most cases the choice of slow manifold \mathcal{M}_ϵ does
 125 not change the analytical and numerical results.

With the usual definitions for stable and unstable manifolds (see, for example, equations (6.3) in [24])

$$\begin{aligned} W^s(\mathcal{M}_0) &= \{(x, y) : \phi_t(x, y) \rightarrow \mathcal{M}_0 \text{ as } t \rightarrow +\infty\}, \\ W^u(\mathcal{M}_0) &= \{(x, y) : \phi_t(x, y) \rightarrow \mathcal{M}_0 \text{ as } t \rightarrow -\infty\}, \end{aligned}$$

126 where ϕ_t denotes the flow of system (5), Fenichel's second theorem ensures that $W^s(\mathcal{M}_0)$ and
 127 $W^u(\mathcal{M}_0)$ persist under perturbation as well:

128 **Theorem 2.2.** *For $\epsilon > 0$ sufficiently small, there exist manifolds $W^s(\mathcal{M}_\epsilon)$ and $W^u(\mathcal{M}_\epsilon)$ which lie*
 129 *$\mathcal{O}(\epsilon)$ close to and are diffeomorphic to $W^s(\mathcal{M}_0)$ and $W^u(\mathcal{M}_0)$ respectively, and are locally invariant*
 130 *under the flow of (2).*

131 In practical terms, Fenichel's theorems show that for $\epsilon > 0$ sufficiently small, the dynamics of
 132 (1)-(2) are a regular perturbation of the limit dynamics (4)-(5) within a small neighbourhood of
 133 the critical manifold.

134 When the manifold \mathcal{M}_0 is not normally hyperbolic, some more advanced tools, such as the
 135 *blow-up method* (see [18]), may need to be invoked. All of the systems we analyse below have one
 136 non-hyperbolic point in the biologically relevant region. Thus, in order to describe the relevant
 137 dynamics we need to use extra techniques besides Fenichel’s theorems. Due to the properties of the
 138 models to be studied, it turns out that the *entry-exit function* [5, 6] is suitable.

139 2.2 Entry-exit function

140 The entry-exit function gives, in the form of a Poincaré map between two sections in phase space,
 141 an estimate of the behaviour of the orbits near the point in which the critical manifold changes
 142 stability (from attracting to repelling), in a class of singularly perturbed systems. Intuitively, the
 143 result can be interpreted as a “build up” of repulsion near the repelling part of the slow manifold,
 144 which needs to compensate the attraction which was built up near the attracting part before the
 145 orbit can leave an $\mathcal{O}(\epsilon)$ neighbourhood of the critical manifold.

146 More specifically, this construction applies to systems of the form

$$\begin{aligned} x' &= f(x, y, \epsilon)x, \\ y' &= \epsilon g(x, y, \epsilon), \end{aligned} \tag{6}$$

147 with $(x, y) \in \mathbb{R}^2$, $g(0, y, 0) > 0$ and $\text{sign}(f(0, y, 0)) = \text{sign}(y)$. Note that for $\epsilon = 0$, the y -axis consists
 of normally attracting/repelling equilibria if y is negative/positive, respectively.

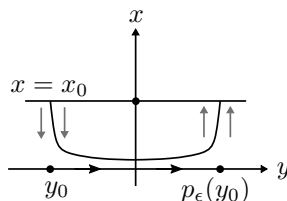


Figure 1: Visualization of the entry-exit map on the line $x = x_0$

148
 149

150 Consider a horizontal line $\{x = x_0\}$, which is $\mathcal{O}(\epsilon)$ -close to the y -axis. An orbit of (6) that
 151 intersects such a line at $y = y_0 < 0$ (entry) re-intersects it again (exit) at $y = p_\epsilon(y_0)$, as sketched
 152 in Figure 1. De Maesschalck [5] shows that, as $\epsilon \rightarrow 0$, the image of the return map $p_\epsilon(y_0)$ to the
 153 horizontal line $x = x_0$ approaches $p_0(y_0)$ given implicitly by

$$\int_{y_0}^{p_0(y_0)} \frac{f(0, y, 0)}{g(0, y, 0)} dy = 0. \tag{7}$$

154 In the following sections, the entry-exit function p_0 plays a crucial role in the analysis of three
 155 different epidemiological models. In particular, the analysis of the SIRWS model relies on a multi-
 156 dimensional version of the entry-exit map, provided in a recent paper by Hsu and Ruan [16].

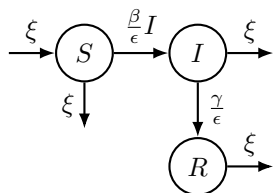
157 3 Analysis of the SIR, SIRS and SIRWS models

158 In this section we analyse three different epidemiological models, giving a short interpretation of the
 159 equations and then proceeding to use the techniques of GSPT, especially the entry-exit function,

160 to deduce information about the behaviour of each one.

161 3.1 SIR model

162 We consider a SIR compartment model (presented in a similar form in [14] and with the same
163 underlying dynamics in [21]) as depicted in Figure 2 and with corresponding equations given as in
164 (8)



$$\begin{aligned}
 \dot{S} &= \xi - \xi S - \frac{\beta}{\epsilon} SI, \\
 \dot{I} &= \frac{\beta}{\epsilon} SI - \frac{\gamma}{\epsilon} I - \xi I, \\
 \dot{R} &= -\xi R + \frac{\gamma}{\epsilon} I,
 \end{aligned} \tag{8}$$

166 Figure 2: Flow diagram for (8).

167 where $S(\tau)$, $I(\tau)$, $R(\tau)$ denote the susceptible, infected and recovered proportion of the population
168 respectively. Since the (S, I, R) variables represent fractions of a population, they are assumed to
169 be non-negative for all $\tau \geq 0$. Observe that the non-negative octant of \mathbb{R}^3 , to be denoted by $\mathbb{R}_{\geq 0}^3$,
170 and in particular the set $\{(S, I, R) \in \mathbb{R}_{\geq 0}^3 \mid 0 \leq S + I + R \leq 1\}$, are invariant under the flow of (8).

171 The parameter ξ in (8) refers to the birth rate and is assumed to be equal to the death rate.
172 Furthermore, as depicted in Figure 2, we also assume that all individuals are born susceptible.
173 Similarly, the parameter β and γ refer, respectively, to the rates at which susceptible individuals
174 are infected and the latter are recovered. In our analysis the parameters ξ , β and γ are of order
175 $\mathcal{O}(1)$. Note that we introduce a small positive parameter $0 < \epsilon \ll 1$, which gives rise to the
176 difference in magnitude between the large infection rate β/ϵ , the large recovery rate γ/ϵ and the
177 birth/death rate. Such a difference represents a highly contagious disease with a short infection
178 period.

179 As stated above, $S(\tau)$, $I(\tau)$ and $R(\tau)$ represent proportions of the population. Consistently the
180 plane $\{S + I + R = 1\}$ is invariant for system (8). Hence, we can assume $R = 1 - S - I$, which
181 allows us to reduce (8) to

$$\begin{aligned}
 \dot{S} &= \xi - \xi S - \frac{\beta}{\epsilon} SI, \\
 \dot{I} &= \frac{\beta}{\epsilon} SI - \frac{\gamma}{\epsilon} I - \xi I.
 \end{aligned} \tag{9}$$

182 By rescaling time, system (9) can also be written as

$$\begin{aligned}
 S' &= \epsilon \xi (1 - S) - \beta SI, \\
 I' &= I(\beta S - \gamma - \epsilon \xi).
 \end{aligned} \tag{10}$$

183 Note that system (10) is a fast-slow system in non-standard form, as it often occurs in biological
184 models [22, 25]. Later we perform a convenient rescaling that brings (10) into a standard form.

185 The corresponding critical manifold is the set $\mathcal{C}_0 = \{(S, I) \in \mathbb{R}^2 \mid I = 0\}$, and the slow flow
186 along it is given by $\dot{S} = \xi(1 - S)$, which implies flow towards the point $S = 1$. In the $\epsilon \rightarrow 0$ limit,
187 we recover from (10) the basic dynamics for the (S, I) couple in a standard SIR system (see [14]),

188 namely

$$\begin{aligned} S' &= -\beta SI, \\ I' &= I(\beta S - \gamma). \end{aligned} \tag{11}$$

189 In particular, it follows from linearization of (11) along \mathcal{C}_0 that the critical manifold is attracting
190 for $S < \frac{\gamma}{\beta}$, repelling for $S > \frac{\gamma}{\beta}$, and loses normal hyperbolicity at $S = \frac{\gamma}{\beta}$.

191 From here on, we assume the basic reproduction number to be $R_0 = \beta/\gamma > 1$. This means that
192 the disease is able to spread through the population. In particular, as stated in the well known next
193 Lemma [15, 21], the previous assumption implies that, for every initial condition $S(0) = S_0 > 1/R_0$,
194 there exists a unique $S_\infty < 1/R_0$ such that a trajectory of (11) with initial conditions (S_0, I_0)
195 converges towards $(S_\infty, 0)$ as $t \rightarrow +\infty$.

196 **Lemma 1.** $\Gamma(S, I) = \gamma \ln(S) - \beta(S + I)$ is a constant of motion for system (11), and all its orbits
197 in the first quadrant are heteroclinic to two points on the S -axis.

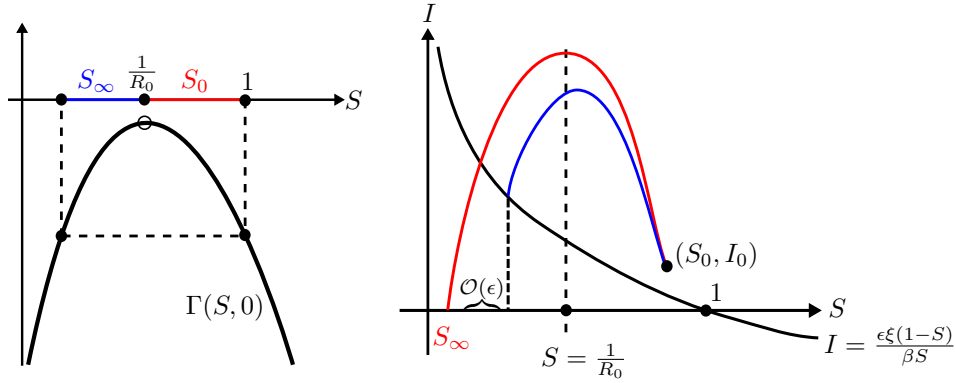


Figure 3: Left: function $\Gamma(S, 0)$, intersection with horizontal lines give the starting and ending points of a heteroclinic orbit of the layer equation (11). Right: qualitative comparison between perturbed and unperturbed SIR systems in fast time scale. In red we show an orbit of (11) given by $\Gamma(S, I) = \Gamma(S_0, I_0)$ and in blue a small perturbation of it corresponding the related orbit of (10).

198 From Lemma 1 we define $S_\infty \in (0, \frac{1}{R_0})$ to be the unique non-trivial solution of the equation
199 $\Gamma(S, 0) = \Gamma(S_0, 0)$ where $S_0 > \frac{1}{R_0}$.

200 For future use, let us define the map

$$\Pi_1 : \{S \in (1/R_0, 1]\} \rightarrow \{S \in (0, 1/R_0)\} \tag{12}$$

201 that maps S_0 into S_∞ , and which is induced by the flow of (11), or is equivalently given by Γ .

202 So far, we know that the solutions of (10) away from the critical manifold are closely given by
203 $\Gamma(S, I)$ as shown in the right side of Figure 3. Therefore, the next step is to focus on a small region
204 close to \mathcal{C}_0 . That is, for the analysis that follows, we assume I to be $\mathcal{O}(\epsilon)$ -small. In particular,
205 and following Lemma 1, if we choose $I_0 \in \mathcal{O}(\epsilon^2)$, we have an explicit relation (up to a $\mathcal{O}(\epsilon)$ error)
206 between S_∞ and S_0 , namely, $\Gamma(S_\infty, 0) \approx \Gamma(S_0, I_0) = \Gamma(S_0, 0) + \mathcal{O}(\epsilon)$.

207 Considering the signs of the derivatives in the perturbed system (10), we see that orbits spiral
208 counterclockwise. Moreover, system (10) has a two equilibria, namely $(S, I) = (1, 0)$ and one

209 which is $\mathcal{O}(\epsilon)$ -close to the point $(1/R_0, 0)$, as shown in Figure 4, given by $(S, I) = (S_E, I_E) :=$
 210 $(\frac{1}{R_0} + \epsilon \frac{\xi}{\beta}, \alpha_\epsilon(S_E))$, where

$$\alpha_\epsilon(S) = \frac{\epsilon \xi (1 - S)}{\beta S} \quad (13)$$

211 is obtained from the nullcline for S in (8). Regular perturbation arguments imply that an orbit
 212 of the perturbed system (10), starting from a point (S_0, I_0) with $I_0 \in \mathcal{O}(\epsilon)$ and $S_0 > S_E$, follows
 213 $\mathcal{O}(\epsilon)$ -closely from below, since the $\mathcal{O}(\epsilon)$ contribution is negative, a power level of $\Gamma(S, I)$, until it
 214 reaches the nullcline of S given by $I = \frac{\epsilon \xi (1 - S)}{\beta S}$, as shown on the right half of Figure 3, at a point
 with S coordinate $\mathcal{O}(\epsilon)$ -close to S_∞ .

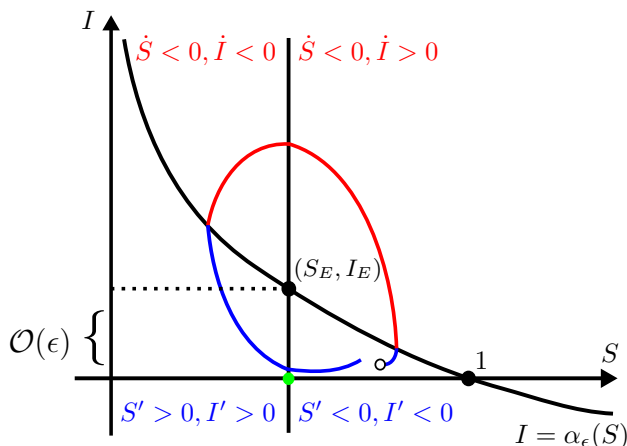


Figure 4: Schematic representation of the orbits of (10) on the two time scales. Red: fast orbit; blue: slow orbit; green: non-hyperbolic point.

215 It is also well known [15, 29] that the endemic equilibrium (S_E, I_E) is globally asymptotically
 216 stable, as stated below.
 217

218 **Theorem 3.1.** *Consider (10). All trajectories with initial conditions $0 \leq S(0) \leq 1$, $0 < I(0) \leq 1$
 219 with $S(0) + I(0) \leq 1$ converge asymptotically towards the (endemic) equilibrium point (S_E, I_E) .*

220 The theorem can be proved using the Lyapunov function

$$L_1(S, I) = S + I - S_E \ln(S) - I_E \ln(I) - C_E, \quad (14)$$

221 with $C_E = S_E + I_E - S_E \ln(S_E) - I_E \ln(I_E)$, together with Lasalle's invariance principle [26]; or
 222 with [29, 32]

$$L_2(S, I) = I - I_E - I_E \ln(I/I_E) + \frac{\beta}{2(2\mu + \gamma)} (S + I - S_E + I_E)^2. \quad (15)$$

223 Here we are going to describe how solutions approach the equilibrium, for $\epsilon > 0$ small. Once
 224 it is shown that solutions are in a neighbourhood of the equilibrium, local methods can be used to
 225 prove convergence to the equilibrium. Such an approach will be used for the other models as well.

226 Our motivation is to present a method of analysis that does not depend on finding a Lyapunov
 227 function, which is, in general, a difficult task.

228 A convenient step, which is justified by the following Lemma, is to bring (10) to a standard
 229 form, in order to then apply the entry-exit formula.

230 **Lemma 2.** Consider (10) and an initial condition (S_0, I_0) with $0 < S_0 \leq \frac{\gamma}{\beta} - \Delta < S_E$ and $I_0 > 0$,
 231 where $\Delta \in \mathcal{O}(1)$ and $I_0 \in \mathcal{O}(\epsilon)$. Let $0 < \Delta_1 < \Delta$, $\Delta_1 \in \mathcal{O}(1)$, and (S^*, I^*) denote the point
 232 where the corresponding trajectory intersects the line $\ell = \{(S, I) \in \mathbb{R}^2 \mid S = \frac{\gamma}{\beta} - \Delta_1\}$. Then, for
 233 sufficiently small $\epsilon > 0$ we have that I^* is exponentially small. Furthermore, the first point at which
 234 the trajectory intersects the line $\pi_\epsilon = \{(S, I) \in \mathbb{R}^2 \mid I = I_0 \epsilon^k\}$ satisfies $S = S_0 + \mathcal{O}(\epsilon \log(\epsilon))$ for
 235 $\epsilon \rightarrow 0$.

236 *Proof.* We first note that the assumption on S_0 simply means that S_0 is bounded away from S_E
 237 uniformly in ϵ . For the proof it is convenient to define new coordinates (S, v) by $(S, I/\epsilon) = (S, v)$.
 238 Then (10) becomes

$$\begin{aligned} S' &= \epsilon(\xi(1 - S) - \beta S v), \\ v' &= v(\beta S - \gamma - \epsilon \xi). \end{aligned} \tag{16}$$

A trajectory of (16) with initial condition (S_0, v_0) with $S_0 < \frac{\gamma}{\beta}$ and $v_0 = I_0/\epsilon \in \mathcal{O}(1)$ quickly
 converges towards and stays $\mathcal{O}(\epsilon)$ -close to the S -axis for some time. We know from the reduced
 system that $S' > 0$ on the critical manifold, this guarantees that the trajectory crosses the line ℓ in
 a small neighbourhood of the critical manifold. Let T denote the (slow) time it takes the trajectory
 to reach ℓ . During such time, $\beta S - \gamma \leq -\beta \Delta_1 < 0$ and therefore

$$v' \leq -Kv \implies v\left(\frac{T}{\epsilon}\right) \leq v_0 \epsilon^{-K \frac{T}{\epsilon}} \implies I\left(\frac{T}{\epsilon}\right) \leq \epsilon v_0 \epsilon^{-K \frac{T}{\epsilon}},$$

239 with $K = \beta \Delta_1 > 0$.

240 The last claim follows immediately from $v(t) \leq v_0 \epsilon^{-Kt}$. □

241 Note in particular from Lemma 2 that, before the trajectory intersects ℓ , its corresponding
 242 I -coordinate is eventually $\mathcal{O}(\epsilon^2)$, which is what we need for the forthcoming arguments.

243 3.2 Applying the entry-exit function

244 We are now going to apply the entry-exit formula to describe the way trajectories pass near the
 245 non-hyperbolic point $(S, I) = (1/R_0, 0)$.

246 From Lemma 1 and 2, we can consider an initial point for system (10) with $S_0 < 1/R_0$ and
 247 $I_0 = \mathcal{O}(\epsilon^2)$. Next, we apply a change of variables defined by

$$S = \frac{u + 1}{R_0}, \quad I = \epsilon v, \tag{17}$$

248 which brings the system to a standard form, with u slow and v fast, that is

$$\begin{aligned} v' &= \gamma(u - \epsilon \xi)v, \\ u' &= \epsilon(\xi(R_0 - u - 1) - \beta v(u + 1)). \end{aligned} \tag{18}$$

249 So, using the notation of Section 2.2,

$$\begin{aligned} f(v, u, \epsilon) &= \gamma(u - \epsilon\xi), \\ g(v, u, \epsilon) &= \xi(R_0 - u - 1) - \beta v(u + 1), \end{aligned} \tag{19}$$

250 which satisfy the hypotheses of the entry-exit function. Indeed, $S < 1$ implies $u < R_0 - 1$, which
251 means $g(0, u, 0) > 0$ in the relevant region. Moreover, $f(0, u, 0) = \gamma u$, which clearly has the same
252 sign as u .

253 Since $v_0 = I_0/\epsilon = \mathcal{O}(\epsilon)$, we can now apply the entry-exit formula, which gives $p_0(u_0)$ as the
254 only positive solution of

$$\int_{u_0}^{p_0(u_0)} \frac{u}{R_0 - 1 - u} du = 0. \tag{20}$$

255 The integral (20) can be solved explicitly, giving $p_0(u_0)$ as the positive solution of

$$-p_0(u_0) + u_0 - (R_0 - 1) \ln \left(\frac{R_0 - 1 - p_0(u_0)}{R_0 - 1 - u_0} \right) = 0. \tag{21}$$

256 We now change back to the original (S, I) variables, and introduce, beyond Π_1 defined in (12),
257 the map

$$\Pi_2 : \{S \in (0, 1/R_0)\} \rightarrow \{S \in (1/R_0, 1)\} \tag{22}$$

258 defined by $\frac{p_0(u_0) + 1}{R_0}$, where $u_0 = R_0 S_0 - 1$. Combining together the previous results, we can state
259 the following:

Proposition 1. *Consider the solution of (9) with an initial condition $S_0 > 1/R_0$ and $I_0 = \mathcal{O}(\epsilon^2)$. Then the orbit $\{S_\epsilon(t), I_\epsilon(t), t \in [0, T]\}$ converges for $\epsilon \rightarrow 0$ to the union of the orbit under the fast flow*

$$\{(S, I) : \Gamma(S, I) = \Gamma(S_0, 0), \Pi_1(S_0) \leq S \leq S_0\}$$

and under the slow flow

$$\{(S, 0) : \Pi_1(S_0) \leq S \leq \Pi_2(\Pi_1(S_0))\}$$

260 where T is such that the solution of $S' = \xi(1 - S)$, $S(0) = \Pi_1(S_0)$ satisfies $S(T) = \Pi_2(\Pi_1(S_0))$.

The limit orbit is sketched in Figure 5. Considering the composition of Π_1 and Π_2 gives the Poincaré map

$$\Pi : \{S \in [S_E, 1], I = I_0\} \rightarrow \{\Pi_2(\Pi_1(S)) \in [S_E, 1], I = I_0\}.$$

In this notation, we define $P_0 = \Pi_1(S_0)$, $S_1 = \Pi_2(P_0) = \Pi(S_0)$. These correspond, in the u -coordinate, to

$$u_0 = R_0 P_0 - 1 \approx R_0 S_\infty - 1, \quad p_0(u_0) = R_0 S_1 - 1.$$

We rewrite (21) as

$$P_0 - S_1 - \left(1 - \frac{1}{R_0}\right) \ln \left(\frac{1 - S_1}{1 - P_0}\right) = 0.$$

261 Which means that S_1 , the exit point, is the only root greater than P_0 of

$$F(x) = x - P_0 + \left(1 - \frac{1}{R_0}\right) \ln \left(\frac{1 - x}{1 - P_0}\right). \tag{23}$$

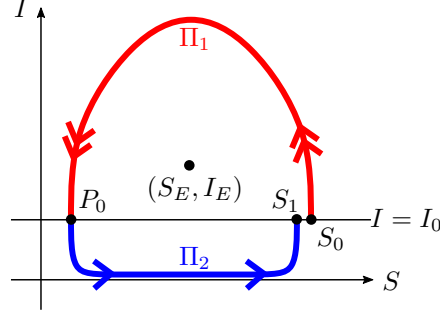


Figure 5: Sketch of the fast and slow dynamics defining the maps Π_1 and Π_2 . The fact that $S_1 < S_0$ is shown below.

262 It is clear that when the trajectory is in a neighbourhood of (S_1, I_0) , as implied by the entry-
 263 exit map, one can reapply Proposition 1, obtaining $P_1 = \Pi_1(S_1)$ (reached through the fast flow),
 264 $S_2 = \Pi_2(P_1)$ (slow flow), and so on, obtaining two sequences

$$S_0, S_1 = \Pi_2(P_0), \dots, S_n = \Pi_2(P_{n-1}), \dots \quad P_0 = \Pi_1(S_0), \dots, P_n = \Pi_1(S_n), \dots \quad (24)$$

265 **Lemma 3.** *The sequence $\{S_n\}$ is decreasing and bounded below by $1/R_0$; the sequence $\{P_n\}$ is*
 266 *increasing and bounded above by $1/R_0$.*

267 *Proof.* We recall $S_1 = \Pi_2(P_0) = \Pi(S_0)$, so if, for any $S_0 \in (1/R_0, 1)$, such value is smaller/greater
 268 than S_0 , $\{S_n\}$ is decreasing/increasing.

269 We notice that $\Pi(S_0) < S_0$ if and only if $\Pi_2(P_0) < \Pi_1^{-1}(P_0)$, where $\Pi_1^{-1}(P_0) > P_0$ is the only such
 270 root of

$$G(x) = x - P_0 + \frac{1}{R_0} \ln \left(\frac{P_0}{x} \right), \quad (25)$$

271 which comes from $\Gamma(x, 0) = \Gamma(P_0, 0)$; we recall that Γ describes the trajectories of the layer equation.
 272 The functions F and G are sketched in Figure 6.

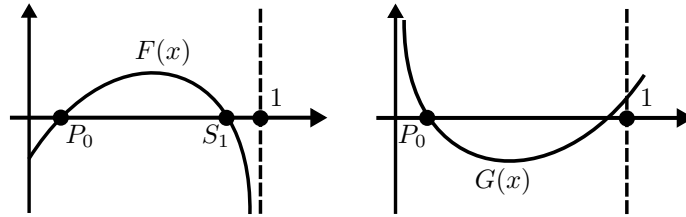


Figure 6: Sketch of the functions F and G , which implicitly define Π_2 and Π_1^{-1} , respectively.

Then, since G is increasing for $x > 1/R_0$,

$$\Pi_2(P_0) < \Pi_1^{-1}(P_0) \iff G(\Pi_2(P_0)) < 0.$$

273 The fact that $S_1 < S_0$ can be shown as a particular case of the following, more general propo-
 274 sition, by taking $a = P_0$, $b = 1/R_0$, $x^* = S_1$.

275 **Lemma 4.** Let $0 < a < b < 1$, $F(x) = x - a + (1 - b) \ln(\frac{1-x}{1-a})$, $G(x) = x - a + b \ln(\frac{a}{x})$. Let
 276 $x^* \in (a, 1)$ be the only zero greater than a of F . Then $G(x^*) < 0$.

Proof. We use the auxiliary function $H(x) = F(x) + \frac{b}{1-b}G(x)$, which, under the hypotheses, is decreasing for $x \in (0, 1)$. Next we have that $H(a) = F(a) + \frac{b}{1-b}G(a) = 0$ which implies

$$0 > H(x^*) = F(x^*) + \frac{b}{1-b}G(x^*) = \frac{b}{1-b}G(x^*) \implies G(x^*) < 0.$$

277

□

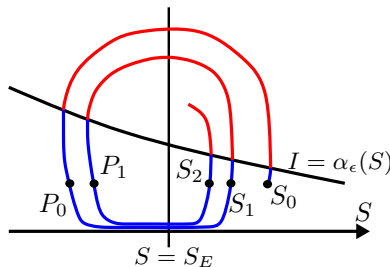


Figure 7: $\alpha_\epsilon(S) = \mathcal{O}(\epsilon)$; the red parts of the orbit are fast for both variables, the blue parts are fast for I , slow for S .

278 Since Π_1 is a decreasing function, from the fact that $\{S_n\}$ is decreasing, it follows that $\{P_n\}$ is
 279 increasing. □

280 **Proposition 2.** The sequences $\{S_n\}$ and $\{P_n\}$ defined in (24) both converge to $1/R_0$.

281 *Proof.* The convergence can be shown reasoning by contradiction, for example by looking at the
 282 sequence S_i . We know it is decreasing, and bounded below by $1/R_0$, so if it is not converging
 283 to this value, it must be converging to some other value $S_{\text{lim}} > 1/R_0$. But if this is the case,
 284 $\Pi(S_{\text{lim}}) < S_{\text{lim}}$, which contradicts the nature of S_{lim} .

285 Completely analogously we can see that $P_i \rightarrow 1/R_0$. □

286 Extending Proposition 1, one can easily show that, if $S_0 > 1/R_0$ and $I_0 = \mathcal{O}(\epsilon^2)$, the orbits
 287 $\{S_\epsilon(t), I_\epsilon(t), t \in [0, T]\}$ for any T converge for $\epsilon \rightarrow 0$ to a finite union of orbits (under the fast flow)
 288 from $(S_n, 0)$ to $(P_n, 0)$, and slow flows on the S -axis from $(P_n, 0)$ to $(S_{n+1}, 0)$.

289 The same can be shown for any initial condition, since starting from any (S_0, I_0) with $I_0 > 0$,
 290 the solutions will approach a point $(S_\infty, 0)$ with $S_\infty < 1/R_0$, so that setting $P_0 = S_\infty$, one can
 291 repeat the above argument.

292 What can we say of the orbits $\{S_\epsilon(t), I_\epsilon(t)\}$ for ϵ small but fixed as $t \rightarrow \infty$? When $1/R_0 - P_n =$
 293 $\mathcal{O}(\epsilon)$, the argument of Lemma 2 does not work. Hence, we cannot say, and indeed it is no longer
 294 true, that $I(t)$ becomes $\mathcal{O}(\epsilon^2)$ afterwards, and we cannot apply the entry-exit Lemma as above.

295 However, the previous argument shows that $\{S_\epsilon(t), I_\epsilon(t)\}$ reaches an ϵ -neighbourhood of the
 296 equilibrium (S_E, I_E) . Linearization at the equilibrium then shows that all trajectories of (10)
 297 starting in the set $\{(S, I) \in \mathbb{R}^2 \mid S \geq 0, I > 0, S + I \leq 1\}$ converge towards (S_E, I_E) , as already
 298 known (Theorem 3.1). This analysis provides an alternative proof, valid for $\epsilon > 0$ sufficiently small.

299 Biologically, the above analysis tells us that between two consecutive peaks of infection there is
 300 a long ($\mathcal{O}(1/\epsilon)$) time during which the fraction of infected population is exponentially small. On the
 301 other hand, the duration of high infected portion of the population is rather small (it occurs on the
 302 fast time scale). Ultimately, however, under the setting of this section the only possible asymptotic
 303 outcome is convergence towards the endemic equilibrium (S_E, I_E) via damped oscillations.

304 3.3 SIRS model

305 We now consider a SIRS compartment model. The SIRS model is a slight modification of the SIR
 306 model and thus we keep the same notation. The SIRS model is given by the following system:

$$\begin{aligned}
 \dot{S} &= -\frac{\beta}{\epsilon}SI + \delta R, \\
 \dot{I} &= \frac{\beta}{\epsilon}SI - \frac{\gamma}{\epsilon}I, \\
 \dot{R} &= \frac{\gamma}{\epsilon}I - \delta R.
 \end{aligned}
 \tag{26}$$

308 Figure 8: Flow diagram for (26).

309
 310 In this model there is no birth nor death, so the population remains constant. The small
 311 positive parameter $0 < \epsilon \ll 1$ gives rise to the difference in magnitude between the large infection
 312 rate β/ϵ , the large recovery rate γ/ϵ and the rate of loss of immunity δ . This difference models
 313 a highly contagious disease with a short infection period with possibility of reinfection. The main
 314 distinctions with the SIR system presented in Section 3.1 are the absence of demographic dynamics
 315 (no birth/death) and the possible loss of immunity (meaning that individuals can move from R
 316 to S). As we will see shortly, however, this important biological difference does not modify the
 317 qualitative behaviour of the system.

318 As we noticed in Section 3.1, $\dot{N} = \dot{S} + \dot{I} + \dot{R} = 0$, that is, the total population remains constant,
 319 so we assume without loss of generality $N(0) = 1$, which implies $N(\tau) \equiv 1$ for all $\tau \geq 0$; this allows
 320 us, using $R = 1 - S - I$, to reduce the system to

$$\begin{aligned}
 \dot{S} &= -\frac{\beta}{\epsilon}SI + \delta(1 - S - I), \\
 \dot{I} &= \frac{\beta}{\epsilon}SI - \frac{\gamma}{\epsilon}I.
 \end{aligned}
 \tag{27}$$

321 Proceeding as in the first model, we introduce the fast time variable $t = \tau/\epsilon$, which gives

$$\begin{aligned}
 S' &= -\beta SI + \epsilon\delta(1 - S - I), \\
 I' &= I(\beta S - \gamma),
 \end{aligned}
 \tag{28}$$

322 where now the prime (') indicates the derivative with respect to t .

323 The critical manifold is, as before, the set $\mathcal{C}_0 = \{(S, I) \in \mathbb{R}^2 \mid I = 0\}$, and the slow flow along it
 324 is given by $\dot{S} = \delta(1 - S)$, which implies flow towards the point $(S, I) = (1, 0)$.

325 The $\epsilon \rightarrow 0$ limit system corresponding to (28) is

$$\begin{aligned}
 S' &= -\beta SI, \\
 I' &= I(\beta S - \gamma),
 \end{aligned}
 \tag{29}$$

326 which is exactly the limit system we obtained in Section 3.1. Hence, we can apply the same
 327 qualitative reasoning as before, with some small changes: in the perturbed system the nullcline for
 328 S is slightly different, giving $I = \alpha(S) = (\epsilon\delta(1 - S))/(\beta S + \epsilon\delta)$, and the value of S_E is exactly $1/R_0$.

329 The previous ansatz for the Lyapunov function does not work here; we could find another one,
 330 following what was done in [29], but we instead follow the analysis with the entry-exit function
 331 which, as we show below, does not change.

The trajectory starting from (S_0, I_0) , with $I_0 \in \mathcal{O}(\epsilon^2)$, follows the same qualitative behaviour:
 after it intersects $I = \alpha(S)$ at a point $(S_\infty + \mathcal{O}(\epsilon), \mathcal{O}(\epsilon))$, it eventually intersects the horizontal line
 $I = I_0$. At that moment, we change the variables as before:

$$S = \frac{u + 1}{R_0}, \quad I = \epsilon v,$$

332 and we obtain a system in standard form:

$$\begin{aligned} v' &= \gamma uv, \\ u' &= \epsilon(-\beta v(u + 1) + \xi(R_0 - u - 1 - \epsilon v)). \end{aligned} \tag{30}$$

333 In the notation of the entry-exit function, then,

$$\begin{aligned} f(v, u, \epsilon) &= \gamma u, \\ g(v, u, \epsilon) &= -\beta v(u + 1) + \xi(R_0 - u - 1 - \epsilon v), \end{aligned} \tag{31}$$

334 which satisfy the hypotheses in the relevant region; hence, we can compute $p_0(u_0)$ with exactly the
 335 same integral equation

$$\int_{u_0}^{p_0(u_0)} \frac{u}{R_0 - 1 - u} du = 0, \tag{32}$$

336 and the procedure we followed for the SIR model can be applied to this SIRS one identically to
 337 show the global convergence to the unique equilibrium.

338 By following a similar analysis as the one performed so far one can also show that considering
 339 a SIRS model with demography would not change the qualitative behaviour of the system.

340 The results obtained so far for the SIR and SIRS models are summarized in the following
 341 Proposition.

342 **Proposition 3.** *The SIR, SIRS without and with demographic dynamics, with infection and recovery*
 343 *rates $\mathcal{O}(1/\epsilon)$ big compared to the other parameters, are all qualitatively equivalent. Their main*
 344 *common features are:*

- 345 • *boundedness of solutions in the set $\{(S, I, R) \in \mathbb{R}_{\geq 0}^3 \mid 0 \leq S + I + R \leq 1\}$,*
- 346 • *population either constant, or converging uniformly and exponentially fast to a constant, which*
 347 *allows to reduce the number of compartments from 3 (S, I, R) to 2 (S, I) ,*
- 348 • *existence of an endemic equilibrium point of the form $(S_E, I_E) = (\frac{1}{R_0} + \mathcal{O}(\epsilon), \mathcal{O}(\epsilon))$,*
- 349 • *fast-slow decomposition in the I and S coordinate, respectively, $\mathcal{O}(\epsilon)$ -close to the critical manifold*
 350 $\mathcal{C}_0 = \{(S, I) \in [0, 1]^2 \mid I = 0\}$,
- 351 • *counterclockwise spiralling of the orbits towards (S_E, I_E) , and consequent absence of periodic*
 352 *orbits.*

353 *These common features mean that, in the long run, the population in each of these models converges*
 354 *to an equilibrium $\mathcal{O}(\epsilon)$ close to $(S, I, R) = (1/R_0, 0, 1 - 1/R_0)$, in the first octant of \mathbb{R}^3 ; each of the*
 355 *three variables have damped oscillations around the equilibrium value.*

356 In the next section we study a more complete (but also more complicated) epidemic model,
 357 where the techniques developed so far shall be extended.

358 3.4 SIRWS model

359 We consider the SIRWS compartment model suggested by Dafilis et al. in [4]. As in the previous
 360 models, we assume that some parameters are $\mathcal{O}(\epsilon)$ small compared to others, making the corre-
 361 sponding processes slow, and the remaining ones fast (the changes correspond to every occurrence
 362 of ϵ in system (33)). This allows us to build on the analysis done in sections 3.1 and 3.3, and to
 363 apply the entry-exit function to a more challenging model.

364 The model we are concerned with in this section is given by:

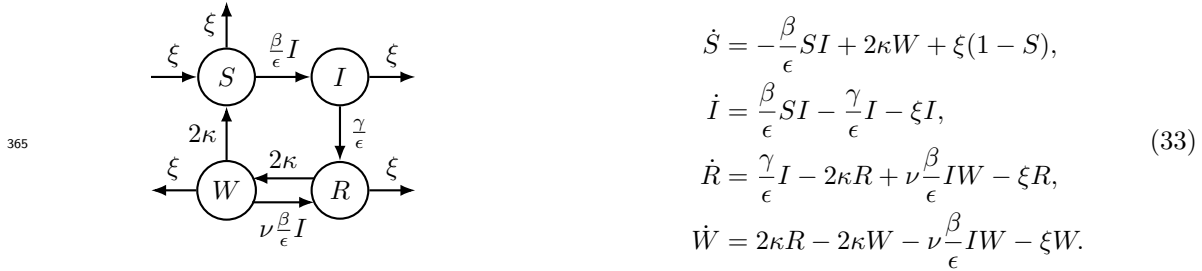


Figure 9: Flow diagram for (33)

366 As in the previous models, susceptible individuals ($S(\tau)$) become infectives ($I(\tau)$) upon contact
 367 with infectious individuals, who, at rate γ/ϵ become immune at their first stage ($R(\tau)$), and then,
 368 at a rate 2κ , become second-stage (‘weakly’) immune ($W(\tau)$). Weakly immune individuals may
 369 then lose totally their immunity at rate 2κ , or, upon contact with infectious individuals, revert
 370 back to fully immune individuals ($R(\tau)$), thanks to the so-called immunity boosting. The constant
 371 ν is the ratio between the rate at which immunity boosting occurs in weakly immune individuals,
 372 and the rate at which susceptibles become infected. Finally, we assume a constant birth rate ξ ,
 373 equal to the death rate, and that all individuals are born susceptible. Through the introduction of
 374 the small parameter ϵ we consider a highly contagious disease with a very short infection period,
 375 compared to other typical times of the system; indeed, the average length of the infectious period
 376 is ϵ/γ , while the average length of life is $1/\xi$ and the total average length of the immune period
 377 is $1/\kappa$ for individuals whose immunity is not boosted. Such relation between the parameters has
 378 been assumed, for example, for diseases such as pertussis, as described in [27], where the authors
 379 estimated $\beta = 260$, $\gamma = 17$, $\xi = 0.01$, $\kappa = 0.1$, $\nu = 20$; hence, the analysis which follows may be
 380 useful in the modelling of such diseases.

381 Analogous to the previous models, the set $\{(S, I, R, W) \in \mathbb{R}_{\geq 0}^4 \mid 0 \leq S + I + R + W \leq 1\}$ is
 382 invariant. We can thus scale the total population to 1, so that we can use $R = 1 - S - I - W$. We
 383 notice that system (8) can be recovered from system (33) by setting $\kappa = \nu = 0$, and ignoring the
 384 consequently decoupled W coordinate.

385 As we shall describe in our analysis below, incorporating the waning state W modifies consider-
 386 ably the dynamics of the model; in fact, it induces the possibility of periodic limit cycles, a feature
 387 that the previous simpler models did not have. This is particularly important when comparing the
 388 dynamics of the SIRWS model with that of the SIRS model where, even if recovered portions of
 389 the population may become again susceptible, there is still no “long run periodic behaviour”.

390 As we have done before, introducing the fast time variable $t = \tau/\epsilon$ brings the system into the
 391 form

$$\begin{aligned} S' &= -\beta SI + \epsilon(2\kappa W + \xi(1 - S)), \\ I' &= \beta SI - \gamma I - \epsilon\xi I, \\ R' &= \gamma I + \nu\beta IW - \epsilon(2\kappa R + \xi R), \\ W' &= -\nu\beta IW + \epsilon(2\kappa R - 2\kappa W - \xi W). \end{aligned} \tag{34}$$

392 **Remark 1.** Note that the critical manifold is (similarly to the previous models) given by

$$C_0 = \{(S, I, R, W) \in [0, 1]^4 \mid I = 0\}. \tag{35}$$

393 Furthermore, in the $\epsilon \rightarrow 0$ limit, S and I become independent of R and W , and orbits follow
 394 the same behaviour we have seen in the fast phases of the first two models. In other words, the
 395 (S, I) -orbits of the layer equation follow a power level of $\Gamma(S, I) = \gamma \ln(S) - \beta(S + I)$, and converge
 396 towards $(S_\infty, 0)$ ¹. These observations motivate the following lemma.

397 **Lemma 5.** Consider the layer equation corresponding to (34). Then, as $(S, I) \rightarrow (S_\infty, 0)$ one has
 398 $W \rightarrow W_\infty := W_0 \exp^{-\nu R_0(S_0 + I_0 - S_\infty)}$, where $W_0 = W(0)$.

Proof. We note that

$$\int_0^\infty \left(S'(u) + I'(u) \right) du = -\gamma \int_0^\infty I(u) du \implies S_0 + I_0 - S_\infty = \gamma \int_0^\infty I(u) du,$$

399 due to the fact that $\lim_{t \rightarrow +\infty} I(t) = 0$. Next, note from (34) that in the limit $\epsilon = 0$ one has
 400 $\frac{W'}{W} = -\nu\beta I$, which implies $W(t) = W_0 \exp^{-\nu\beta \int_0^t I(u) du}$. Letting $t \rightarrow \infty$ leads to the result, recalling
 401 that $R_0 = \frac{\beta}{\gamma}$. \square

402 Since we have already shown that the layer equation is in the (S, I) -coordinates the same as
 403 before, we proceed just in the same way, that is, we apply first the change of coordinates

$$S = \frac{u + 1}{R_0}, \quad I = \epsilon v,$$

404 which gives a system in standard singular perturbation form, with u, W slow and v fast, namely

$$\begin{aligned} v' &= (\gamma u - \epsilon\xi)v =: f(v, u, \epsilon)v, \\ u' &= \epsilon(-\beta v(u + 1) + 2\kappa R_0 W + \xi(R_0 - u - 1)) =: \epsilon g(v, u, W, \epsilon), \\ W' &= \epsilon(-\nu\beta v W + 2\kappa - 2\kappa \frac{u + 1}{R_0} - 4\kappa W - \xi W) + \mathcal{O}(\epsilon^2). \end{aligned} \tag{36}$$

¹We recall that S_∞ is defined as the nontrivial solution of $\Gamma(S, 0) = \Gamma(S_0, 0)$.

405 And, accordingly, in the slow time scale τ :

$$\begin{aligned}
\epsilon \dot{v} &= (\gamma u - \epsilon \xi)v, \\
\dot{u} &= -\beta v(u+1) + 2\kappa R_0 W + \xi(R_0 - u - 1), \\
\dot{W} &= -\nu \beta v W + 2\kappa - 2\kappa \frac{u+1}{R_0} - 4\kappa W - \xi W + \mathcal{O}(\epsilon).
\end{aligned}
\tag{37}$$

406 Naturally, the critical manifold in these new coordinates is $\mathcal{C}_0 = \{(u, v, W) \in \mathbb{R}^3 \mid v = 0\}$.

407 In order to use the entry-exit formula, as described in [16, equation (12)], we first check that
408 indeed

$$\begin{aligned}
g(0, u, W, 0) &= 2\kappa W R_0 + \xi(R_0 - u - 1) > 0, \\
f(0, u, 0) &= \gamma u \leq 0 \iff u \leq 0.
\end{aligned}
\tag{38}$$

409 However, the presence of W in the equation for \dot{u} makes the entry-exit integral

$$\int_{u_0}^{p_0(u_0)} \frac{u}{2\kappa W(u)R_0 + \xi(R_0 - u - 1)} du = 0
\tag{39}$$

410 not immediately computable, as we would need to find an expression for $W(u)$. To deal with this
411 issue, let us look at the (S, W) -dynamics in the slow time variable t on the critical manifold $I = 0$:

$$\begin{aligned}
\dot{S} &= 2\kappa W + \xi(1 - S), \\
\dot{W} &= 2\kappa(1 - S) - (4\kappa + \xi)W.
\end{aligned}
\tag{40}$$

412 This system of ODEs can be solved explicitly, assuming initial conditions
413 $(S(0), W(0)) = (S_\infty, W_\infty)$, the limit values of the fast loop, we have:

$$\begin{aligned}
S(\tau) &= 1 + [S_\infty - 1 + 2\kappa(S_\infty + W_\infty - 1)\tau] \exp(-(2\kappa + \xi)\tau), \\
W(\tau) &= [W_\infty - 2\kappa(S_\infty + W_\infty - 1)\tau] \exp(-(2\kappa + \xi)\tau) \\
&= 1 - S(\tau) - (1 - S_\infty - W_\infty) \exp(-(2\kappa + \xi)\tau).
\end{aligned}
\tag{41}$$

414 The phase-portrait of (40) is illustrated in Figure 10, where the only feasible region is the triangle
415 $0 \leq S + W \leq 1$, $S, W \geq 0$, and all trajectories converge to $(S, W) = (1, 0)$.

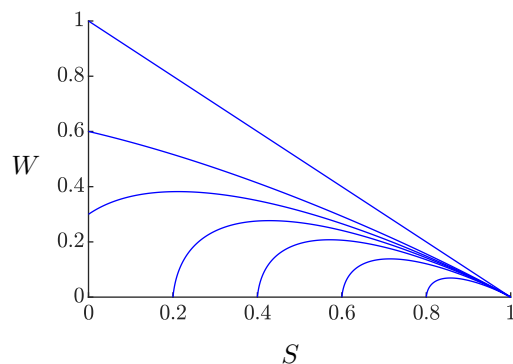


Figure 10: Phase plane for the S, W couple; values for $\kappa = 0.1$ and $\xi = 0.0125$ taken from [4]

Note that, in general, the integral (39) is not explicitly computable. Hence, let $du = [2\kappa R_0 W + \xi(R_0 - u - 1)]d\tau$; then one can transform (39) into an integral equation which provides the exit time T_E , namely, after substituting $du = [2\kappa R_0 W + \xi(R_0 - u - 1)]d\tau$ in (39) one has

$$\int_0^{T_E} u(\tau)d\tau = 0.$$

416 In other words, T_E is defined as the time it takes to go from $u = u_0$ to $u = p_0(u_0)$, and therefore
 417 it is also the time during which a trajectory of (34) stays $\mathcal{O}(\epsilon^2)$ -close to the critical manifold. This
 418 implies, remembering $u(\tau) = R_0 S(\tau) - 1$, that

$$\int_0^{T_E} (R_0 S(\tau) - 1)d\tau = 0. \quad (42)$$

Using the explicit equation for $S(\tau)$ given in (41), and introducing, for ease of notation, $A := 2\kappa + \xi$, $B := 2\kappa(S_\infty + W_\infty - 1)$, $C := S_\infty - 1$ so that

$$S(\tau) = 1 + C \exp(-A\tau) + Bt \exp(-A\tau),$$

419 the equation for the exit time T_E (42) becomes

$$-\frac{R_0 \exp(-AT_E)(ABT_E + AC + B)}{A^2} + (R_0 - 1)T_E + \frac{R_0(AC + B)}{A^2} = 0. \quad (43)$$

420 Clearly $T_E = 0$ is a solution. Moreover, there is only one strictly positive solution, since $S(\tau)$ is
 421 strictly increasing and tends to 1 as $\tau \rightarrow +\infty$. Such solution provides the exit time.

422 Substituting the positive solution T_E of (43) it in (41) we obtain the exit point $(S(T_E), W(T_E))$.
 423 However, due to the implicit formulae we have obtained above, such a computation is only suitable
 424 numerically (see Section 3.4.1). Despite the previous obstacle, we can still check how the exit points
 425 depend on certain parameters. For example, from the first equation of (41) we observe that

$$\frac{\partial S}{\partial \xi}(\tau, \xi) = -\tau[S_\infty - 1 + 2\kappa(S_\infty + W_\infty - 1)\tau] \exp^{-(2\kappa + \xi)\tau} > 0, \quad (44)$$

426 which immediately suggests that the exit time is decreasing in ξ . Namely, let $T_{E,i}$ denote the exit
 427 time with $\xi = \xi_i$ and $i = 1, 2$. If $\xi_1 < \xi_2$ then, using (44), one sees that $T_{E,1} > T_{E,2}$.

428 To provide more insight on the dynamics of the SIRWS model, we are now going to complement
 429 our previous study with a numerical analysis, where the computed exit time T_E shall play an
 430 essential role.

431 3.4.1 Periodic orbits

432 Recall that in the SIR and SIRS models no periodic trajectories are possible. In this section we
 433 show that the SIRWS does have periodic solutions, and of particular biological relevance, stable
 434 limit cycles. Our motivation is that if a stable limit cycle exists, then a disease would have periodic
 435 outbursts. Furthermore, due to the time scales present in the model, there is the danger of missing
 436 such periodicity if only short time scale analysis is considered. Moreover, information regarding
 437 the parameter regions in which damped/sustained oscillations occur can give directions as to which
 438 parameter(s) to modify in order to have a desired control of the epidemic.

439 As it is usual in GSPT, the general idea to show existence of limit cycles of the perturbed
440 (fast-slow) system is to first find a singular cycle, see for example [22, 36]. A singular cycle is
441 a concatenation of limiting slow and fast orbits that form a cycle. Afterwards, given that some
442 conditions are met, we argue that such singular cycle gives rise to a limit cycle of the fast-slow
443 system. We further remark that a mixture of analytical and numerical methods is relevant since
444 we have to combine local analytical results with global numerical results, which is a key theme in
445 multiple time scale systems [12, 13, 23].

446 The steps to form a singular cycle of the SIRWS model are as follows:

- 447 1. Choose a section $J_1 = \left\{ (S, I, W) = (S_0, 0, W) \mid S_0 > \frac{1}{R_0}, W \in (0, 1 - S_0) \right\}$. This section is
448 transversal to the reduced slow flow and is located on the unstable region of the critical manifold.
- 449 2. Consider the map Π_1 defined by the layer equation. Under such a map one obtains a new section
450 on the critical manifold $J_2 := \Pi_1(J_1)$. The coordinates of J_2 are given by $(S_\infty, 0, W_\infty)$, as in
451 Lemma 5.
- 452 3. Consider the map Π_2 defined by the slow flow *for a time* T_E implicitly given by (43), i.e.
453 $\Pi_2(J_2) = (S(T_E), W(T_E))$ with $(S(\tau), W(\tau))$ given by (41), and let $J_3 := \Pi_2(J_2)$. Recall from
454 the last part of section 3.4 that we can tune the exit time, for example, by changing the parameter
455 ξ , without changing the map Π_1 .
- 456 4. If J_3 intersects transversally J_1 , then we have a robust singular cycle given precisely by the orbit
457 corresponding to a fixed point of $\Pi_2 \circ \Pi_1$, see Figure 11 for a schematic representation of these
458 four arguments.

459 In the present context, robust means that the singular cycle persists under small smooth pertur-
460 bations as a periodic orbit of the fast-slow system precisely due to the transverse intersection of
461 J_1 and J_3 [37] (if it occurs).

It is clear that for the particular SIRWS model, there is a priori no guarantee that such a
transverse intersection occurs for a particular set of parameters and initial conditions. To clarify
that indeed such a fixed point exists upon variation of parameter values, we refer to the situation
shown in Figure 12 varying the parameter ξ , we argue as follows: let $F^\xi = (F_1^\xi, F_2^\xi) = \Pi_2 \circ \Pi_1 : \mathcal{C}_0 \rightarrow \mathcal{C}_0$
using the parameter ξ , and $X = \{\xi : J_3 \cap J_1 \neq \emptyset\}$. We can then define, for $\xi \in X$,
 $\bar{w}(\xi)$ as the value of w such that $F_1^\xi(S_0, w) = S_0$. Note moreover that for all w , the inequalities
 $0 < F_2^\xi(S_0, w) < 1 - S_0$ hold, as can be seen by (41).

Consider finally

$$g : X \rightarrow \mathbb{R}, \quad g(\xi) = \bar{w}(\xi) - F_2^\xi(S_0, w)$$

462 If $X = [\xi_1, \xi_2]$, we have $\bar{w}(\xi_1) = 0$ and $\bar{w}(\xi_2) = 1 - S_0$, or vice versa. Hence $g(\xi_1) < 0 < g(\xi_2)$,
463 or vice versa. In either case, there exists $\bar{\xi} \in (\xi_1, \xi_2)$ such that $g(\bar{\xi}) = 0$, i.e. $F_1^{\bar{\xi}}(S_0, \bar{w}(\bar{\xi})) = S_0$
464 and $F_2^{\bar{\xi}}(S_0, \bar{w}(\bar{\xi})) = \bar{w}(\bar{\xi})$ as claimed.

466 Moreover, since we know that both Π_1 and Π_2 are contractions in the W -direction (refer to
467 (34) and to Figure 10), such a singular cycle is locally attracting. Hence it persists as a locally
468 attracting periodic orbit for $\epsilon > 0$ sufficiently small. We remark, however, that this does not
469 mean that there are no other limit cycles for $\epsilon > 0$ sufficiently small. As we show in our numerical
470 analysis of the forthcoming section, there is in fact a range of parameter for which a stable and

471 an unstable limit cycle co-exist. The existence of the unstable limit cycle, however, does not
 472 follow from our previous perturbation arguments.

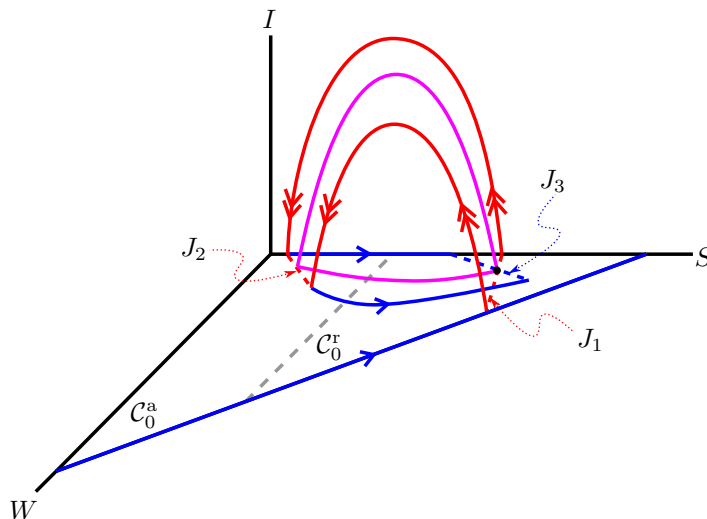


Figure 11: Schematic representation of the singular cycle, shown in magenta. The red arrows depict the map $\Pi_1 : (S_0, W_0) \mapsto (S_\infty, W_\infty)$ so that $\Pi_1(J_1) = J_2$. The blue arrows depict the map Π_2 given by the reduced flow on C_0 and induced by (40) (for a finite time $T_E(S_\infty, W_\infty)$) so that $\Pi_2(J_2) = \Pi_2(\Pi_1(J_1)) = J_3$. If the sections J_1 and J_3 intersect, then such an intersection defines closed singular orbits. If J_1 and J_3 intersect transversally, then such intersection persists for $\epsilon > 0$ sufficiently small giving rise to a periodic orbit of the SIRWS model.

473 Naturally, the above procedure is only sufficient to show existence of limit cycles that pass close
 474 to the critical manifold and provides no information on other possible limit cycles of the fast-slow
 475 system, compare with [35]. Yet our attention is precisely focused on describing those limit cycles
 476 arising from the time scale separation.

477 An example of the above procedure is shown in Figure 12 where we set $\{\beta = 260, \gamma = 17, \kappa =$
 478 $0.1, \xi = 0.0125, \nu = 5\}$, values taken from [4]. Figures in the left column show the evolution of J_1
 479 (dashed red) in the fast system (red) and of J_2 , too small to be visible, in the slow system (blue).
 480 Figures in the right column zoom to the interval J_3 (blue) for each parameter value, and its position
 481 relative to J_1 (dashed red). Note that

- 482 • For $\xi = 0.01$ (Figures 12 (a) and (b)) the interval J_3 lies to the right of J_1 , so there might be a
 483 larger limit cycle further away from J_1 .
- 484 • For $\xi = 0.0125$ (so Figures 12 (c) and (d)) the interval J_3 intersects transversally J_1 , and the
 485 intersection certifies the existence the singular periodic orbit.
- 486 • For $\xi = 0.015$ (so Figures 12 (e) and (f)) the interval J_3 lies to the left of J_1 , so there might be a
 487 smaller limit cycle further away from J_1 , or the system might converge to the unique equilibrium
 488 point in the first octant.

489 It is worth noting that we chose to investigate the role of ξ , the birth/death rate, due to its
 490 biological relevance. However, by the same method one is able to numerically approach the existence
 491 of limit cycles upon variation of any other parameter. It is important to note that, in the limit
 492 systems, there is a clear separation between “fast parameters” (β, γ, ν) and “slow parameters” ($\xi,$
 493 κ); changing a single parameter will only influence either the layer or the reduced dynamics, and
 494 not both.

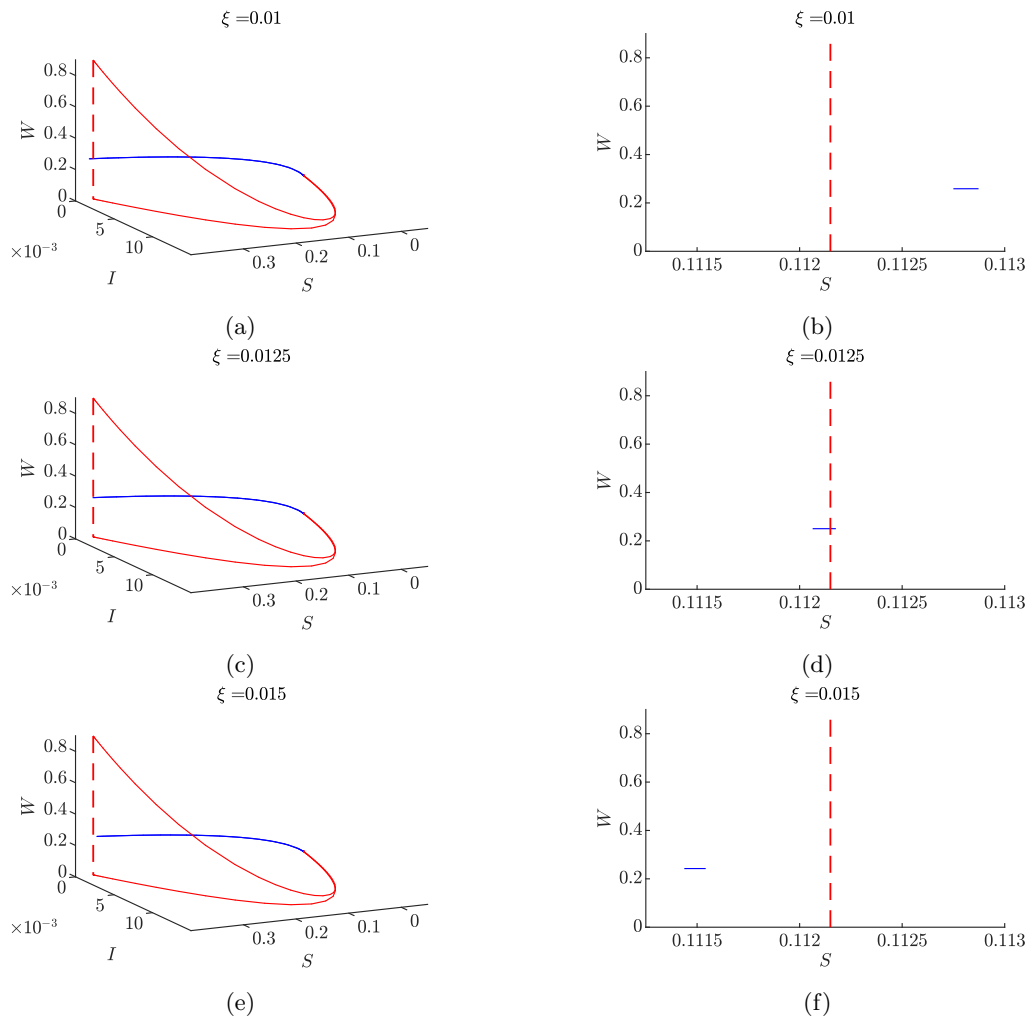
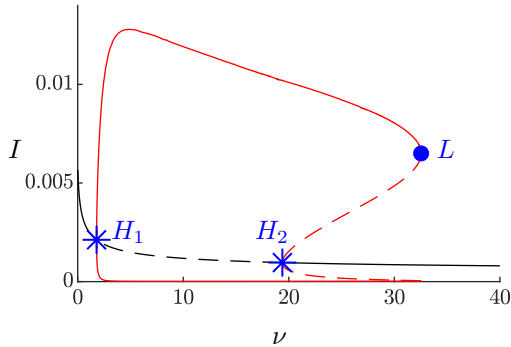
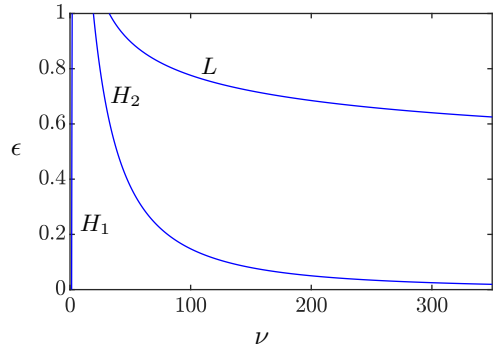


Figure 12: Numerical illustration of the effect of changing ξ on the slow dynamics. This numerical analysis shows that there is an interval around $\xi \sim 0.0125$ for which periodic orbits of (34) exists, for $\epsilon > 0$ sufficiently small.

495 Since we have already demonstrated the existence of limit cycles, the next question to investigate
 496 is the possible bifurcations that may arise upon variation of the parameters. Such analysis is
 497 presented in the forthcoming section.



(a) One-parameter (ν) bifurcation diagram for (34): blue stars labelled H_1 and H_2 correspond to Hopf points; blue dot labelled L corresponds to the Limit Point of Cycles (LPC); red lines correspond to stable (solid) and unstable (dashed) limit cycles; the stable (solid) and unstable (dashed) equilibrium point is depicted by the black line.



(b) The blue lines represent the Hopf points H_1 and H_2 , and the LPC point L , plotted in Figure (13a), which are then continued while decreasing ϵ ; compare with Figure 13a. We observe that H_1 does not tend to $\nu = 0$ as $\epsilon \rightarrow 0$ while H_2 and L diverge.

Figure 13: One and two parameter bifurcation diagrams for (34).

498 3.4.2 Bifurcation analysis

499 In this section we carry out a bifurcation analysis, motivated by the one developed in [4], which we
500 perform with MatCont [8]. Our goal is to investigate the way the bifurcation diagrams change as ϵ is
501 decreased, i.e., we want to understand via numerical continuation how the fast-slow singular limit is
502 approached; see also [7, 11, 17] where such a strategy has considerably improved our understanding
503 of several fast-slow models. In our context, decreasing ϵ means, from a biological point of view,
504 modelling an epidemiological system in which the difference in duration between life expectancy and
505 infectious episodes becomes large. In the limit as $\epsilon \rightarrow 0$, infectious episodes become instantaneous,
506 and the analysis of this limit case helps to understand the behaviour of the system for $\epsilon > 0$ small
507 enough.

508 In fact, we note that the system studied in [4] is system (34), for the particular choice of $\epsilon = 1$.
509 In what follows, we set $\beta = 260$, $\gamma = 17$, $\kappa = 0.1$, as in [4], and vary ϵ , ξ , ν , and later β as
510 well. Notice that the values of the parameters β , γ , κ and ξ already appear of different order of
511 magnitude. It would be possible to use a different parametrization, letting $\tilde{\beta} = 0.26$, $\tilde{\gamma} = 0.017$ and
512 $\epsilon = 0.001$. All the following analysis would be identical, except that the values obtained for ϵ , β
513 and γ would be multiplied by 10^{-3} .

514 For consistency, we start by replicating Figure 5 from [4], by setting $\epsilon = 1$ and $\xi = 0.01$, in
515 Figure 13a. For all parameter values there is a unique equilibrium in $\mathbb{R}_{\geq 0}^4$, as can be easily proved,
516 but its stability changes varying ν through a subcritical and a supercritical Hopf bifurcation.

517 Next, in order to get the dependence of the bifurcation points with respect to ϵ , we continue the
518 two Hopf points H_1 and H_2 and the Limit Point of Cycles (LPC) L in a (ν, ϵ) bifurcation diagram,
519 obtaining the diagram shown in Figure 13b.

520 We notice from Figure 13b that H_1 converges to a positive value for $\nu \sim 1.32$ as $\epsilon \rightarrow 0$, while
 521 H_2 and L diverge; the latter much faster than the former. Moreover, we know from the analysis
 522 performed in Section 3.4 that as $\epsilon \rightarrow 0$ the equilibrium curve (black curve in Figure 13a) approaches
 523 the $\{I = 0\}$ axis. These two observations suggest that as $\epsilon \rightarrow 0$ the bifurcation diagram on Figure
 524 13a gets stretched. One must also point out that the computation of the bifurcation diagrams for
 525 small ϵ becomes considerably expensive due to the high stiffness of the problem.

526 We next produce the analogous to Figure 13a, but for a smaller value of ϵ , namely $\epsilon = 0.05$, in
 527 Figure 14. In order to do so, due to stiffness of the problem, it is necessary to rescale the system
 528 by introducing a new variable $v = \ln(I)$. We emphasize that this rescaling is motivated by the
 529 fact that trajectories get exponentially close to the critical manifold, recall Lemma 2. Moreover,
 530 this rescaling might be useful for bifurcation analysis of systems with similar dynamics in which
 531 an exchange of stability of the critical manifold occur at a non-hyperbolic point, and trajectories
 532 of interest pass exponentially close to such a singularity. With the aforementioned rescaling one
 533 obtains the following system of ODEs:

$$\begin{aligned}
 S' &= -\beta S \epsilon^v + \epsilon(2\kappa W + \xi(1 - S)), \\
 v' &= v(\beta S - \gamma - \epsilon \xi), \\
 W' &= -\nu \beta W \epsilon^v + \epsilon(2\kappa(1 - S - \epsilon^v - W) - 2\kappa W - \xi W).
 \end{aligned}
 \tag{45}$$

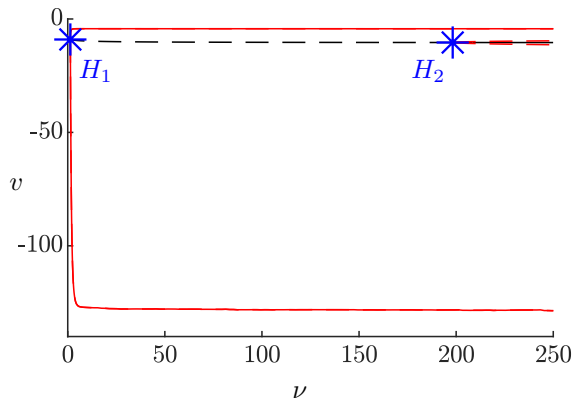


Figure 14: One-parameter (ν) bifurcation diagram for (45): blue stars labelled H_1 and H_2 correspond to Hopf points; red lines correspond to stable (solid) and unstable (dashed) limit cycles; the stable (solid) and unstable (dashed) equilibrium point is depicted by the black line.

534 Thus, the bifurcation diagram in Figure 14 is obtained from (45) and confirms the behaviour
 535 anticipated in Figure 13b: as ϵ decreases, the distance between H_1 and H_2 increases, thus stretching
 536 the parameter region in which stable periodic solutions are to be observed. Most importantly, as is
 537 already evident in Figure 13b, we have that for ϵ sufficiently small the LPC is undetectable, implying
 538 that an eventual transition to stable (endemic) equilibrium due to increase of the immunity boosting
 539 rate ν is not possible any more.

540 Another important parameter is β , which regulates the infection rate. Thus, in order to further
 541 investigate the role of ϵ in the model, we next present in Figure 15 a (ν, β) bifurcation diagram.

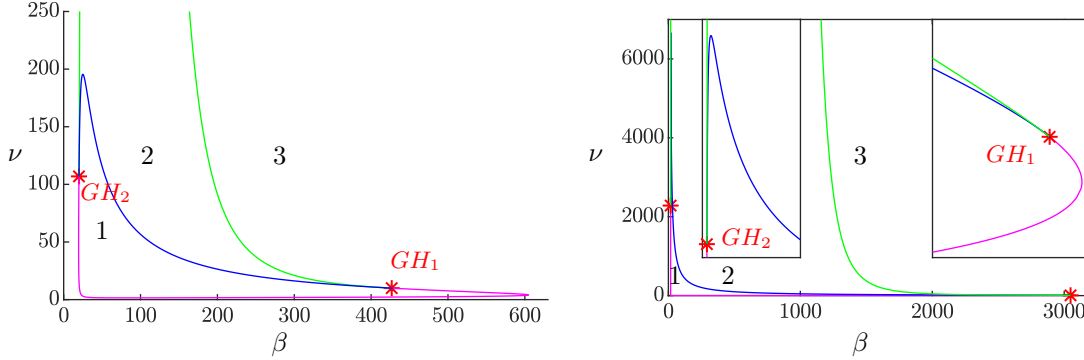


Figure 15: Two parameter bifurcation diagram for (34). Left and right represent $\epsilon = 1$ and $\epsilon = 0.05$, respectively. The red points labelled GH_i are generalised Hopf points. The blue (resp. magenta) branch is a curve of subcritical (resp. supercritical) Hopf bifurcation while the green branches correspond to limit point of cycles. We label the regions in the diagram according to the attractor as 1: Limit cycles, 2: Bistability, and 3: Point attractor. The insets in the right picture are “zoom-ins” near the two GH points.

542 For ease of notation, let us denote by $\nu(P)$ the value of ν corresponding to a point P . From Figure
 543 15 we have that $\nu(GH_1) \approx 9.96$ and $\nu(GH_2) \approx 106.9$ for $\epsilon = 1$. Furthermore, for $\nu \leq \nu(GH_1)$, the
 544 system only exhibits stability of the equilibrium or of the limit cycle (zones 1 and 3). For $\nu(GH_1) <$
 545 $\nu \leq \nu(GH_2)$ there are two intervals of values for β which correspond to a stable equilibrium, one
 546 to a stable limit cycle and one to bistability (zones 1, 2, and 3). For $\nu(GH_2) < \nu \leq \nu_{\max}$, with
 547 $\nu_{\max} \approx 195.46$, there are two intervals of values for β which correspond to a stable equilibrium, one
 548 to a stable limit cycle and two to bistability, one of them being very thin. At $\nu = \nu_{\max}$ the two
 549 Hopf points H_1 and H_2 collide, and a codimension-2 Hopf-Hopf bifurcation occurs.

550 For $\epsilon = 0.05$, the diagram is qualitatively the same, but as already pointed-out before the
 551 diagram gets stretched both in β and in ν . The points GH_1 and GH_2 correspond now to $\nu \approx 7.04$
 552 and $\nu \approx 2282.6$, respectively. In particular, the bistability region 2 is enlarged.

553 To complement the previous description, and similar to Figure 9 (a) to (d) in [4] in Figures 16a-
 554 16c, we present the β -bifurcation diagram for different values of ν and continue all the Hopf points
 555 for decreasing ϵ , as shown in Figures 16d-16f.

556 As before, and for ease of notation, we denote by $\beta(P)$ the value of β corresponding to a point
 557 P . For each value of ν considered, we find two values $17 < \beta(H_1) < \beta(H_2)$ (17 was the fixed value
 558 of γ in each simulation; recall $R_0 = \beta/\gamma$) corresponding to Hopf points, and we continue them in
 559 ϵ , as shown in Figures 16d-16f. For $17 \leq \beta \leq \beta(H_1)$ the equilibrium point is stable, and there is no
 560 limit cycle. For $\beta(H_1) < \beta \leq \beta(H_2)$ the equilibrium point is unstable, and the limit cycle stable.
 561 For $\nu > \nu(GH_1)$ (resp. $\nu > \nu(GH_2)$), there is an interval (resp. there are two intervals) of values
 562 of $\beta(H_2) < \beta \leq \beta(L)$ (with L a LPC, whose existence and position depend on the choice of ν) for
 563 which the system exhibits bistability; eventually these two limit cycles collapse, and for $\beta > \beta(L)$
 564 the system is characterized by a unique asymptotically stable equilibrium. Note, interestingly, that
 565 as the Hopf-Hopf bifurcation is approached, a new LPC (L_2 in Figure 16c) becomes visible.

566 We note that in the limit $\epsilon \rightarrow 0$, one has $\beta(H_1) \rightarrow 17$. This is due to the influence on
 567 the dynamics of the basic reproduction number $R_0 = \beta/\gamma$, which should remain greater than

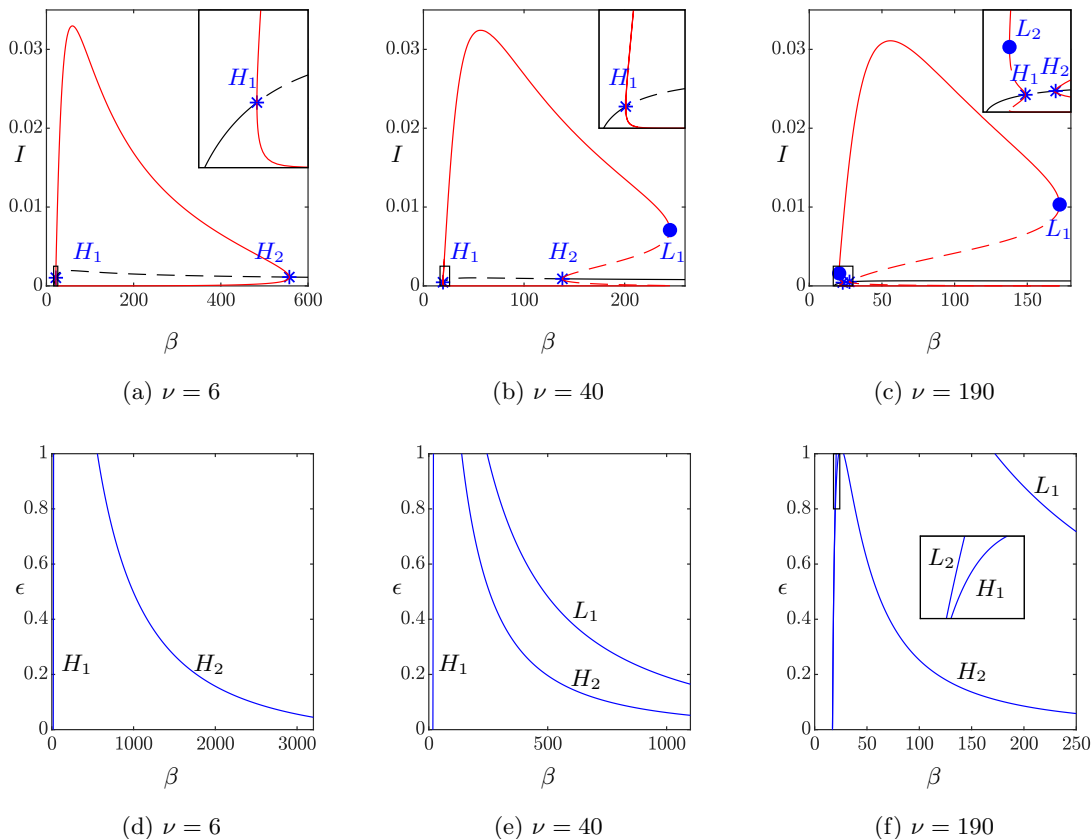


Figure 16: **First row:** one-parameter (β) bifurcation diagram for (34): blue stars labelled H_1 and H_2 correspond to Hopf points; blue circles labelled L_1 and L_2 correspond to Limit Point of Cycles; red lines correspond to stable (solid) and unstable (dashed) limit cycles; the stable (solid) and unstable (dashed) equilibrium point is depicted by the black line. The insets correspond to zoom-in near $\beta = 17$. **Second row:** continuation of the Hopf and LPC points while decreasing ϵ . We observe that H_1 (and L_2 , when it exists) tends to $\beta = 17$ as $\epsilon \rightarrow 0$, while H_2 (and L_1 , when it exists) diverges. The inset in (f) shows a zoom-in at the continuation of H_1 and L_2 from $\epsilon = 1$ to $\epsilon = 0.8$.

568 1 for the endemic equilibrium to exist. Related to this, one has that $\beta(L_2) \rightarrow 17$ as $\epsilon \rightarrow 0$,
569 whenever $\nu > \nu(GH_2)$. The values $\beta(H_2)$ and $\beta(L_1)$, instead, diverge to $+\infty$ as $\epsilon \rightarrow 0$; the region
570 corresponding to the stable limit cycle stretches, as in the ν case. Lastly, we compute a (ξ, ν) -
571 diagram and compare them for $\epsilon = 1$ and $\epsilon = 0.05$ in Figure 17, as we did for (β, ν) in Figure
572 15.

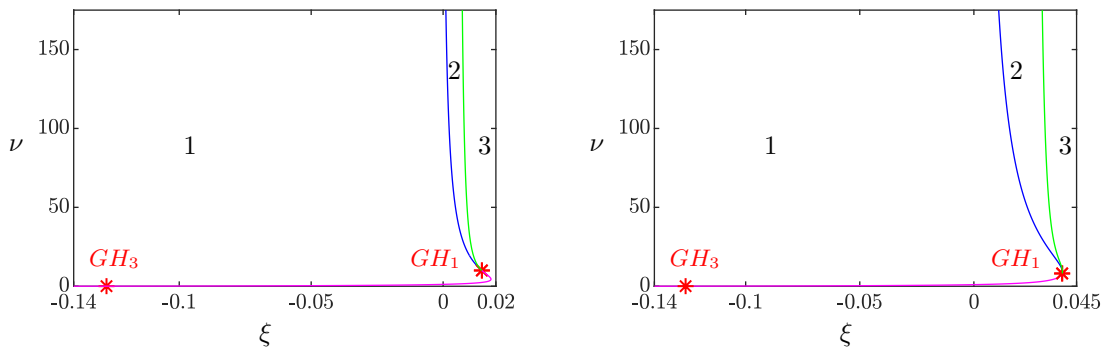


Figure 17: Two parameter bifurcation diagram for (34). Left and right represent $\epsilon = 1$ and $\epsilon = 0.05$, respectively. The red points labelled GH_i are generalised Hopf points. The blue (resp. magenta) branch is a curve of subcritical (resp. supercritical) Hopf bifurcation while the green branch corresponds to a limit point of cycles. Thus, we label the regions in the diagram according to the attractor as 1: Limit cycles, 2: Bistability, and 3: Point attractor.

573 We observe in Figure 17 that not only the bifurcation diagram is stretched as ϵ decreases but
574 also that the bistable region (region 2) is enlarged. GH_1 corresponds to $\xi \approx 0.0147$ for $\epsilon = 1$ and to
575 $\xi \approx 0.03871$ for $\epsilon = 0.05$. Furthermore, in Figure 17 we show the existence of another Generalized
576 Hopf point GH_3 (not considered in [4]), corresponding to $\xi \approx -0.1276$ for $\epsilon = 1$ and to $\xi \approx -0.1263$
577 for $\epsilon = 0.05$. We do not show the 2-parameter continuation of GH_3 since such a computation is
578 not numerically feasible due to the high stiffness of the system in such parameter range. However,
579 the previous observation suggests that all the bifurcation branches corresponding to GH_3 are close
580 to each other.

581 The numerical analysis shown in this section supports the existence of stable limit cycles for an
582 increasing parameter range as $\epsilon \rightarrow 0$. Nonetheless, the dependence of the behaviour of the orbits
583 on the parameters stays the same for sufficiently small parameters. This means that as in the $\epsilon = 1$
584 case, one still observes parameter ranges corresponding to the stability of the endemic equilibrium,
585 and other parameter ranges corresponding to stable periodic orbits.

586 Based on the analysis performed so far, we can now give an interpretation of our results: first of
587 all, the interplay between birth/death rate ξ and immune boosting ν remains qualitatively similar
588 to the one described in [4], for small ϵ . However, the Hopf point H_2 moves according to the
589 increasing difference in the time scales involved in the respective dynamics. H_1 does not converge
590 to 0, supporting the result obtained in [4], where the authors showed that, for ν small enough,
591 the dynamics are close to a SIRS system. The main difference, however, is that as ϵ decreases the
592 role of the parameters can drastically change due to the changes in the bifurcation diagram. For
593 example, for $\epsilon = 1$, a life expectancy of 50 years ($\xi = 0.02$) corresponds to convergence to the

594 endemic equilibrium for all the possible values of ν . In contrast, for smaller values of ϵ the same ξ
595 could correspond to stability of the limit cycle, bistability, or stability of the endemic equilibrium,
596 depending on the value of ν (see Figure 17). Moreover, the effect of increasing life expectancy,
597 i.e. decreasing ξ , results in the transition from point stability to stability of a limit cycle, possibly
598 passing through a region of bistability. This means that, the higher the life expectancy of a certain
599 population, the larger the interval for ν for which a stable limit cycle exists. Biologically, this
600 means that ν must be sufficiently small to obtain a stable endemic equilibrium, otherwise periodic
601 epidemic outbursts turn out to be robust.

602 4 Summary and Outlook

603 We have analysed the behaviour of three models given as a nonstandard singularly perturbed ODE.
604 The first two models presented in Sections 3.1 and 3.3 proved to behave, under mild hypotheses
605 on the parameters, qualitatively in the same way. In particular, their trajectories converge to
606 the only (endemic) equilibrium in the open first quadrant, as long as the initial population of
607 infected individuals is strictly positive. The SIRWS model, instead, proved to be much richer, with
608 parameter regimes allowing for damped oscillations or sustained oscillations, or both.

609 For our analysis we have combined techniques from Geometric Singular Perturbation Theory,
610 and in particular the entry-exit function, introduced in section 2.2. One must point-out that GSPT
611 is usually employed for singular perturbation problems in standard form, and just recently it has
612 been shown that non-standard problems can also be dealt with. More precisely, GSPT allowed us
613 to show the existence of stable limit cycles for certain parameter ranges. Based on such analysis, we
614 further performed numerical studies and computed several insightful bifurcation diagrams, which
615 allowed us to provide a complete qualitative description of the perturbed SIRWS model.

616 We concluded comparing previous results appearing in [4], and extending them by taking into
617 account the role of the (small) parameter ϵ , which does not change the overall qualitative behaviour
618 of the system, but it does drastically change the parameter ranges corresponding to each dynamic
619 regime. Finally, our studies show that GSPT together with numerical tools seem to be suitable to
620 analyze and comprehend epidemiological models with vastly different rates.

621 Once the bifurcation structure of epidemic models is known, one can then be more ambitious
622 and aim to not only control epidemic outbreaks better after they have occurred but even try to
623 anticipate them using early-warning signs [30, 38]. Therefore, our results on bifurcation structure
624 presented here are strongly expected to contribute to the design of these warning signs.

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