Mathematical Modeling of Sterile Insect Technology for Control of Anopheles Mosquito

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Abstract

The Sterile Insect Technology (SIT) is a nonpolluting method of control of the invading insects that transmit the disease. The method relies on the release of sterile or treated males in order to reduce the wild population of anopheles mosquito. We propose two mathematical models. The first model governs the dynamics of anopheles mosquito. The second model, the SIT model, deals with the interaction between treated males and wild female anopheles. Using the theory of monotone operators, we obtain dynamical properties of global nature that can be summarized as follows. Both models are dissipative dynamical systems on the positive cone \(\mathbb{R}_+^4\). The value \(R = 1\) of the basic offspring number \(R\) is a forward bifurcation for the model of the anopheles mosquito, with the trivial equilibrium \(0\) being globally asymptotically stable (GAS) when \(R \leq 1\), whereas \(0\) becomes unstable and one stable equilibrium is born with well determined basins of attraction when \(R > 1\). For the SIT model, we obtain a threshold number \(\hat{\lambda}\) of treated male mosquitos above which the control of wild female mosquitos is effective. That is, for \(\lambda > \hat{\lambda}\) the equilibrium \(0\) is GAS. When \(0 < \lambda \leq \hat{\lambda}\), the number of equilibria and their stability are described together with their precise basins of attraction. These theoretical results are rephrased in terms of possible strategies for the control of the anopheles mosquito and they are illustrated by numerical simulations.

Keywords: sterile insect technology, compartmental modeling, mosquito control, monotone operators.

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1. Introduction

Malaria is one of the most prevalent vector-host diseases, whereby the disease is not transmitted directly from host to host, but through a vector. Malaria

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is caused by a protozoa of the genus plasmodium and is transmitted by the female anopheles mosquito (vector). According to the World Health Organization (WHO), approximately one to three million people die of malaria every year, the vast majority of which are pregnant women and children who live mostly in Africa and in South America. Moreover, it is now acknowledged that northern countries and in particular those in the South of Europe such as France and Italy could become infected again by malaria. Therefore, the control of anophel- les mosquito, the vector responsible for the transmission of diseases, is a major prevention strategy.

Chemicals have been and are still extensively used all over the world to control wild mosquito populations. However, on a long run mosquitos can develop resistance to chemical products. Besides, the WHO only allows a limited number of insecticides in view of polluting disasters. As a viable alternative, nonpolluting methods also known as biological control tools are more and more studied, with a special focus on the ecology and behavior of the involved species. One of the most promising such method is the Sterile Insect Technique/Technology. The SIT is indeed a nonpolluting method of insect control, which relies on the release of sterile males. Mating of released sterile males with wild females leads to nonhatching eggs or to reducing drastically the number of hatching eggs and this drives slowly the wild population to decline. Thus, releasing sufficiently many sterile males and/or doing this over a sufficiently long period of time can lead to local reduction or elimination of the wild population.

The SIT has been known for more than half a century [26, 1]. It was first used in 1954 on the Island of Curacao in the Netherlands Antilles in order to control the new world screwworm fly (Cochliomyia hominivorax (Coquerel)). Sterile males were released for six months after which, the pest was completely eliminated [4]. Among regions where other releases were done, we can mention Southern USA, Mexico, Central America [41], and Lybia [30]. Since then, the SIT has been used successfully to almost eradicate various insect populations ranging from pest species (e.g. Mediterranean fruit fly), Ceratitis capitata Wiedeman [22], codling moth Cydia pomonella (L.) [9, 35], tsetse fly Glossina austeni [35], Culex quinquefasciatus to the vector of Bancroftian filariasis [36]. The book [18] provides a comprehensive overview on the SIT and its applications.

As far as the anopheles vector is concerned, the largest SIT release programme against specifically the anopheles albimanus took place in El Salvador in the 1970s [31], over a 5-month period. About 4.3 million mosquito pupae were mass-produced, sterilized, and released around Lake Apastepeque. The analysis of the data of the anopheles albimanus population [11] from the release and the nearby control area demonstrated how effective the sterile males were in preventing a normal seasonal rise in the vector density. Subsequently, a more extensive trial took place from 1977 to 1979 on the Pacific coast of El Salvador [32] with up to 0.5 million sterile males or 1.25 million sterile male pupae being released every day. Recently (2007) in Italy, Bellini [10], released sterile males, which contributed to reduce the wild population of aedes albopictus and to fight the Chikungunya virus [17] (see [15, 16] for further details about Chikungunya). Furthermore, mosquitos genetically modified by using the RIDL (Release of
Insects carrying a Dominant Lethal Technique, were released in the Caíman Island and in Malaysia, by the Oxitec Company to fight Dengue Fever (see [25] for an overview on RIDL approach for *Aedes aegypti*).

Since the first field releases, various modelling and/or mathematical works have been done on SIT using either discrete models [12, 29], or continuous temporal models with continuous release (see for instance [5, 6, 19, 20, 28, 29, 38] and references therein), with pulsed releases [17, 42], or spatio-temporal models with one dimensional spatial component and continuous (proportional) releases [20, 34, 39]. See also [7] for an overview on SIT mathematical modelling.

The mathematical models investigated in this work are designed according to the approach in [13], [14], where the transmission of malaria is modelled by including variability in population in such a way that both the human and the anopheles mosquito populations follow the logistic growth law. Though having the same compartmental structure as the SIT study in [19] regarding the control of *aedes aegypti*, also investigated in [38], the differential equations in our model are constructed differently. Furthermore, the equation for the sterile males is much simpler (for other models, see for instance [29]). The purpose of this work is twofold. Firstly, for the dynamical systems under consideration, we are interested in properties of global nature, including the dissipativity of the system, the global asymptotic stability of the trivial equilibrium and the precise description of the basins of attraction of multiple stable equilibria. This goal is achieved by an alternative approach, namely, the theory of monotone operators [37], [40], which unlike many classical studies is not subjected to any Lyapunov function. The major advantage of the obtained global properties of dynamical systems is translated into the second purpose of the work. That is to systematically analyse the impact of the SIT, as a measure for the control of the anopheles mosquito population. We identify efficient strategies that lead to the reduction of the mosquito population below a certain threshold, which is epidemiologically relevant.

The rest of paper is organized as follows. In the next section, we give the properties of monotone operators that are relevant to our study. This is followed (Section 3) by the presentation of the basic mathematical model of wild anopheles mosquito population and by the discussion of its global dynamical properties. Section 4 is devoted to the study of the SIT mathematical model, where the key finding is the identification of a threshold number of treated male mosquitoes or sterile mosquitoes above which the control of wild female mosquitoes is effective. The theoretical results are discussed and supported by numerical simulations in Section 5. The decrease of the wild population of female mosquitoes upon the release of sterile male mosquitoes is further characterized in Section 6 in terms of the so-called yield number of the SIT, which is a certain measure of the reduction of the wild population. Concluding remarks as to how this work fits in the literature and can be extended are given in Section 7. For convenience, we have included an appendix for the proofs of the three main results, which as indicated earlier, are based on the theory of monotone operators [2].
2. Preliminaries on monotone operators

Consider the system of ODEs
\[
\frac{dx}{dt} = g(t, x),
\]
where \( D \subseteq \mathbb{R}^n \) and \( g : [0, +\infty) \times D \to \mathbb{R}^n \) is continuous. Typically \( D \) is assumed open to avoid complications. However, in view of the models in this paper we assume only that \( D \subset \text{closure}(\text{interior}(D)) \) and that for some \( \delta > 0 \) the vector fields defined by \( g(t, \cdot) \), \( t \in [0, \delta) \), are all directed inwards at the points of \( \partial D \).

This is enough to ensure that for every \( a \in D \) there exists \( T_a > 0 \) such that the system (1) has a solution \( x(a, t) \) on the interval \( [0, T_a) \) which satisfies \( x(a, 0) = a \).

We further assume that \( g \) is such that the solution initiated at \( a \) is unique. To avoid new notations we assume that \([0, T_a)\) is the maximal (nonnegative) interval of existence of \( x(a, t) \).

**Definition 1.** The system (1) is called cooperative if for every \( i, j \in \{1, 2, \ldots, n\} \) such that \( i \neq j \) the function \( g_i(t, x_1, \ldots, x_n) \) is monotone increasing with respect to \( x_j \).

If \( g \) is differentiable and the domain \( D \) is convex then, equivalently, the system is cooperative if the Jacobian \( \frac{dg(t, x)}{dx} \) is a Metzler matrix for every \( t \in [0, +\infty) \) and \( x \in D \). Let us recall that a matrix is called Metzler if its nondiagonal entries are nonnegative [8].

The following theorem is often referred to as Kamke’s theorem, see e.g [23].

**Theorem 2.** Let system (1) be cooperative. Then for every \( a, b \in D \)
\[
a \leq b \implies x(a, t) \leq x(b, t), \ t \in [0, \min\{T_a, T_b\}].
\]

The condition on \( g \) in Definition 1 is sometimes called quasimonotonicity with respect to \( x \). It is linked in [40] to the monotonicity of differential operators. Kamke’s theorem follows from a more general inequality given in [40].

**Theorem 3.** If \( g \) is quasimonotone with respect to \( x \) then for any two differentiable functions \( y, z : [0, T) \to D \) we have
\[
\begin{align*}
y(0) &\leq z(0) \\
y'(t) - g(t, y(t)) &\leq z'(t) - g(t, z(t)), \ t \in [0, T]
\end{align*}
\]

\[\implies y(t) \leq z(t), \ t \in [0, T].\]

The inequalities between vectors are considered here in their usual coordinate-wise sense, that is, for any \( a, b \in \mathbb{R}^n \),
\[
a \leq b \quad \iff \quad a_i \leq b_i, \ i = 1, \ldots, n.
\]

In addition, we use the relations
\[
a < b \quad \iff \quad a \leq b, \ a \neq b, \\
a \ll b \quad \iff \quad a_i < b_i, \ i = 1, \ldots, n.
\]
Most of the theory for cooperative systems is developed for the case of autonomous systems
\[ \frac{dx}{dt} = f(x), \quad (2) \]
where the right hand side is independent of \( t \). In this case, Theorem 2 equivalently means that the evolution semi-group operator \( \varphi_t : D_t \rightarrow D \) defined by \( \varphi_t(a) = x(a,t) \) is monotone increasing on its domain \( D_t = \{ a \in D : T_a > t \} \) for every \( t > 0 \). Hence also the name monotone systems [37]. A system of the form (2) is called irreducible if its Jacobian \( \frac{df}{dx} \) is an irreducible matrix for every \( x \in D \). For cooperative irreducible systems the Kamke’s theorem admits a stronger form as stated below, see [37, Theorem 4.1.1].

**Theorem 4.** If the system (2) is cooperative and irreducible then for every \( a,b \in D \)
\[ a < b \implies x(a,t) \ll x(b,t), \quad t \in [0, \min\{T_a, T_b\}). \]

The next theorem characterizes monotone solutions of cooperative systems and is part of [37, Proposition 3.2.1].

**Theorem 5.** Assume (2) is cooperative, and let \( a \in D \) such that \( f(a) \geq 0 \) \( (f(a) \leq 0) \) then the solution \( x(a,t) \) is monotone increasing (decreasing) function of \( t \in [0, T_a) \).

The combined application of the monotonicity of the evolution operator \( \varphi_t \) given in Theorems 2 and 4 and the monotonicity of the solutions given in Theorem 5 is an efficient tool for studying dynamical systems. Let us recall that (1) defines a dynamical system on \( D \) if \( x(a,t) \) is defined for all \( t \geq 0 \) and \( x(a,t) \in D \). In terms of the notation introduced earlier, this means that \( T_a = \infty \) for all \( a \in D \). Using the monotonicity theorems stated here, one can prove, for example, stability and attractiveness of equilibria without using the Jacobian of the right hand side. Furthermore, this approach provides a method for characterizing the basins of attraction and addressing other related issues. As usual we call an equilibrium asymptotically stable if it is both stable and attractive. An asymptotically stable equilibrium is called globally asymptotically stable if the basin of attraction is the whole domain \( D \). Basins of attraction are often represented as \( n \)-dimensional intervals: given \( a, b \in \mathbb{R}^n \) with \( a \leq b \)
\[ [a,b] = \{ x \in \mathbb{R}^n : a \leq x \leq b \}. \]

**Theorem 6.** Let \( a,b \in D \) be such that \( a < b \), \( [a,b] \subseteq D \) and \( f(b) \leq 0 \leq f(a) \). Then (1) defines a (positive) dynamical system on \([a,b]\). Moreover, if \([a,b]\) contains a unique equilibrium \( p \) then \( p \) is globally asymptotically stable on \([a,b]\).

**Proof :** Under the stated assumptions it follows from Theorem 5 that \( x(a,t) \) and \( x(b,t) \) are respectively monotone increasing and monotone decreasing. Moreover, Theorem 2 implies that \( x(a,t) \leq x(b,t), \quad t \in [0, \min\{T_a, T_b\}). \)
Therefore, at least one of the solutions remains in $[a, b]$ in its maximal interval of existence. Then the compactness of $[a, b]$ implies that $T_a = T_b = \infty$ so that for any $z \in [a, b]$ we have $T_z = \infty$ and
\[ a \leq x(a, t) \leq x(z, t) \leq x(b, t) \leq b, \quad t \in [0, \infty). \]

Therefore, (1) defines a (positive) dynamical system on $[a, b]$. The global asymptotic stability of a unique equilibrium $p \in [a, b]$ follows from a general theorem [27], but in this setting it also admits an elementary proof. Due to the monotonicity of system (2), both solutions converge to points of $[a, b]$ as $t \to \infty$. Since these limits are equilibria of the dynamical system and $p$ is the only equilibrium in $[a, b] \cap D$, it follows that $\lim_{t \to \infty} x(a, t) = \lim_{t \to \infty} x(b, t) = p$. It remains to show the stability of $p$. Assume first that $a < p < b$. Let $V$ be an open neighborhood of $p$ in $[a, b]$. Then there exists $t_1 > 0$ such that $x(a, t) \in V$ and $x(b, t) \in V$ for all $t \geq t_1$. Clearly the set $W = (x(a, t_1), x(b, t_1))$ is an open neighborhood of $p$. An application of Theorem 2 yields
\[ x(a, t_1 + t) < x(z, t) < x(b, t_1 + t), \quad z \in W. \]

Therefore, $x(z, t) \in W \subset V$ for every $z \in W$, which proves the stability of $p$. When $a = p$ and $b = p$, a simplified version of the arguments above can be used.

3. Population dynamics of anopheles mosquito

The life cycle of a mosquito consists of two main stages: aquatic (egg, larva, pupa) and adult (with males and females). After emergence from pupa a female mosquito needs to mate and get a blood meal before it starts laying eggs. Then every 4-5 days it will take a blood meal and lay 100-150 eggs at different places (10-15 per place). For the mathematical description, we will consider two stages [6]: the aquatic stage and the adult stage. Furthermore, we split the adult stage into three sub-compartments, with females and males, which leads to the following compartments:
A - population in aquatic stage
Y - young females, not yet laying eggs
F - fertilized and eggs laying females
M - number of males

The life cycle is described through the flow chart on Fig. 1. The mathematical model is the system of ordinary differential equations (3)–(6).

\[
\frac{dA}{dt} = \phi F - (\gamma + \mu_1 + \mu_2 A)A, \quad (3)
\]

\[
\frac{dY}{dt} = r\gamma A - (\beta + \mu_Y)Y, \quad (4)
\]

\[
\frac{dF}{dt} = \beta Y - \mu_F F, \quad (5)
\]

\[
\frac{dM}{dt} = (1 - r)\gamma A - \mu_M M. \quad (6)
\]

Note that Eq. (3) can be considered as a logistic population with immigration. Following [13], [14], it is formulated by using density dependent mortality rate, \(-\mu_2 A^2\), rather than carrying capacity. A female needs to mate successfully only once. The eggs are laid in the so-called gonotrofic cycle. It consists of taking a blood meal, maturation of the eggs and oviposition. Before a female begins laying eggs, two essential events need to take place, mating and taking a blood meal, occurring in varying order. We consider a female to be in the Y compartment from its emergence from pupa until her gonotrofic cycle has began, that is the time of mating and taking the first blood meal, which takes typically 3-4 days. The death rate during that period reflects essentially only death from predators and adverse climatic conditions. Therefore, it is generally lower than the death rate for the F compartment. Typically the male mosquitos are (depending on the temperature) about half or at least 40% of the total population. In the model the fraction of the emerging female mosquitoes is denoted by \(r\), with \(1 - r\) being the fraction of emerging male mosquitoes. Mating is a complex process that is not fully understood [24]. The male mosquito can mate practically through all its life. A female mosquito needs one successful mating to breed its lifelong. It is admitted that mosquitoes locate themselves in space and time to ensure they are available to mate [24]. Therefore, it is reasonable to assume that in any case the immature female will mate and, thus, move to compartment F, or die. Thus \(1/(\beta + \mu_Y)\) represents the mean time a female stays in compartment Y. Mathematically this means that equation (5) can be decoupled from the system. Sometimes \(\beta\) is referred to as "mating rate", which, as explained above, can be a bit misleading and does not define well the boundary between compartments Y and F: the terminology contact rate would be better for \(\beta\). We clearly fixed this boundary at the beginning of the first gonotrofic cycle of a female, that is immediately after the mating.
and first blood meal. Then the rate (per day) of laying eggs in the breeding sites is \( \phi F \), where \( \phi \) is the average amount of eggs laid per fertilized female per day. In the model, we use density a dependent death rate for the aquatic stage since anopheles larvae are density sensitive, which imply an additional density mortality rate. In [19], the size of the population is also restricted only in the aquatic stage but in a different way by an explicit carrying capacity beyond which no eggs are laid. In equation (3) the parameters \( \mu_1 \) and \( \mu_2 \) denote, respectively, the density independent and the density dependent death rates of the aquatic stage. In equations (4), (5), and (6), \( \mu \) with respective index refers to the death rate for the specific compartment (which is density independent). The system (3)–(6) has two equilibria: the origin 0 and the nontrivial equilibrium \( x^\# = (A^\#, Y^\#, F^\#, M^\#)' \) given by

\[
A^\# = \frac{\gamma + \mu_1}{\mu_2}(R - 1), \quad Y^\# = \frac{r\gamma(\gamma + \mu_1)}{\mu_2(\beta + \mu_Y)}(R - 1),
\]

\[
F^\# = \frac{r\gamma\beta(\gamma + \mu_1)}{\mu_2\mu_F(\beta + \mu_Y)}(R - 1), \quad M^\# = \frac{(1 - r)\gamma(\gamma + \mu_1)}{\mu_2\mu_M}(R - 1),
\]

where \( R = \frac{\phi\beta r\gamma}{(\beta + \mu_Y)(\gamma + \mu_1)\mu_F} \) is the basic offspring number. The nontrivial equilibrium \( x^\# \) is nonnegative, that is, it has a biological meaning if and only if \( R \geq 1 \).

Denoting \( x = (A, Y, M, F)' \) the system (3)–(6) can be written in the form (2) where the function \( f \) is defined via the right hand side of (3)–(6). It is easy to see that the system is cooperative on \( D = \mathbb{R}_+^4 \). Moreover, \( f \) is continuous on \( D \) and the vector field defined by \( f \) is directed inwards on \( \partial D \). Hence Theorems 2, 5 and 6 are applicable.

The essential properties of the model (3)–(6) as a dynamical system are summarized in the following theorem. A major point that differentiates this theorem and its proof from those in the literature (e.g. [19]) are the global nature of the properties and the irrelevance of the Lyapunov function. (For an alternative approach see [19])

**Theorem 7.** The set of ODEs (3)–(6) defines a dissipative dynamical system on \( D = \mathbb{R}_+^4 = \{ x \in \mathbb{R}^4 : x \geq 0 \} \). Moreover,

(i) If \( R \leq 1 \) then 0 is globally asymptotically stable on \( D \).

(ii) If \( R > 1 \) then the system has two equilibria 0 and \( x^\# \) on \( D \) where \( x^\# \) is stable with basin of attraction \( D \setminus \{ x = (A, Y, M, F) \in \mathbb{R}_+^4 : A = Y = M = F = 0 \} \) and 0 is unstable with the nonnegative \( M \)-axis being a stable manifold.

**Proof:** The inequality

\[
\frac{\gamma + \mu_1 + \mu_2 A}{\gamma + \mu_1} > 4R
\]
holds for all sufficiently large $A$. Let $m > 0$ and let $A_m$ be so large that in addition to (9) the following inequalities also hold:

\begin{align}
A_m & \geq m, \\
F_m & := \frac{(\gamma + \mu_1 + \mu_2 A_m)A_m}{2\phi} \geq m, \\
Y_m & := \frac{\mu F}{2\beta} F_m \geq m, \\
M_m & := \frac{2(1 - r)\gamma A_m}{\mu M} \geq m.
\end{align}

Let $b_m = (A_m, Y_m, F_m, M_m)$. Then

\begin{equation}
\begin{pmatrix}
-\phi F_m \\
r\gamma A_m \left(1 - \frac{2 + \mu_1 + \mu_2 A_m}{4R(\gamma + \mu_1)}\right) \\
-\beta Y_m \\
-\frac{1}{2}\mu M M_m
\end{pmatrix}
< 0.
\end{equation}

Applying Theorem 6 with $a = 0$ and $b = b_m$ we obtain that (2) defines a dynamical system on $[0, b_m]$. However, $b_m$ can be selected larger than any $x \in \mathbb{R}^4_+$. Hence, (2) defines a dynamical system on $D = \mathbb{R}^4_+$.

(i) In this case the only equilibrium in $D$ is the origin $0$. It follows from Theorem 6 that $0$ is globally asymptotically stable on $[0, b_m]$ for any $m > 0$. Hence it is globally asymptotically stable on $D$.

(ii) Since $R > 1$ the inequality

\begin{equation}
\frac{\gamma + \mu_1 + \mu_2 A}{\gamma + \mu_1} < \sqrt{R}
\end{equation}

holds for all sufficiently small values of $A$. Let $\varepsilon > 0$ and let $A_\varepsilon$ be so small that in addition to (15) the following inequalities also hold

\begin{align}
A_\varepsilon & \leq \varepsilon, \\
F_\varepsilon & := \sqrt{R} \left(\frac{\gamma + \mu_1 + \mu_2 A_\varepsilon}{\phi}\right) \leq \varepsilon, \\
Y_\varepsilon & := \sqrt{R} \frac{\mu F}{3} F_\varepsilon \leq \varepsilon, \\
M_\varepsilon & := \frac{(1 - r)\gamma A_m}{\mu M \sqrt{R}} \leq \varepsilon.
\end{align}
Let \( a_\varepsilon = (A_\varepsilon, Y_\varepsilon, F_\varepsilon, M_\varepsilon)' \). Then

\[
\begin{pmatrix}
1 - \frac{1}{\sqrt{R}} & \phi F_\varepsilon \\
1 - \frac{\gamma + \mu_1 + \mu_2}{\sqrt{R(\gamma + \mu_1)}} & r_\gamma A_\varepsilon \\
1 - \frac{1}{\sqrt{R}} & \beta Y_\varepsilon \\
(\sqrt{R} - 1)\mu M_\varepsilon & 0
\end{pmatrix} > 0.
\] (20)

Hence it follows from Theorem 6 that \( x^\# = (A^\#, Y^\#, F^\#, M^\#)' \) is globally asymptotically stable on \([a_\varepsilon, b_m]\). Since \( a_\varepsilon \) can be selected to be smaller than any \( x > 0 \) and \( b_m \) can be selected to be larger than any \( x > 0 \) we have that \( x^\# \) is asymptotically stable on \( D = \mathbb{R}_+^4 \) with basin of attraction at least the interior of \( D \). We prove that the basin of attraction is \( \tilde{D} = D \setminus \{x = (A, Y, M, F) \in \mathbb{R}_+^4 : A = Y = F = 0\} \) by using that the system is irreducible on \( \tilde{D} \), which can be seen easily. Let \( z \in \tilde{D} \). Then \( z > 0 \) and it follows from Theorem 4 that \( x(z, t) \gg 0 \) for \( t > 0 \). Hence by what has been proved already \( \lim_{t \to +\infty} x(z, t) = x^\# \). Moreover, on the \( M \)-axis equation (5) is reduced to a decreasing equation. Hence all solutions converge to \( 0 \). Therefore, the basin of attraction of \( x^\# \) is precisely \( \tilde{D} \). This also implies that \( 0 \) is unstable with the \( M \)-axis being a stable manifold.

Note that the basic offspring number \( R \) does not depend on \( \mu_2 \) while the equilibrium values of all compartments are inversely proportional to \( \mu_2 \). Having \( \mu_2 > 0 \) is essential for the dissipativity of the system. For \( R > 1 \) the equilibrium \( x^\# \) has also the role of a carrying capacity although it does not appear explicitly in the formulation of the model.

4. The SIT model

The SIT is a nonpolluting method of insect control that relies on the release of sterile insects. Mating of released sterile males with wild females leads to nonhatching eggs. Thus, if males are released in sufficient numbers and/or over a sufficiently long period of time, it can lead to the local reduction or elimination of the wild population. Typically the aim is to lower this population under a certain threshold so that the entomological risk is low. Thus SIT can be also used as a preventive tool, or after a standard vector control campaign, with adulticide, larvicide or/and mechanical control. The success of the SIT is based on the fact that the sterile mosquito are in many ways the same as the wild ones, most importantly, they will mate with wild females. Nevertheless there are some differences. One for example is their distribution. Since they are distributed manually, the place of release depends on available roads and resources as well as on the level of knowledge about the distribution of wild mosquitoes (breeding sites, feeding grounds). Therefore we assume that only a fraction \( p \) of the
released mosquito can join the wild mosquito population. Further difference is some change in the biology, like the mating competitiveness of the sterile male mosquito, due to irradiation [21]. In general, this can be captured by a ratio \( q \), representing the mean mating competitiveness of the sterile males [24]. Note that \( q \) could be less or greater than 1 [10, 24]. We denote by \( M_T \) the number of "wild mosquitoes equivalent" of sterile mosquitoes. This means that the actual number of sterile mosquitos is \( \frac{1}{pq} M_T \). The death rate \( \mu_T \) also depends on the procedure. Given that the sterile mosquitos are released at a rate of \( \psi(t) \) at time \( t \), the population of treated males \( M_T \) is modeled by equation (26). Under the assumption that, after the stated adjustments, the mosquitos in the compartments \( M \) and \( M_T \) are equally likely to mate, a female mating female mosquito has probability \( \frac{M}{M + M_T} \) to be with wild mosquito and probability \( \frac{M_T}{M + M_T} \) to be with a sterile mosquito. Hence the transfer rate \( \beta \) from the compartment \( Y \) splits into transfer rate of \( \frac{\beta M}{M + M_T} \) to compartment \( F \) and a transfer rate of \( \frac{\beta M_T}{M + M_T} \) to compartment \( U \) of females that would be laying sterile (not hatching) eggs. Note that the total mating rate \( \frac{\beta M}{M + M_T} + \frac{\beta M_T}{M + M_T} = \beta \) remains unchanged by the introduction of the sterile mosquito. We should remark that the different strategy used in [19] results in the biologically unrealistic situation of lowering the mating rate due to the introduction of the sterile mosquitos. A modified flow diagram involving also sterile males is given in Fig. 2. Then the mathematical model is represented as the system of 6 differential equations (21)–(26). Note that, in contrast to the model (3)–(6), neither \( M \) nor \( M_T \) can be decoupled. The effect of the introduction of the sterile mosquitos is in the reduction of the \( Y \) to \( F \) transfer rate from \( \beta \) to \( \frac{\beta M}{M + M_T} \). The females fertilized by sterile mosquitos also lay eggs, but these eggs do not hatch. Therefore they have no effect on the aquatic stage of the population. Other techniques, like RID, lead to eggs which hatch and the larvae die only later, having further negative impact on the wild larva population by competing with it for resources, i.e. reducing the carrying capacity of the breeding sites. This is not the case with the SIT technique. Only the wild mosquito develop in the breeding sites. Let us note that in some previous models, e.g. [33], the sterile mosquito population is assumed to have a direct impact on the available carrying capacity for the wild population, which is biologically incorrect.
The model (21)–(26) can be simplified in the following way. First, the equation for $U$ can be decoupled from the system. Secondly, the size of $M_T$ is controlled by human intervention and independent from the rest of the population. Indeed, given a continuous function $\psi(t)$ the linear equation (26) has the solution

$$M_T(t) = e^{-\mu_T t} \left( M_T(0) + \int_0^t e^{\mu_T s} \psi(s) ds \right).$$

(27)
Then, the mathematical model is a nonautonomous system of four differential equations as follows:

\[
\begin{align*}
\frac{dA}{dt} &= \phi F - (\gamma + \mu_1 + \mu_2 A)A, \\
\frac{dY}{dt} &= r\gamma A - (\beta + \mu_Y)Y, \\
\frac{dM}{dt} &= (1 - r)\gamma A - \mu_M M, \\
\frac{dF}{dt} &= \frac{\beta M}{M + M_T(t)} Y - \mu_F F. 
\end{align*}
\] (28) (29) (30) (31)

In addition to having fewer equations, a major advantage of the system (28)–(31) is that its right hand side is quasi-monotone. This fact in turn explains why the badly needed global properties in Theorem 8 and Theorem 9 below are obtained at low cost i.e. “no Lyapunov function”.

**Theorem 8.** Assume that \( \psi(t) \) is a continuous nonnegative valued function of \( t \in [0, \infty) \) so that \( M_T = M_T(t) \) in (27) has the same property. Then the system of ODEs (28)–(31) defines a dissipative dynamical system on \( D = \mathbb{R}^4_+ \).

**Proof:** We use the notation as in the proof of Theorem 7. The system (28)–(31) is nonautonomous. Denoting its right hand side by \( g(t, x) \) it assumes the form (1). The function \( g \) depends on \( t \) via \( M_T(t) \), that is, it can be written as \( g(t, x) = \Phi(M_T(t), x) \). It is easy to see that \( g(t, x) = \Phi(M_T(t), x) \leq \Phi(0, x) = f(x) \) where \( f \) is the right hand side of (3)–(6). Let \( z \in D \). Denote by \( x(z, t) \) and \( x_0(z, t) \) the solutions of (28)–(31) and (3)–(6), both initiated at \( z \). As shown in Theorem 7, \( x_0(z, t) \) is defined for all \( t \geq 0 \). Let \( [0, T_z) \) be the maximal interval of existence of \( x(z, t) \). We have

\[
\frac{d}{dt} x(z, t) - f(x(z, t)) \leq \frac{d}{dt} x(z, t) - g(t, x(z, t)) = 0 = \frac{d}{dt} x_0(z, t) - f(x_0(z, t))
\]

Then it follows from Theorem 3 that \( x(z, t) \leq x_0(x, t), t \in [0, T_z) \). Furthermore, using that \( 0 \) is an equilibrium of (28)–(31) we obtain by Theorem 2 that \( x(z, t) \geq 0, t \in [0, T_z) \). Now using the dissipativity of (3)–(6) as obtained in Theorem 7, it is easy to obtain that \( T_z = +\infty \) and that \( x(z, t) \) is eventually absorbed into a neighborhood of either \( 0 \) or \([0, x_#]\).

The dynamical system (28)–(31) is not autonomous. The right hand side depends directly on \( t \) via the function \( M_T(t) \). In order to characterize the behavior of the solutions more specifically than in Theorem 8 we need further knowledge about the function \( M_T \). In what follows we assume that the manual intervention \( \psi(t) \) is such that the function in (27) satisfies:

\[ \lim_{t \to \infty} M_T(t) = \lambda \text{ for some } \lambda \in (0, +\infty). \] (32)

For sufficiently large values of \( t \) equation (31) can be replaced by the simpler equation

\[
\frac{dF}{dt} = \frac{\beta M}{M + \lambda} Y - \mu_F F. \] (33)
Moreover, in addition to providing approximation to (28)–(31) for large \( t \) the system (28)–(30), (33) has the same invariant sets and respective stability. Hence, we carry out our further analysis on the autonomous dynamical system (28)–(30), (33). Equating the right hand side to zero we obtain that at any equilibrium of this system the ratio \( \alpha = \frac{M}{\lambda} \) satisfies the equation

\[
\alpha^2 - (R - \lambda Q - 1)\alpha + \lambda Q = 0,
\]

where \( R \) is the basic offspring number (8) of the wild mosquito population and

\[
Q = \frac{\mu_2\mu_M}{(\gamma + \mu_1)(1 - r)\gamma}.
\]

In view of Theorem 7 we are only interested in the SIT model (28)–(30), (33) when \( R > 1 \) which we assume in the sequel. It is easy to see that \( \hat{\lambda} = \left(\frac{\sqrt{R} - 1}{Q}\right)^2 \) is a threshold value of \( \lambda \) determining the number of positive roots of equation (34). More precisely, if \( \lambda < \hat{\lambda} \), then \( \hat{\lambda} \) has two positive roots \( \alpha^* \) and \( \alpha^{**} \) (\( \alpha^* > \alpha^{**} \)), if \( \lambda = \hat{\lambda} \) it has one positive root \( \hat{\alpha} \), and if \( \lambda > \hat{\lambda} \) there are no positive roots. Let \( \hat{x} = (\hat{A}, \hat{Y}, \hat{M}, \hat{F})' \), \( x^* = (A^*, \hat{Y}^*, \hat{M}^*, F^*)' \) and \( x^{**} = (A^{**}, \hat{Y}^{**}, \hat{M}^{**}, F^{**})' \) be the equilibria of the dynamical system (28)–(30), (33) corresponding to \( \hat{\alpha} \), \( \alpha^* \) and \( \alpha^{**} \), respectively. It is easy to see that \( 0 < \hat{x}, 0 < x^* < x^{**} \) whenever these equilibria exist (see Appendix for detailed computations). The properties of the model as a dynamical system are summarized in the next theorem.

**Theorem 9.** The equations (28)–(30), (33) define a dissipative dynamical system on \( D = \mathbb{R}^4_+ \) for any \( \lambda \in (0, +\infty) \). Moreover, we have that:

(a) If \( \lambda > \hat{\lambda} \), then \( 0 \) is globally asymptotically stable equilibrium.

(b) If \( \lambda = \hat{\lambda} \), the system has two equilibria \( 0 \) and \( \hat{x} \). The set \( \{x \in \mathbb{R}^4: 0 \leq x < \hat{x}\} \) is in the basin of attraction of \( 0 \) while \( \{x \in \mathbb{R}^4: x \geq \hat{x}\} \) is in the basin of attraction of \( \hat{x} \).

(c) If \( 0 < \lambda < \hat{\lambda} \), the system has three equilibria \( 0, x^* \) and \( x^{**} \). The set \( \{x \in \mathbb{R}^4: 0 \leq x < x^*\} \) is in the basin of attraction of \( 0 \), while \( \{x \in \mathbb{R}^4: x > x^*\} \) is in the basin of attraction of \( x^{**} \).

**Proof:** In the notation used in the proof of Theorem 8, the system (28)–(30), (33) can be written as

\[
\frac{dx}{dt} = \Phi(\lambda, x).
\]

Denote by \( x_\lambda(z, t) \) the solution of (36) satisfying \( x_\lambda(z, 0) = z \). Consider the point \( b_m \) as given by (10)–(13). Using (14) we have

\[
\Phi(\lambda, b_m) \leq \Phi(0, b_m) = f(b_m) < 0.
\]

Then it follows from Theorem 6 that (36) defines a dynamical system on \([0, b_m]\).
a) Let $\lambda > \hat{\lambda}$. In this case the system (36) has only one nonnegative equilibrium, namely 0. Therefore, by Theorem 6, 0 is globally asymptotically stable on $[0, b_m]$. Since $b_m$ can be selected to be larger than any point in $\mathbb{R}_4^+$, this implies that 0 is globally asymptotically stable on $\mathbb{R}_4^+$. 

c) In this case $\lambda < \hat{\lambda}$ and the dynamical system (36) has three equilibria 0, $x^*$ and $x^{**}$. For an arbitrary $0 < \delta < \alpha^* - \alpha^{**}$, let $c_\delta = (A_\delta, Y_\delta, M_\delta, F_\delta)'$ where

\[
M_\delta = \frac{\lambda}{\alpha^* - \delta}, \tag{37}
\]
\[
A_\delta = \frac{\mu M}{(1 - r)\gamma} M_\delta, \tag{38}
\]
\[
Y_\delta = \frac{r \gamma}{\beta + \mu Y} A_\delta = \frac{r \gamma M}{(\beta + \mu Y)(1 - r)\gamma} M_\delta, \tag{39}
\]
\[
F_\delta = \frac{\beta M_\delta}{\mu F (M_\delta + \lambda)} Y_\delta = \frac{\beta r \gamma M}{(1 + \alpha^* - \delta)\mu F (\beta + \mu Y)(1 - r)\gamma} M_\delta. \tag{40}
\]

Substituting in the expression for $\Phi$ we have

\[
\Phi(\lambda, c_\delta) = \begin{pmatrix}
\delta (\alpha^* - \alpha^{**} - \delta) \\
(\alpha^* - \delta)(1 + \alpha^* - \delta) \\
0 \\
0 \\
0
\end{pmatrix}. \tag{41}
\]

Let $\delta > 0$. Then $\Phi(\lambda, c_\delta) > 0$ and $c_\delta > x^*$. Applying Theorem 6 with $a = c_\delta$ and $b = b_m$ we obtain that for $m$ sufficiently large equation (36) defines a dynamical system on $[c_\delta, b_m]$ and that $x^{**}$ is globally asymptotically stable on $[c_\delta, b_m]$. Using the fact that

\[M_\delta = M^* + \frac{\lambda \delta}{\alpha^*(\alpha^* - \delta)} \leq \frac{\delta}{Q},\]

it is easy to see that for any point $z > x^*$ one can find $\delta > 0$ so that $c_\delta < z$ (see Appendix for detailed computations). In addition, $b_m$ can be larger than any point in $\mathbb{R}_4^+$. Therefore, $x^{**}$ is a stable equilibrium and $\{x \in \mathbb{R}_4^+ : x \gg x^*\}$ is in its basin of attraction. Using that the system (36) is irreducible one can replace $\gg$ with $>$ by an application of Theorem 4 as in the proof of Theorem 7.

For $\delta < 0$ we have $\Phi(\lambda, c_\delta) < 0$ and $c_\delta < x^*$. Then following the same method we prove that 0 is stable and $\{x \in \mathbb{R}_4^+ : x < x^*\}$ is in its basin of attraction.

The proof of b) is similar to c) with respective modification to take into account that $\alpha = \hat{\alpha} = \alpha^* = \alpha^{**}$. ■

5. Discussion and simulations

It follows from Theorem 9(a) that sufficiently large output of sterile mosquito, namely $M_F(t) = \lambda > \hat{\lambda}$, is an effective control of the mosquito population. The
threshold number \( \hat{\lambda} \) can be written in the form

\[
\hat{\lambda} = \frac{\sqrt{R} - 1}{\sqrt{R} + 1} M^#.
\] (42)

Hence for large values of \( R \), as in the case of *anopheles*, we have \( \hat{\lambda} \approx M^# \). This indicates that for effective control using only the SIT method the output of sterile male mosquito need to be of similar magnitude as the equilibrium value of the male mosquito as determined by the environment. Further, we note that the qualitative behavior of the solutions of the model presented in Theorem 9 is independent of the values of the parameters of the model as long as \( R > 1 \).

Moreover, the threshold number \( \hat{\lambda} \) depends on the parameters only via \( R \) and \( M^# \), where both quantities are directly measurable in a practical setting. In the simulation presented on Figs. 3, 4 and 5, we use the parameter values in Table 1 obtained from [14].

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Description</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \phi )</td>
<td>Number of eggs at each deposit per capita (per day)</td>
<td>50</td>
</tr>
<tr>
<td>( \gamma )</td>
<td>Maturation rate from larvae to adult (per day)</td>
<td>0.1</td>
</tr>
<tr>
<td>( \mu_1 )</td>
<td>mortality rate of the aquatic stage</td>
<td>( \approx 0.25 )</td>
</tr>
<tr>
<td>( \mu_2 )</td>
<td>density mortality rate of the aquatic stage</td>
<td>( 10^{-9} )</td>
</tr>
<tr>
<td>( 1/\mu_T )</td>
<td>Average lifespan of sterile male (in days)</td>
<td>7</td>
</tr>
<tr>
<td>( 1/\mu_M )</td>
<td>Average lifespan of male mosquitoes (in days)</td>
<td>7</td>
</tr>
<tr>
<td>( 1/\mu_F )</td>
<td>Average lifespan of female mosquitoes (in days)</td>
<td>10</td>
</tr>
<tr>
<td>( 1/(\beta + \mu_Y) )</td>
<td>Average time in compartment Y (in days)</td>
<td>3.33</td>
</tr>
</tbody>
</table>

Table 1: Entomological parameters [14]

The dynamics of all compartments are very similar to each other. Hence, only the graphs of the total flying mosquito population, that is, \( Y + M + F \) are presented on these figures. For these values of the parameters we have that \( \hat{\lambda} \) is about 78% of \( M^# \). In absolute terms it may turn out that \( \hat{\lambda} \) is so large that it may not be practical to implement SIT with \( \lambda > \hat{\lambda} \) and/or to sustain it for sufficiently long time. As demonstrated on Fig. 4, for sterile insect release
\( \lambda < \hat{\lambda} \) the equilibrium \( x^{**} \) attracts all trajectories with initial value higher than \( x^* \). Thus, SIT is reducing the mosquito population from the wild equilibrium \( x^\# \) to a new equilibrium \( x^{**} < x^\# \).

6. The yield of SIT

Our aim in this section is to quantify the controlling effect SIT has on the mosquito population. Without human intervention the wild mosquito population settles around its natural equilibrium \( x^\# \), which has coordinates
If a sterile male mosquito is released, this results in a certain decrease of the wild population. To quantify this decrease we need to make the basic assumption that the sterile mosquitoes are released at the same rate until the population settles around a new equilibrium. Then the reduction from $x^\#$ to the new equilibrium describes the effect of SIT. Assume first that $\lambda < \hat{\lambda}$. Then the new equilibrium is $x^{**} = (A^{**}, Y^{**}, M^{**}, F^{**})^t$. Explicitly the equilibrium $x^{**}$ is given by

\begin{align*}
M^{**} &= \frac{\lambda}{\alpha^{**}}, \\
A^{**} &= \frac{\mu_M}{(1 - r)\gamma} M^{**}, \\
Y^{**} &= \frac{r\gamma}{\beta + \mu_Y} A^{**} = \frac{r\mu_M}{(\beta + \mu_Y)(1 - r)} M^{**}, \\
F^{**} &= \frac{\beta M^{**}}{(M^{**} + \lambda)\mu_F} Y^{**} = \frac{\beta r\mu_M}{(1 + \alpha^{**})\mu_F(\beta + \mu_Y)(1 - r)} M^{**},
\end{align*}

where $\alpha^{**} = \frac{1}{2} \left( R - 1 - \lambda Q - \sqrt{(R - 1 - \lambda Q)^2 - 4\lambda Q} \right)$.

It is easy to see from (7) that the coordinates of $x^\#$ can be represented in similar terms. More precisely, we have

\begin{align*}
M^\# &= \frac{R - 1}{Q}, \\
A^\# &= \frac{\mu_M}{(1 - r)\gamma} M^\#, \\
Y^\# &= \frac{r\mu_M}{(\beta + \mu_Y)(1 - r)} M^\#, \\
F^\# &= \frac{\beta r\mu_M}{\mu_F(\beta + \mu_Y)(1 - r)} M^\#.
\end{align*}

Now one can see that the relative reduction of the compartments $A, Y, M$ between the two equilibria is the same, that is

$$\frac{M^\# - M^{**}}{M^\#} = \frac{A^\# - A^{**}}{A^\#} = \frac{Y^\# - Y^{**}}{Y^\#}.$$ 

It is easy to see that $\frac{F^\# - F^{**}}{F^\#} \neq \frac{M^\# - M^{**}}{M^\#}$. However, we have to recall that with the SIT application the wild female mosquito is divided into the compartments $F$ and $U$. It follows from (24) that

$$U^{**} = \frac{\beta \lambda}{(M^{**} + \lambda)\mu_F} Y^{**} = \frac{\alpha^{**}\beta r\mu_M}{(1 + \alpha^{**})\mu_F(\beta + \mu_Y)(1 - r)} M^{**}.$$ 

Thus we obtain

$$\frac{F^\# - F^{**} - U^{**}}{F^\#} = \frac{M^\# - M^{**}}{M^\#}.$$ 

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Then the number
\[
\varphi := \frac{M^# - M^{**}}{M^#} = \frac{A^# - A^{**}}{A^#} = \frac{Y^# - Y^{**}}{Y^#} = \frac{F^# - F^{**} - U^{**}}{F^#}
\]
describes the reduction of the wild mosquito population both in its totality and per compartment. We call the number \( \varphi \) shortly SIT yield. Using (43) and (47) we obtain
\[
M^{**} = \lambda \frac{1}{\alpha^{**}} = \lambda \frac{R - 1 - \lambda Q + \sqrt{(R - 1 - \lambda Q)^2 - 4\lambda Q}}{2\lambda Q}
\]
\[
= \frac{1}{2} \left( M^# - \xi + \sqrt{(1 - \xi)^2 - \frac{4\xi}{R - 1} M^#} \right)
\]
where \( \xi = \frac{\lambda}{M^#} \) is the release of sterile males as a fraction of the wild male natural equilibrium. Then we derive an explicit expression for \( \varphi \) as follows
\[
\varphi = \frac{M^# - M^{**}}{M^#} = \frac{1}{2} \left( 1 + \xi - \sqrt{(1 - \xi)^2 - \frac{4\xi}{R - 1}} \right).
\]
This is valid when \( \lambda \leq \hat{\lambda} \) or equivalently
\[
\xi \leq \hat{\xi} := \frac{\hat{\lambda}}{M^#} = \frac{\sqrt{R - 1}}{\sqrt{R + 1}}.
\]
For \( \xi > \hat{\xi} \) the mosquito population decreases to 0. So its relative reduction from \( x^# \) to the new equilibrium \( 0 \) is 1 so that the SIT yield is
\[
\varphi(\xi) = \begin{cases} 
1 + \xi - \sqrt{(1 - \xi)^2 - \frac{4\xi}{R - 1}} & \text{if } \xi \leq \hat{\xi}, \\
1 & \text{if } \xi > \hat{\xi}.
\end{cases}
\]
One should note that \( \varphi \) as a function of \( \xi \) depends on the parameters of the model only through the basic offspring number \( R \). It particular, it is independent of the density dependent death rate \( \mu_2 \) of the aquatic stage. Hence qualitatively the response of the mosquito population to SIT is independent of its size. The yield function for various values of \( R \) is graphically presented on Fig. 6.
Sterile male release such that \( \lambda \leq \hat{\lambda} \), or equivalently \( \xi \leq \hat{\xi} \), results in suppressing the mosquito population towards extinction only if at the beginning of the release process this population is relatively small, namely, less than \( y^* \).

The equilibrium \( y^* \) is obtained similarly to (43)–(46) with \( \alpha^* \) instead of \( \alpha^{**} \). Then a sterile male mosquito release of \( \lambda = \xi M \# \) can suppress a population not exceeding \( y = \eta y \# \) where

\[
\eta := \frac{M^*}{M \#} = \frac{A^*}{A \#} = \frac{Y^*}{Y \#} = \frac{F^* + U^*}{F \#}.
\]

Further, \( \eta \) can be obtained as a function of \( \xi \) in the form

\[
\eta = \frac{1}{2} \left( 1 - \xi - \sqrt{(1 - \xi)^2 - \frac{4\xi}{R - 1}} \right).
\]

The graph of the function \( \eta = \eta(\xi) \) is presented on Fig. 7 for several values of \( R \). One can observe that for the mosquito population for large \( R \), as it is typically the case for mosquito, the population size needs to be relatively quite small to be controlled via release \( \xi < \hat{\xi} \). For example if \( R = 40 \), release of \( \xi = 60\% \) can only control a mosquito population of about 4\% of the wild equilibrium.
Remark 10. In [16, 17] the authors have considered mechanical control as an additional vector control tool for the Chikungunya Disease. Mechanical control consists in reducing the breeding sites, that is in reducing eggs deposit. In our model, mechanical control would modify the parameter $\phi$. Indeed, let $c \in [0, 1]$ be the parameter that represents the mechanical control: when $c = 1$, there is no mechanical control, and when $c = 0$, the mechanical control is total (elimination of all breeding sites). Then the time evolution of the aquatic stage becomes

$$\frac{dA}{dt} = c\phi F - (\gamma + \mu_1 + \mu_2 A),$$

such that the new basic offspring number is given by $R_m = cR$. Thus following Figs. 6 and 7, it is also clear, that a combination of SIT and Mechanical Control can be helpful to improve the yield of the Control.

7. Conclusion

By establishing a relevant and simple mathematical model, we studied, through it, the life cycle of mosquitos (from the aquatic stage to the adult stage that includes the female vectors). We establish, in terms of the basic offspring number, properties of global nature such as the stability of the equilibria and their basins of attraction. We added to this initial model a compartment of treated or sterile male mosquitos in order to get the SIT model. We obtained a threshold number and established theoretically and computationally that the control of wild female mosquitos is effective provided that the number of released sterile male mosquitos is above this threshold number. Furthermore, below this number, we determined properties of global nature for the equilibria of the system.

Our study reveals that the success of the SIT depends on the entomological parameters of the wild *anopheles* mosquito as well as on parameters related...
to the sterile males, i.e. $p$, $q$, $\psi$, and $\mu_T$, which determine the said threshold number. This leads us to being interested in future work in combining the SIT with Mechanical Control, in order to improve the control. Comparing the SIT approach with standard chemical vector control is also of interest. Another possible extension would be to couple our SIT model with the epidemiological model studied in [3].

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[26] E F Knipling, Possibilities of insect control or eradication through the use of sexually sterile males, J. Econ. Entomol. 48 (1955) 495.


Appendix: Computations

A. Equilibria of the model (28)–(30), (33)

When equating the right hand side of the considered system of ODEs, from equations (33) and (29) we have

\[ F = \frac{\beta M}{(M + \lambda)\mu_F} Y = \frac{\beta M}{(M + \lambda)\mu_F} \frac{r\gamma}{\beta + \mu_Y} A \]

Substituting into the right hand side of (28) we obtain

\[ \phi \frac{\beta M}{(M + \lambda)\mu_F} \frac{r\gamma}{\beta + \mu_Y} A - (\gamma + \mu_1 + \mu_2 A) A = 0 \]
Excluding the trivial case \( A = 0 \) the equation simplifies to

\[
\frac{\phi \beta r \gamma M}{\mu_F(\beta + \mu_Y)(M + \lambda)} - \gamma - \mu_1 - \mu_2 A = 0
\]

Using (30), we have \( A = \frac{\mu_M M}{(1 - r)\gamma} \), which implies the following equation about \( M \)

\[
\frac{\phi \beta r \gamma M}{\mu_F(\beta + \mu_Y)(M + \lambda)} - \gamma - \mu_1 - \mu_2 \frac{\mu_M M}{(1 - r)\gamma} = 0.
\]

Substitution \( \alpha = \frac{\lambda}{M} \) leads to

\[
\frac{\phi \beta r \gamma}{\mu_F(\beta + \mu_Y)(\gamma + \mu_1)} \frac{1}{1 + \alpha} - \gamma - \mu_1 - \frac{\mu_2 \mu_M}{(1 - r)\gamma(\gamma + \mu_1)} \frac{\lambda}{\alpha} = 0
\]

or equivalently

\[
\frac{\phi \beta r \gamma}{\mu_F(\beta + \mu_Y)(\gamma + \mu_1)} \frac{1}{1 + \alpha} - 1 - \frac{\mu_2 \mu_M}{(1 - r)(\gamma + \mu_1)} \frac{\lambda}{\alpha} = 0. \tag{51}
\]

The coefficient of \( \frac{1}{1 + \alpha} \) is exactly the basic offspring number \( R \). Denote the coefficient of \( \frac{\lambda}{\alpha} \) by \( Q \). Then (51) is equivalent to the following quadratic equation with respect to \( \alpha \)

\[
\alpha^2 - (R - 1 - \lambda Q)\alpha + \lambda Q = 0 \tag{52}
\]

As mentioned earlier it is assumed that \( R > 1 \). The discriminant is

\[
\Delta = (R - 1 - \lambda Q)^2 - 4\lambda Q = (R + 1 - \lambda Q)^2 - 4R = ((\sqrt{R} - 1)^2 - \lambda Q)(\sqrt{R} + 1)^2 - \lambda Q).
\]

The equation two real positive roots iff \( \Delta > 0 \) and \( R - 1 - \lambda Q > 0 \). Then using that \((\sqrt{R} - 1)^2 < R - 1 < (\sqrt{R} + 1)^2 \) and in view of the expression for \( \Delta \), equation (52) has two positive roots iff \((\sqrt{R} - 1)^2 - \lambda Q > 0 \) or equivalently

\[
\lambda < \hat{\lambda} := \frac{(\sqrt{R} - 1)^2}{Q}.
\]

In this case the roots are

\[
\alpha^* = \frac{1}{2} \left( R - 1 - \lambda Q + \sqrt{(R - 1 - \lambda Q)^2 - 4\lambda Q} \right)
\]

\[
\alpha^{**} = \frac{1}{2} \left( R - 1 - \lambda Q - \sqrt{(R - 1 - \lambda Q)^2 - 4\lambda Q} \right)
\]

Further, if \( \lambda = \hat{\lambda} \) equation (52) has one positive root \( \hat{\alpha} = \sqrt{R} - 1 \) and it has no positive roots when \( \lambda > \hat{\lambda} \).
Equilibria:

\[
M^* = \frac{\lambda}{\alpha^*} \\
A^* = \frac{\mu_M}{(1-r)\gamma} M^* \\
Y^* = \frac{r\gamma}{\beta + \mu_Y} A^* = \frac{r\gamma \mu_M}{(\beta + \mu_Y)(1-r)\gamma} M^* \\
F^* = \frac{\beta M^*}{(M^* + \lambda)\mu_F} Y^* = \frac{\beta}{(1+\alpha^*)\mu_F} \frac{r\gamma}{\beta + \mu_Y} A^* \\
= \frac{\beta r\gamma \mu_M}{(1+\alpha^*)\mu_F(\beta + \mu_Y)(1-r)\gamma} M^*
\]

For \(x^{**}\) and \(\hat{x}\) one uses the above formulas with \(\alpha^{**}\) and \(\hat{\lambda}\).

**B. Inequalities in the proof of Theorem 9**

The first coordinate of the function \(\Phi\) is

\[
\Phi_1(\lambda, x) = \frac{\phi \beta r\gamma}{\mu_F(\beta + \mu_Y)} \frac{1}{1+\alpha} - \gamma - \mu_1 - \frac{\mu_2 \mu_M \lambda}{(1-r)\gamma} \\
= - \frac{\alpha^2 - (R - 1 - \lambda Q)\alpha + \lambda Q}{\alpha(\alpha + 1)} = \frac{(\alpha^* - \alpha)(\alpha - \alpha^{**})}{\alpha(\alpha + 1)}.
\]

Using that \(0 < \delta < \alpha^* - \alpha^{**}\) we obtain

\[
\Phi_1(\lambda, c_\delta) = \frac{\delta (\alpha^* - \delta - \alpha^{**})}{(\alpha^* - \delta)(\alpha^* - \delta + 1)} > 0
\]

The inequality \(c_\delta > x^*\) is derived as follows:

\[
M_\delta = \frac{\lambda}{\alpha^* - \delta} > \frac{\lambda}{\alpha^*} = M^* \\
A_\delta = \frac{\mu_M}{(1-r)\gamma} M_\delta > \frac{\mu_M}{(1-r)\gamma} M^* = A^* \\
Y_\delta = \frac{r\gamma}{\beta + \mu_Y} A_\delta > \frac{r\gamma}{\beta + \mu_Y} A^* = Y^* \\
F_\delta = \frac{\beta M_\delta}{(M_\delta + \lambda)\mu_F} Y_\delta = \frac{\beta}{(1+\alpha^* - \delta)\mu_F} Y_\delta \\
> \frac{\beta}{(1+\alpha^*)\mu_F} Y_\delta > \frac{\beta}{(1+\alpha^*)\mu_F} Y^* = F^*
\]
Further, we also have

\[
M_\delta - M^* = \frac{\lambda}{\alpha^* - \delta} - \frac{\lambda}{\alpha^*} = \frac{\lambda \delta}{\alpha^*(\alpha^* - \delta)} \leq \frac{\lambda \delta}{\alpha^* \alpha^{**}} = \frac{\delta}{Q}
\]

\[
A_\delta - A^* = \frac{\mu_M}{(1 - r)\gamma}(M_\delta - M^*)
\]

\[
Y_\delta - Y^* = \frac{r\gamma}{\beta + \mu_Y}(A_\delta - A^*)
\]

\[
F_\delta - F^* = \frac{\beta}{(1 + \alpha^* - \delta)\mu_F} Y_\delta - \frac{\beta}{(1 + \alpha^*)\mu_F} Y^*
\]

\[
= \left( \frac{\beta}{(1 + \alpha^* - \delta)\mu_F} - \frac{\beta}{(1 + \alpha^*)\mu_F} \right) Y_\delta + \frac{\beta}{(1 + \alpha^*)\mu_F} (Y_\delta - Y^*)
\]

\[
= \frac{\delta \beta}{(1 + \alpha^{**})(1 + \alpha^*)\mu_F} Y_\delta + \frac{\beta}{(1 + \alpha^*)\mu_F} (Y_\delta - Y^*).
\]

Therefore \(c_\delta\) is arbitrary close to \(x^*\).