A cascade of destabilizations: Combining Wolbachia and Allee effects to eradicate insect pests

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Abstract

1. The management of insect pests has long been dominated by the use of chemical insecticides, with the aim of instantaneously killing enough individuals to limit their damage. To minimize unwanted consequences, environmentally friendly approaches have been proposed that utilize biological control and take advantage of intrinsic demographic processes to reduce pest populations.

2. We address the feasibility of a novel pest management strategy based on the release of insects infected with Wolbachia, which causes cytoplasmic incompatibilities in its host population, into a population with a pre-existing Allee effect. We hypothesize that the transient decline in population size caused by a successful invasion of Wolbachia can bring the population below its Allee threshold and, consequently, trigger extinction.

3. We develop a stochastic population model that accounts for Wolbachia-induced cytoplasmic incompatibilities in addition to an Allee effect arising from mating failures at low population densities. Using our model, we identify conditions under which cytoplasmic incompatibilities and Allee effects successfully interact to drive insect pest populations towards extinction. Based on our results, we delineate control strategies based on introductions of Wolbachia-infected insects.

4. We extend this analysis to evaluate control strategies that implement successive introductions of two incompatible Wolbachia strains. Additionally, we consider methods that combine Wolbachia invasion with mating disruption tactics to enhance the pre-existing Allee effect.

5. We demonstrate that Wolbachia-induced cytoplasmic incompatibility and the Allee effect act independently from one another: the Allee effect does not modify the Wolbachia invasion threshold, and cytoplasmic incompatibilities only have a marginal effect on the Allee threshold. However, the interaction of these two processes can drive even large populations to extinction. The success of this method can be amplified by the introduction of multiple Wolbachia cytotypes as well as the addition of mating disruption.

6. Our study extends the existing literature by proposing the use of Wolbachia introductions to capitalize on pre-existing Allee effects and consequently eradicate insect pests. More generally, it highlights the importance of transient dynamics, and the relevance of manipulating a cascade of destabilizations for pest management.
1 | INTRODUCTION

Although most insect species provide crucial ecosystem services (Losey & Vaughan, 2006), a minority of taxa that we consider pests (~1%) have an overwhelming influence on the development of population management in theory and in practice. Among the various environmentally friendly approaches that have been envisaged to control unwanted species, we focus on a research avenue that proposes the exploitation of Allee effects, i.e. the decrease in survival or reproduction at small population sizes and the consequent reduction in population growth (Berec, Angulo, & Courchamp, 2007; Courchamp, Clutton-Brock, & Grenfell, 1999; Dennis, 1989; Liebhold & Tobin, 2008; Stephens, Sutherland, & Freckleton, 1999). The central ideas surrounding this work are twofold: management tactics could be combined in order to (1) reduce a population size down below the Allee threshold—the population size at which the per capita growth rate decreases (in the case of “weak” Allee effects) or becomes negative (in the case of “strong” Allee effects)—which, in turn, increases the probability of stochastic extinction, and/or (2) amplify the mechanisms underpinning a pre-existing Allee effect to increase the Allee threshold itself (Tobin, Berec, & Liebhold, 2011). Capitalizing on Allee effects to manage undesirable species is particularly advantageous because it drives populations into extinction vortices without needing to eliminate every last individual.

Control methods centred on manipulating mating success as an alternative to chemical pesticides have long been recognized as desirable (e.g. Baumhover et al., 1955; Knipling, 1955). The application of such control methods has been considered for managing many insect pests including the Oriental fruit fly (Steiner et al., 1970), Indian meal moth and almond moths (Sower & Whitmer, 1977), and Gypsy moths (Beroza & Knipling, 1972; Knipling, 1970). Importantly, mating disruption has been successfully used to control populations with pre-existing Allee effects. The Gypsy moth (Lymantia dispar), for example, is one of the few insect species for which both a component (mate-finding) and demographic Allee effect have been explicitly identified (Johnson, Liebhold, Tobin, & Bjornstad, 2006; Tobin, Onufrieva, & Thorpe, 2013; Tobin, Whitmire, Johnson, Bjornstad, & Liebhold, 2007). The Gypsy moth is an invasive forest pest in North America and triggered a major containment programme to slow the spread towards the western United States (Liebhold, Halverson, & Elmes, 1992; Sharov, Leonard, Liebhold, Roberts, & Dickerson, 2002). Mating disruption has been a major tactic used to control newly established low-density populations along the invasion front, with evidence supporting that it is more efficient as well as economically cheaper than classic treatments with the pesticide Bacillus thuringiensis (Sharov, Leonard, Liebhold, & Clemens, 2002; Sharov, Leonard, Liebhold, Roberts, et al., 2002). This highlights the potential benefits of identifying other pest species that have pre-existing Allee effects and determining whether environmentally desirable forms of control may similarly be effective.

Several recent theoretical developments have focused on taking advantage of Allee effects to promote pest eradication (e.g. Blackwood et al., 2012; Boukal & Berec, 2009; Liebhold & Bascompte, 2003; Yamanaka & Liebhold, 2009). These models capture the underlying population dynamics of a pest and evaluate the success of population management tactics such as culling, release of sterile males and mating disruption to determine whether these methods can create or enhance pre-existing Allee effects (Fauvergue, 2013 provides a comprehensive review). While there is evidence that such population management strategies will be successful for populations with pre-existing Allee effects, the range of species that might benefit from these tactics may be much greater than currently known. In a meta-analysis focused on the presence of Allee effects in natural animal populations (Kramer, Dennis, Liebhold, & Drake, 2009), terrestrial arthropods were found associated with the largest number of studies (22) and the highest proportion (77%) exhibiting an Allee effect. Mating failure at low density appeared as the most frequent mechanism. Additionally, Fauvergue (2013) found evidence supporting the presence of mate-finding Allee effects in 19 out of 34 published studies that investigated the interplay of population size and mating success in insects. Indirectly, the central role of Allee effects in insect population dynamics is supported by the efficiency of eradication programmes based on the disruption of reproduction. Pest management based on the reduction of mating success via mass trapping, mating disruption with sex pheromones or the release of sterile males has indeed proved successful in several instances (Baumhover et al., 1955; Knipling, 1955; Krafsur, 1998; Sukling, Tobin, McCullough, & Herms, 2012; Sukling et al., 2014).

In this article, we investigate Wolbachia-induced cytoplasmic incompatibility (CI) as a novel method for triggering reproductive failures and consequently bringing a pest population below its Allee threshold. Wolbachia are endosymbiotic bacteria that infect at least 20% of all insect species and up to two-thirds in some estimations (Hilgenboecker, Hammerstein, Schlattmann, Telschow, & Werren, 2008). Wolbachia have various effects on their insect hosts, the most widespread and prominent being CI (Stouthamer, Breeuwer, & Hurst, 1999). Under CI, matings between an infected male and a female that is either uninfected or infected with an incompatible cytotype result in offspring mortality during embryonic development. Fitness advantages of infected females as well as maternal inheritance are key features that promote invasion of Wolbachia into a host population: above a threshold frequency, a given Wolbachia strain is expected to invade until near-fixation (Barton & Turelli, 2011; Caspari & Watson, 1959; Hancock, Sinkins, & Godfray, 2011; Hoffmann & Turelli, 1997; Turelli & Hoffmann, 1991). As a result of the associated CI and subsequent reduction in reproductive rate, Wolbachia invasion via the release of...
infected hosts is a candidate biological control agent against arthropod pests (Bourtzis, 2008).

In practice, there are multiple ways to implement a management strategy centred on inducing CIs via introduction of Wolbachia. For example, similar to the use of “Sterile Insect Technique” (SIT), males bearing a Wolbachia strain incompatible with that of the target population can be released in large numbers. CIs arising from the mating of females and infected males would substantially limit the total offspring in the subsequent generation, resulting in a decrease in overall population growth rate and thereby increasing the possibility of local population extinction (Atyame et al., 2015; Laven, 1967; Zabalou et al., 2004). Incompatible males can be obtained via transfection, even between completely different species of host insects (e.g. Braig, Guzman, Tesh, & Oneill, 1994; Hoffmann et al., 2011). At the population level, the conceptual underpinnings for mass releases of incompatible males do not depart from that of SIT, for which interactions with the Allee effect have already been thoroughly analysed (Barclay, 1982; Barclay & Mackauer, 1980; Berec, Maxim, & Bernhauerová, 2016; Boukal & Berec, 2009; Fauvergue, 2013; Lewis & Van Den Driessche, 1993; Yamanaka & Liebold, 2009).

An alternative management tactic using CI relies on the inoculation of a relatively small number of insects of both sexes with a Wolbachia strain incompatible with that of the target population. This method is investigated in the model introduced in Dobson, Fox, and Jiggins (2002), which combines insect population dynamics with releases of individuals infected with Wolbachia. During a successful invasion of Wolbachia, a transient reduction in the insect population size occurs. This decline results from the temporary increase in the fraction of incompatible matings, which peaks in the midst of the invasion process. Therefore, systematic introductions of different Wolbachia cytotypes could be applied to artificially sustain an unstable coexistence of multiple incompatible infections within an insect population, allowing the population size to be reduced and maintained at low levels (Dobson et al., 2002).

Our goal was to determine when the latter implementation of Wolbachia introductions can drive a population to extinction in the presence of Allee effects. Specifically, we derive a mathematical model built upon Dobson et al.’s (2002) approach of CI management that additionally accounts for Allee effects as well as environmental and demographic stochasticity. We also consider mating disruption in our model as a potential complementary tactic. We use this model to address three primary questions: (1) What is the influence of Allee effects present within a host population on Wolbachia invasion dynamics? (2) What is the influence of CI on the demographic Allee effect? (3) What is the influence of a combination of Wolbachia-induced CI, Allee effects, mating disruption and stochasticity on the probability of host extinction?

2 | MATERIALS AND METHODS

2.1 | Population model

Our model extends the framework introduced by Dobson et al. (2002) by accounting for pre-existing Allee effects, the release of pheromone sources as a method of mating disruption, as well as both demographic and environmental stochasticity. In this section, we first introduce a model that considers the population dynamics in the absence of individuals infected with Wolbachia.

We considered populations such that the dynamics can be modelled in discrete time with non-overlapping generations. The population model explicitly tracks the total population size at each time t, given by \( N_t \). Our population model can be expressed in terms of either census size or density and hereafter, we refer to our model in terms of size. Therefore, while the deterministic model can take non-integer values, population sizes less than one are considered extinct. In contrast, the stochastic model forces integer population sizes. We assume that each time step can be broken into two stages: the first (at time \( t + 0.5 \)) captures reproduction, and the second (at time \( t + 1 \)) captures density-dependent survivorship of offspring to adults. The total number of offspring is given by:

\[
N_{t+0.5} = \bar{n} F_t g_1(M_t) g_2(F_t)
\]

where \( \bar{n} \) is the maximum per capita female fecundity, \( F_t \) is the number of females in the population and \( M_t \) is the number of males. \( g_1(M_t) \) captures a component Allee effect that results from the failure of mates finding one another at low densities such that:

\[
g_1(M_t) = \frac{M_t}{M_t + \theta}
\]

Here, \( \theta \) is the strength of the Allee effect; a convenient interpretation of this term is that \( \theta \) represents the number of males at which half of the females successfully find a mate or, equivalently, the maximum mating rate is reduced by half (Boukal & Berec, 2009).

The function \( g_2(F_t) \) in Equation 1 captures the decline in fecundity resulting from techniques to control populations via mating disruption. We assume that pheromone sources are introduced into the environment and maintained in the population at a fixed number \( P \) and only a fraction \( F_t/(F_t + P) \) males successfully find a mate (Fauvergue, 2013), or:

\[
g_2(F_t) = \frac{F_t}{F_t + P}
\]

We now assume that there is a 50:50 sex ratio (i.e. \( F_t = M_t = N_t/2 \)) and define constants \( m = \bar{n}/2 \) (which is now the overall per capita fecundity), \( \theta = 2\theta \) and \( P = 2P \). Now, we can write both \( g_1 \) and \( g_2 \) as functions of \( N_t \) and Equation 1 is now:

\[
N_{t+0.5} = mN_t \frac{N_t}{N_t + \theta} \frac{N_t}{N_t + P}
\]

For a simpler biological interpretation, we consider \( \theta \) relative to the carrying capacity of the population in the absence of any control (denoted as \( K \)) as an indicator of the intensity of the Allee effect (i.e. \( \theta/K \)). Now, when \( \theta = K \) (or, equivalently, \( \theta/K = 1 \)), half of the total population successfully mates at its carrying capacity. Hereafter, we refer to \( \theta/K \) as the “relative strength of the Allee effect” and restrict its values to the range \((0–0.5)\), with 0.5 likely on the upper end of biologically reasonable values for \( \theta/K \). Similarly, we consider \( P \), the number of pheromone sources, relative to \( K \).
Finally, we assume that survivorship of offspring to adults is density-dependent so that:

$$N_{t+1} = N_{t+0.5} S_N$$

where

$$S_N = \frac{S_0}{1 + (aN_{t+0.5})^\gamma}$$

(5)

and the constant $a$ is related to the carrying capacity (more details provided in the Supporting Information S1). $\gamma$ is related to intraspecific competition and $S_N$ is survivorship in the absence of intraspecific competition (Slatkin & Smith, 1979). Combining Equations (4)–(7), we are left with the model:

$$N_{t+1} = mN_t \left( \frac{N_t}{N_t + \theta} \right) \left( \frac{N_t}{N_t + P} \right) \frac{S_0}{1 + (aN_{t+0.5})^\gamma}$$

(6)

Our goal was to consider the dynamical consequences of population control methods (i.e. Wolbachia introductions and the release of pheromone sources). Therefore, we analytically determine the equilibrium values for the population model (7) in the absence of these management tactics ($P = 0$). Throughout the remainder of the paper, we distinguish between the “maximum reproductive rate” (which is given by $mS_0$ and therefore accounts for both fecundity and density-independent survivorship) and the overall “reproductive rate” (which is the reproductive rate in the presence of all other demographic processes in our model, namely the Allee effect). As detailed in the Supporting Information S1), we find three equilibria (assuming the maximum reproductive rate is greater than 1 and setting $\gamma = 1$). The first equilibrium corresponds to population extinction, and the other two are given by:

$$mS_0 - 1 \pm \sqrt{(mS_0 - 1)^2 - 4am\theta}$$

$$2am$$

(8)

When Allee effects are not present ($\theta = 0$), these equilibria collapse to a single equilibrium (the carrying capacity $K$). In contrast, for sufficiently large Allee effects, these equilibria no longer exist and the reproductive rate is always less than 1; consequently, the population will be driven to extinction independent of its initial size. However, when the value of $\theta$ is between these extremes, the smaller equilibrium corresponds to the Allee threshold and the larger equilibrium corresponds to the carrying capacity. The structure of these equilibria is integral to the insect species that we are considering; there is a carrying capacity and a strong Allee effect for positive values of $\theta$ up to a threshold. We therefore conjecture that an alternative form of density dependence that captures these properties will yield qualitatively similar results.

Based on this deterministic framework, we build in environmental and demographic stochasticity. We assume that environmental stochasticity leads to variation in the population’s fecundity between generations (Melbourne & Hastings, 2008). Therefore, we adapt the methods of Schoener, Clobert, Legendre, and Spiller (2003) and account for environmental stochasticity by rewriting Equation 4 as:

$$N_{t+0.5} = n_t g_1(N_t) g_2(N_t)$$

(9)

where the fecundity $m_t$ is drawn at each generation from a normal distribution with mean $m$ (that is truncated so that $m \geq 0$) and variance ($\sigma^2$) that scales with the mean. In the main text, we fix the variance so that it is equal to the mean; however, a sensitivity analysis of the magnitude of the variance is provided in the Supporting Information S3). Furthermore, we note that gamma-distributed environmental stochasticity is a commonly used alternative choice; under our parameterization, the probability density function (pdf) is nearly identical for both normal- and gamma-distributed stochasticity and would therefore yield similar results.

Demographic stochasticity results from variation in fecundity at the individual level (Melbourne & Hastings, 2008). In the absence of demographic stochasticity, the total number of individuals that successfully reproduce is given by:

$$B = N_t g_1(N_t) g_2(N_t)$$

(10)

We assume that each of these individuals at a given time $t$ reproduces with fecundity $m_t$ (as described above), and the total number of eggs produced is a Poisson random variable (Melbourne & Hastings, 2008). As the sum of independent Poisson random variables is also a Poisson random variable, the total offspring of all adults is:

$$N_{t+0.5} \sim \text{Poisson}(m_t B)$$

(11)

Finally, we include stochasticity in density-dependent survivorship, again following Melbourne and Hastings (2008). Given that $S_N$ (as defined in Equation 7) is the probability that offspring survive to adults, we assume that survivorship is binomially distributed so that:

$$N_{t+1} \sim \text{Binomial}(N_{t+0.5}, S_N)$$

(12)

### 2.2 Infection dynamics

We consider the infection dynamics of up to two different cytotypes of Wolbachia (referred to as cytotypes X and Y) and denote the number of uninfected individuals as $W$. Note that all variables and parameters with subscripts $X$ (or $Y$) are related to cytotype $X$ (or $Y$). This model is adapted from Dobson et al. (2002); therefore, we use similar notation throughout.

In the presence of a single cytotype of Wolbachia, there are only unidirectional cytoplasmic incompatibilities (CI); in contrast, in the presence of multiple cytotypes, there may be bidirectional CI. We first introduce the case of a single cytotype and then extend the model to include two cytotypes. Below we describe the mathematical formulation of the infection dynamics, and Table 1 summarizes these processes.

#### 2.2.1 One cytotype

To capture the Wolbachia dynamics, we first determine the proportions of infected and uninfected individuals in the population at time $t$. For example, if there are $W_t$ uninfected individuals and $X_t$ infected with cytotype $X$, then under the assumption that there is a 50:50 sex ratio, the fraction of females infected with cytotype $X$ at $t + 0.5$ is given by:

$$\frac{X_t}{W_t}$$
TABLE 1 Summary of Wolbachia transmission and its effects on reproduction in its host population. The first column states the maternal Wolbachia cytotype and the first row states the paternal Wolbachia cytotype. Each box in the table corresponding to a particular pairing between a female and male provides that proportion of the offspring from that pair that are uninfected (WT), infected with cytotype X and infected with cytotype Y.

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<tr>
<th>( \delta )</th>
<th>W</th>
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<tr>
<td>W</td>
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<td>( \delta )</td>
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<td>X</td>
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<td>( Y: (1 - \mu_x) F_x b_i j_i )</td>
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\[ q_i = \frac{X_i}{W_i + X_i} \]

Similarly, we find the fraction \( i_j \) of all males that are infected (where \( i_j = a_j \)), the fraction \( q_j \) of all males that are uninfected, and the fraction \( c_j \) of all females that are uninfected (again \( q_j = c_j \)).

Based on the proportions of uninfected and infected individuals in the population, we can now determine the fraction of offspring that are infected. Crosses between pairs with an infected female suffer a fecundity loss due to infection (1 - \( F_y \)), where \( F_y \) is the relative fecundity of infected individuals. Vertical transmission of Wolbachia occurs maternally and we assume that transmission is successful with probability \( (1 - \mu_y) \), where \( \mu_y \) is the probability of transmission failure. In the instance of Wolbachia-induced CIs, crosses between infected females and uninfected males in addition to crosses between infected males and infected females give rise to infected offspring. The proportion of viable offspring that are infected with cytotype X after reproduction (i.e. at time \( t + 0.5 \)) is therefore given by:

\[ x_{t+0.5} = a_i (1 - \mu_x) F_x (i_j + q_j) \]  (13)

where a lowercase \( x \) is used to denote proportion rather than number. Second, we can identify the proportion of viable offspring that are uninfected (\( w_{t+0.5} \)). Uninfected individuals can arise from crosses between both uninfected females and uninfected males. Furthermore, matings between both infected females and infected males can yield viable uninfected offspring. This results from failure to vertically transmit Wolbachia to their offspring. When one type of Wolbachia is present within a population, then only unidirectional CI is possible. This type of CI occurs through matings between infected males and uninfected females. Therefore, we assume that all but a fraction \( H_x \) of pairings between infected males and uninfected females undergo CI. Additionally, offspring from pairings between infected males and infected females that fail to transmit Wolbachia are subject to CI-induced mortality. The proportion of viable offspring that are not infected with Wolbachia following reproduction is therefore given by:

\[ w_{t+0.5} = (\mu_x F_x a_i + c_i) (i_j H_x + q_i) \]  (14)

Notice that due to cytoplasmic incompatibilities and the fecundity cost due to infection with Wolbachia, the fraction of the offspring that are viable \( (x_{t+0.5} + w_{t+0.5}) \) is less than 1. Therefore, the total number of offspring as governed by Equation 4 can be rewritten as:

\[ N_{t+0.5} = m (x_{t+0.5} + w_{t+0.5}) N_i g_i (N_i) g_2 (N_i) \]  (15)

In other words, the product \( g_i (N_i) g_2 (N_i) \) captures the total fraction of adults at time \( t \) who successfully find a mate, and the sum \( x_{t+0.5} + w_{t+0.5} \) is the fraction of all offspring that are viable. Finally, as described in the previous section, density-dependent mortality limits the total number of adults at time \( t + 1 \) (Equation 7).

### 2.2.2 | Two cytotypes

In addition to releasing a single cytotype of Wolbachia, we consider a scenario in which a second cytotype is introduced. When two cytotypes of Wolbachia are present within a population, bidirectional CI occurs when a male with one cytotype mates with a female infected with an incompatible Wolbachia cytotype. Similar to the previous section, we assume that a fraction \( H_y \) (or \( H_x \), depending on the infection type of the male and female) of offspring survives. Offspring pairings between infected males and females of either cytotype that fail to transmit are again subject to CI-induced mortality.

Therefore, in the presence of two strains, we rewrite Equation 13 as:

\[ x_{t+0.5} = a_i (1 - \mu_x) F_x (i_j + j_i H_y + q_i) \]  (16)

where \( j_i \) is the fraction of males infected with cytotype Y. Similarly, the proportion of viable offspring infected with cytotype Y following reproduction is given by:

\[ y_{t+0.5} = b_i (1 - \mu_x) F_y (i_j H_x + j_i + q_i) \]  (17)

where \( b_i \) is the fraction of females infected with cytotype Y. The proportion of viable uninfected offspring is now given by:

\[ w_{t+0.5} = (\mu_y F_y a_i + \mu_x F_y b_i + c_i) (i_j H_x + j_i H_y + q_i) \]  (18)

### 2.3 | Parameterization

Here, we define the maximum reproductive rate as the product of the per capita fecundity and density-independent survivorship (i.e. \( m_S \)). In other words, this is the reproductive rate in the absence of Allee effects and in the limit as the population size approaches zero. Our parameterization of the population model is based on both the parameterization used in Dobson et al. (2002) and common ranges for insect populations (see Table 2). For example, Hassell, Lawton, and May (1976) cites finite net rates of increase (defined as fecundity discounted by density-independent mortality and therefore equivalent to \( m_S \) here) ranging from 1.3 to 75 (with 22 of the 24 species investigated in the range 1.3–13.5) which is consistent with our parameterization (e.g. see Figure 2 caption and the associated results as...
Additional parameters vary for our analysis including the strength of the Allee effect, \( \theta \) and the initial population size. However, our results are intended to assess the general qualitative behaviour of Wolbachia introductions and consequently the actual implementation of such management tactics would require a detailed analysis and parameterization specific to the target population and cytotype.

Our parameterization for the infection dynamics is based on values that are reasonable for Wolbachia cytotypes (Charlat, Calmet, Andrieu, & Mercot, 2005; Dobson et al., 2002; Hoffmann & Turelli, 1997). In the main text, we assume that fecundity loss, transmission failure and survival of CI are equal between all cytotypes. However, the Supporting Information (S3) provides an analysis of the dynamics when the introduced cytotypes are not identical. S3 additionally provides sensitivity analyses for parameters related to the population model, including the strength of environmental stochasticity and the maximum reproductive rate.

3 | RESULTS

In the following sections, we first test our model against well-established results related to Wolbachia invasion as a method of model validation, establish the relationship between Wolbachia and the strength of the Allee effect, and finally evaluate the potential for the release of infected insects to control a population.

3.1 | Model validation

We first determine whether our model captures the same features of the important earlier work (Hancock et al., 2011; Hoffmann, Turelli, & Harshman, 1990; Turelli & Hoffmann, 1991). Hoffmann et al. (1990) derived an analytic expression for the expected equilibrium infection frequencies. After adjusting their notation to match ours and simplifying, the equilibrium infection frequency (which we denote as \( p \)) for a single cytotype of Wolbachia \( X \) should satisfy the equation:

\[
(1-H_x) (1-\mu_F X) \rho^2 + (F_X + H_x - 2)p + 1 - F_X (1-\mu_x) = 0
\]

Their work predicts that there is an unstable equilibrium referred to as the Wolbachia invasion threshold; cytotypes with initial infection frequencies above this threshold will increase until reaching the higher stable equilibrium, indicating a successful invasion. Following Charlat et al. (2005), we considered invasion dynamics by estimating the infection frequency at generation \( t + 1 \) as a function of the frequency at generation \( t \). Doing so allows us to create a simple graphical representation of the stable and unstable equilibria (Figure 1). Unless stated otherwise, the default parameter values are listed in Table 2.

Our simulation results are consistent with the analytically derived equilibrium infection frequencies (Figure 1). This verifies that our simulations are in line with the behaviour that we would like to capture from our model and are consistent with the results observed in Charlat et al. (2005). This is not surprising given that our model makes similar assumptions on the mechanisms driving Wolbachia invasion dynamics (e.g., fecundity loss and cytoplasmic incompatibilities). In contrast to earlier studies, our population model is dynamically different because of the inclusion of Allee effects and pheromones. Therefore, we determined the relationship between the invasion threshold and these features of the model. We found that the Wolbachia invasion threshold is not affected by Allee effects or by the application of pheromones to the host insect (data not shown). This is important to note because in all of our simulations and analyses, the invasion threshold does not vary as \( \theta \) and \( P \) are adjusted. Finally, we note that the invasion threshold is not affected by any of our demographic parameters (i.e. \( m, \alpha, \gamma \) and \( S_0 \); data not shown).

3.2 | The effect of Wolbachia on the Allee threshold

To determine the dynamical effects of the presence of Wolbachia infection within a population, we find the Allee threshold in insect populations both in the presence and absence of infection. In this section, we ignore stochasticity as well as the release of pheromones (\( P = 0 \)). For a given proportion of infected individuals, we calculate the reproductive rate between two consecutive generations (i.e. \( N_{t+1}/N_t \)) across all population sizes (Figure 2). The equilibria for our population model occur when \( N_{t+1} = N_t \) and there are three equilibria for the parameters used to produce this figure: the first corresponds to population

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
<th>Value</th>
<th>Source</th>
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<td>( N_0 )</td>
<td>Initial population size</td>
<td>Varies 1 – K</td>
<td></td>
</tr>
<tr>
<td>( m, m_1 )</td>
<td>Per capita fecundity (realized, expected)</td>
<td>25</td>
<td></td>
</tr>
<tr>
<td>( \sigma^* )</td>
<td>Standard deviation of fecundity</td>
<td>5</td>
<td></td>
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<tr>
<td>( S_0^* )</td>
<td>Survivorship in the absence of competition</td>
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<tr>
<td>( \alpha )</td>
<td>Related to carrying capacity</td>
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<td>[1]</td>
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<tr>
<td>( \gamma )</td>
<td>Related to intraspecific competition</td>
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<td>[1]</td>
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<tr>
<td>( \mu_F^*, \nu_Y )</td>
<td>Maternal transmission failure</td>
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<td>[1]</td>
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<tr>
<td>( F_X^*, F_Y )</td>
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<td>[1]</td>
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<tr>
<td>( H_x^*, H_y )</td>
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<td>[2]</td>
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<tr>
<td>( \theta )</td>
<td>Strength of Allee effect</td>
<td>Varies 1–0.5K</td>
<td></td>
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<tr>
<td>( P )</td>
<td>Number of pheromone sources</td>
<td>Varies 0–0.15K</td>
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</table>
extinction (stable), the second is the Allee threshold (unstable), and finally the third is the carrying capacity (stable).

In addition to considering the population model in the absence of *Wolbachia*-infected individuals, we calculated the reproductive rates when the population is comprised of 10%, 50% and 90% infected individuals (Figure 2). Given our parameterization, the frequency of infected individuals is chosen to lie above the invasion threshold (which is ~0.5%); therefore, this figure captures the dynamics between two consecutive generations during the replacement process when the population contains the specified distribution of infected and uninfected individuals. As a consequence of cytoplasmic incompatibilities, there is an increase in the Allee threshold. Additionally, there is a slight decrease in the carrying capacity that results from the fecundity loss associated with *Wolbachia* infection. Consequently, the maximum reproductive rate decreases as the proportion of infected individuals increases. Finally, the changes in reproductive rate are most significant during the replacement process (e.g. when 50% is infected) and they diminish when the infection is close to fixation (e.g. 90%). The proportion of *Wolbachia*-infected individuals has a significantly smaller effect on the location of the Allee threshold than the strength of the mate-finding Allee effect itself (Figure 2).

3.3 | Implications for population management

3.3.1 | Deterministic results

In this section, we characterize implications for population management through the release of *Wolbachia*-infected individuals into an insect population. As observed by Dobson et al. (2002), there should be a transient decline in the population size during the replacement of uninfected hosts by *Wolbachia*-infected individuals. Therefore, we find the magnitude of this decline in the presence of Allee effects to determine the conditions under which the replacement process brings the population size below the Allee threshold in a deterministic setting, thereby forcing extinction. This is achieved by running our model over a range of values for the strength of the mate-finding Allee effect ($\theta$). We assume that the initial population size is at its carrying capacity (which is found analytically, as shown in the Supporting Information S1).

We find the minimum population size (relative to $K$) over 50 generations following the introduction of one cytotpe (Figure 3b) and two cytotypes (Figure 3c). Here, values of zero for the minimal population size indicate that the transient reduction in population size brought the population below the Allee threshold, therefore leading to deterministic extinction. The first cytotpe is always released in the second generation, and the release of the second cytotpe is optimized so that the release occurs in the generation that causes the largest decline in population size (see Supporting Information S2 for a detailed explanation of the optimization scheme). While in the main text we assume that all cytotypes have the same infection properties, this assumption is challenged in the Supporting Information (S3) and our qualitative results are unchanged. To ensure that the introduction size is above the invasion threshold, in all simulations we assume that the
introduction is large enough so that the initial infection frequency is 10%. This value lies just above the actual threshold of ~8.5% associated with our parameter values. Therefore, the number of infected individuals introduced in our simulations directly depends on the current host population size.

To determine the success of releases under varying reproductive rates, we replicated the results for four different values of \( S_0 \). For all reproductive rates, the release of two incompatible cytotypes of *Wolbachia* is substantially more effective than a release of a single cytotype. In fact, release of a single cytotype is only effective in driving population extinction when reproductive rates are small. This observation holds more generally; the effectiveness of *Wolbachia* introductions increases as the maximum reproductive rate decreases (Figure 3b,c).

These results suggest that CI management may fail in species with high reproductive rates or alternatively in these species, complementary tactics that either decrease the population size or increase the Allee threshold may be required to amplify the effects of *Wolbachia* introductions. Therefore, we consider the use of mating disruption through the release of sex pheromones (\( P \)) as a supplemental management tactic (as in Equations 3 and 4). We determine the minimum population size following no introduction, one introduction and the introduction of two cytotypes while additionally assuming there is a fixed number of pheromone sources present. These results suggest that combining both methods is significantly more effective than either tactic alone (Figure 3d–f). The associated dynamics can be visualized by considering a plot of the population size with time (Figure 4).

### 3.3.2 Stochastic results

The interplay of Allee effects and stochasticity can be especially important at low population sizes, when the population is at higher risk of stochastic extinction. Therefore, in this section, we determine the ability of *Wolbachia* and mating disruption to drive populations with variable initial population sizes to extinction in the presence of Allee effects and stochasticity.

To achieve this, we determine the probability of extinction based on 500 realizations of the stochastic model (i.e. Equations 9–12) over all relevant combinations of the initial population size and strength of the Allee effect (i.e. the initial population size is at most at 80% of carrying capacity and the relative Allee effect, \( \theta/K \) is no greater than 0.5). To determine the relative roles of environmental and demographic stochasticity, we simulate the model while including both types of stochasticity as well as demographic stochasticity alone.
Furthermore, we find the extinction probability under three scenarios: no introduction of Wolbachia-infected individuals, introduction of one cytotype and the introduction of two incompatible cytotypes. As in the deterministic setting, we assume that the introduction of the first cytotype occurs at the second generation. When two cytotypes are introduced, the generation of the second release is determined in the same way as it is found in the deterministic setting: the second introduction is optimized for each realization so that it occurs in the generation (up to 25 generations) that creates the largest transient decrease in population size resulting from the Wolbachia introduction. The number of generations between releases increases as the strength of the mate-finding Allee effect decreases (see Supporting Information S2). As before, each release is implemented so that the proportion of infected individuals of a given cytotype is 10% (just above the invasion threshold).

Here, we consider populations with relatively high reproductive rates (maximum reproductive rate 3.75); the Supporting Information (S3) provides a sensitivity analysis for lower reproductive rates. We find that the introduction of a single cytotype of Wolbachia leads to negligible increases in the extinction probability across all values of the relative strength of the Allee effect (Figure 5a,b). However, environmental stochasticity increases the uncertainty associated with extinction near the Allee threshold, and there are more apparent increases in extinction probability for Wolbachia introductions for very strong Allee effects (Figure 5d,e). As described in the previous section, the success of Wolbachia releases increases for lower reproductive rates. This finding holds in the stochastic setting (see Supporting Information S3).

In contrast to a single introduction, the release of two incompatible cytotypes is much more effective (Figure 5c,f). Interestingly, when \( \theta \) is relatively high, Wolbachia introductions succeed in driving population extinction independent of the initial population size. This result has the important implication that the success of Wolbachia introductions in driving extinction do not necessarily rely on having a pest population at the initial stage, or at the front, of the invasion. Moreover, in settings with high levels of environmental stochasticity, extinction is possible for much smaller Allee effects than predicted by the deterministic model (Figure 5f).

As explored in the deterministic framework, combining Wolbachia introductions with other methods that increase the Allee threshold (e.g. mating disruption) will likely further increase the success of the overall management strategy. Here, we perform a more global exploration of management options that combine both Wolbachia introductions and pheromone releases. We again consider the population dynamics under three different management regimes: mating disruption only, mating disruption and the introduction of one cytotype, and mating disruption and the introduction of two cytotypes of Wolbachia. In this case, we assume that the relative strength of the Allee effect is fixed and relatively low so that \( \theta/K \) is 0.1; under this condition, Wolbachia introductions alone do not drive the populations to extinction in the deterministic setting (Figure 3b,c). We again find that environmental stochasticity increases the extinction probability for a smaller number of pheromone sources and, importantly, utilizing both mating disruption and CI is much more effective than using mating disruption alone (Figure 6). Therefore, these two methods can serve as complementary tactics for pest management.

## 4 | DISCUSSION

We investigated a population management strategy that considers Wolbachia-induced CI in the presence of Allee effects. In particular, we developed a stochastic population model, building upon the seminal approach of Dobson et al. (2002) and the continuously expanding body of literature investigating the use of Allee effects for the eradication of pest species (Liebhold & Bascompte, 2003; Liebhold et al., 2016; Tobin et al., 2011). Our model demonstrates that the introduction of a relatively small number of incompatible individuals into a pest population that has a strong pre-existing Allee effect can drive the pest population to extinction with no further intervention. These methods are successful more broadly when multiple strains of Wolbachia are introduced. We also demonstrate that extinction is possible for surprisingly large pest populations and that combinations of more than one strain of Wolbachia and mating disruption via sex pheromones work synergistically to increase the population’s extinction risk. Biological control has been studied for decades as an environmentally friendly alternative to pesticide use (e.g. Bale, van Lenteren, & Bigler, 2008; Baumhover et al., 1955; Knipling, 1955; Murdoch, Chesson, & Chesson, 1985), and our study adds to this

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**FIGURE 4** Sample trajectories of the population dynamics (relative to K) when two incompatible cytotypes of Wolbachia are introduced into a population when \( s_0 = 0.15 \). The relative strength of the Allee effect is set to \( \theta/K = 0.15 \), which corresponds to an Allee threshold of ~5.8% of the carrying capacity (as indicated by the horizontal dashed grey line). The black dashed line assumes that there are no pheromone sources, and the solid line assumes that the relative number of pheromone sources \( P/K \) is 0.1. As also evidenced in Figure 3 (e.g. by considering the dashed curve in Figure 3a,b when \( \theta/K = 0.15 \)), intervention with Wolbachia introductions only under these parameter conditions is not successful. A combination of Wolbachia introduction and pheromone sources, however, is successful in achieving population extinction.
FIGURE 5  The grey scale in each plot represents the extinction probability for a given parameter combination based on 500 realizations of the model. In each plot, the initial population size and the strength of the Allee effect $\theta$, both relative to $K$, are varied. We note that the carrying capacity of the population in the absence of Allee effects is $5,500$ with these parameters; therefore, the introduction sizes—which adjust the population size so that there is a 10% infection frequency—do not exceed $\sim 612$ insects. (a–c) Demographic stochasticity only. (d–f) Both demographic and environmental stochasticity. First column: no introduction. Second column: introduction such that infection frequency is at 10%. Third column: two subsequent introductions, both of which ensure the infection frequency is 10% for each cytotype at time of introduction (see Supplementary Information S2 for generation of second introduction). The dashed grey curve is the Allee threshold (i.e. initial populations below the grey curve go to extinction in the deterministic model). Here, $S_0 = 0.15$

FIGURE 6  Fixing the relative strength of the Allee effect $\theta/K = 0.1$, the grey scale in each plot represent the extinction probability for a given parameter combination based on 500 realizations of the model. In each plot, the initial population size and the number of pheromone sources ($P$) relative to $K$ are varied. (a–c) Demographic stochasticity only. (d–f) Both demographic and environmental stochasticity. First column: no introduction. Second column: introduction such that infection frequency is at 10%. Third column: two subsequent introductions. Here, $S_0 = 0.15$

work by providing insight into ways that Wolbachia invasions can take advantage of intrinsic population processes—that is, Allee effects—to manipulate and control pest populations.

An important first step of our modelling work was to uncover the basic interactions between Allee effects and CI. We show that these interactions are weak or non-existent: the Wolbachia invasion
threshold does not depend on the strength of the Allee effect of its insect host, and the Allee threshold has only a marginal decrease in the presence of CI. Therefore, invasion of a particular Wolbachia strain into a population only depends on the critical infection frequency above which invasion succeeds in a deterministic setting (Barton & Turelli, 2011). This invasion threshold corresponds to a proportion of infected hosts above which infection spreads up to almost fixation, and is determined by parameters such as the reduction in egg hatch rate caused by CI, the fitness costs of Wolbachia carriage and the fraction of offspring that inherit the bacteria from an infected mother (Turelli, 1994). The invasion threshold found with our simulation model is consistent with that derived analytically (Turelli & Hoffmann, 1991), and unaffected by the intensity of a mate-finding Allee effect. In addition to adding validation to our model, this result holds interest because most theoretical approaches of Wolbachia invasion dynamics are purely genetic and consider changes in infection frequency without considering host population dynamics. One exception is the work of Hancock et al. (2011) which suggested that Wolbachia invasion thresholds predicted analytically hold for closed populations. Our results are consistent even when, as assumed here, host reproductive rate is affected by both positive and negative density dependence.

In the presence of strong Allee effects, a population below the Allee threshold will be forced to extinction in a deterministic setting, making the Allee effect a central paradigm for conservation (Derdec & Courchamp, 2007; Stephens & Sutherland, 1999), invasions (Taylor & Hastings, 2005), biological control introduction (Fauvergue, Malusa, Giuge, & Courchamp, 2007; Fauvergue, Vercken, Malusa, & Hufbauer, 2012) and as hypothesized in the present work, eradication (Tobin et al., 2011). Whether an Allee effect is weak or strong (and the value of the Allee threshold in the latter case) depends on the strength of the underlying component Allee effect(s) relative to other density-dependent processes. Our simulations of various levels of cytoplasmic incompatibilities in a population with a pre-existing mate-finding Allee effect suggest that the Allee threshold is much less sensitive to variations in the frequency of Wolbachia-infected individuals (0%–90% infected individuals) than it is to variations in mating success (Figure 2b). Wolbachia-induced CI does decrease population growth rate, as expected, but it has a minimal effect on the extinction threshold. Therefore, Wolbachia-induced CI may be considered a culling population management tactic where population size is temporarily decreased as a result of cytoplasmic incompatibilities.

Despite their initial apparent independence, CI and the Allee effect yield interesting properties when acting in concert. Our first analysis that considered the combined occurrence of Allee effects and CI in a deterministic context reveals that the transient decrease in population size is large enough to trigger extinction when the reproductive rate of the host species is relatively low. Extinction caused by the introduction of a single Wolbachia cytotype in populations with higher reproductive rates is only observed for very strong Allee effects (Figure 3a). However, the strength of the Allee effect required for extinction lowers with the introduction of an additional incompatible Wolbachia strain. The resulting insect extinction probability, estimated by simulating the model in the presence of stochasticity, confirmed the interaction between the two processes. In the absence of Wolbachia, we determined the population’s extinction probability as it varies with its population size and the strength of the component Allee effect (Figure 5a,d). Introducing infected individuals results in the extinction of populations that would have persisted otherwise (i.e. a population that is above its Allee threshold can be brought to extinction). Introducing a second incompatible cytotype of Wolbachia increases CIs within the population and, consequently, increases the extinction domain by reducing the severity of Allee effect necessary to trigger extinction (Figure 5).

Nonetheless, our model predicts that although Allee effects and CI combine to drive populations to extinction—even surprisingly large populations—these extinctions may occur for unrealistically severe Allee effects. For instance, after the introduction of two incompatible Wolbachia strains into a population with a maximum reproductive rate of 3.75, extinction is expected when θ/K exceeds 0.25 (Figure 5c–f); that is, extinction occurs if only 25% of all females successfully mate when the population is at 50% of the carrying capacity. Field estimations of mate-finding Allee effects in insects are rare, but it is probable that mating failures only occur at low densities. For instance, in the Gypsy moth L. dispar, mating failures occurred below a density of ~4 (estimated via the rate of male captures on sex pheromone traps), whereas the carrying capacity was estimated similarly around 800 (Tobin et al., 2007, 2013), so that estimation of θ/K in this species could be one or two orders of magnitude lower than that yielding extinction in our model.

Our simulations demonstrate that eradication is much more likely if the introduction of Wolbachia-infected individuals is combined with mating disruption via the release of sex pheromone sources (Figure 6). Interestingly, eradication is not only restricted to small populations but also applies to populations that have reached carrying capacity. Moreover, our results suggest that the two tactics act synergistically: the decrease in population size obtained when CI and mating disruption are combined is higher than the cumulative decrease obtained with each tactic applied separately. Our model therefore supports previous studies that highlight the potential benefit of simultaneously using multiple complementary management tactics (Berec et al., 2016; Blackwood et al., 2012; Suckling et al., 2012). If different tactics benefit from one another, additional methods for controlling a pest population should also be considered. For example, other methods for population control such as parasitism or predation by native natural enemies may also be complementary. Additionally, while our focus was on cytoplasmic incompatibilities, there is evidence that Wolbachia and other bacteria are capable of other reproductive manipulations including male-killing (Dyer & Jaenike, 2004; Richardson, Schiffer, Griffin, Lee, & Hoffman, 2016). Similar conclusions were also made in the recent modelling study of Berec et al. (2016), who suggest that SI T is improved when combined with male-killing bacteria. This suggests the existence of additional avenues for utilizing Wolbachia in the context of pest management.

Theoretical models of population management are currently flourishing much faster than empirical evidence can be obtained; therefore, it is important to discuss the relevance of our predictions and
the feasibility of our proposed methods. First, we have shown that extinction in insect species with high reproductive rates may not be feasible because extinction would require an unrealistic amount of sex pheromone lures to successfully complement the Wolbachia-induced transient decrease in population size. However, in the majority of our simulations, we assigned the maximum reproductive rate a value of 3.75 or, equivalently, per capita fecundity as $m = 25$ and density-independent survivorship of larvae as $S_L = 0.15$. This value is just below the median of the reproductive rates estimated in Hassell et al. (1976) across 24 different insect species. Therefore, we conjecture that these methods would still apply to a variety of pest species. Second, the reproductive rate is just one of the parameters in our model. Although the main text is supplemented with sensitivity analyses, our work is not intended to provide robust quantitative guidelines for a practical application of our proposed management strategies. Instead, our analysis provides general properties of the interactions between Allee effects and cytoplasmic incompatibilities on the population dynamics of an insect pest species. Should our proposed management strategy be applied to a specific insect pest that exhibits Allee effects, thorough investigations would be needed to better quantify parameters that dictate the strength of the processes that underpin the model, including Wolbachia diversity and incompatibility as well as component and demographic Allee effects.

To date, the potential use of Wolbachia-induced CI for pest management is supported by a few but important studies on mosquitoes (Hoffmann et al., 2011; Laven, 1967) and fruit flies (Zabalou et al., 2004). Furthermore, evidence for the existence of both a component and a strong demographic Allee effect exists in the Gypsy moth, which could explain the relative success of mating disruption in this species (Tobin et al., 2007, 2013). Although such empirical advances are indeed promising, they deserve a much stronger body of data and robust cause and effect demonstrations (Fauvergue, 2013). From this perspective, our model was not built as a predictive tool for a specific species. It was rather developed as a heuristic theory yielding qualitative predictions which will hopefully encourage future experimental approaches on the consequences of CI and Allee effects on population extinction.

There is a long and prolific body of research in population dynamics that focuses on understanding the mechanisms stabilizing species near their carrying capacities (e.g. Bernstein, 2000; Hassell & May, 1973; May, 1978). More recently, global climate change and the biodiversity crisis, including population declines, extinctions or biological invasions, point towards the increasing relevance of non-equilibrium ecology (Rohde, 2006) and the biology of small populations (Fauvergue et al., 2012). Transient dynamics are increasingly emphasized (Hastings, 2004) and sometimes considered in the specific context of population management (Ezard et al., 2010; Kidd & Amarasekare, 2012). As first highlighted by Dobson et al. (2002), incompatible cytoplasms transiently coexist during the process of a successful Wolbachia invasion. Consequently, there is also a transient decline in both the reproductive rate and population size of the host population. Here, the transients only last a few generations and this perturbation of the population’s microbiome is the first step in a destabilizing cascade. We show here that the population can then be pushed towards a second step of destabilization, triggered by a mate-finding Allee effect that can be reinforced by the application of mating disruption, which potentially drives the population to extinction.

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**AUTHORS’ CONTRIBUTIONS**

J.C.B. and X.F. conceived the ideas. J.C.B., X.F. and R.V. discussed methodology and modelling choices. J.C.B. and R.V. developed and ran the model. J.C.B. and X.F. led the writing of the manuscript and all authors gave final approval for publication.

**DATA ACCESSIBILITY**

This article does not contain data.

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**REFERENCES**


SUPPORTING INFORMATION

Additional Supporting Information may be found online in the supporting information tab for this article.

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A cascade of destabilizations: combining *Wolbachia* and Allee effects to eradicate insect pests:
Electronic Supplementary Materials

**Contents**

<table>
<thead>
<tr>
<th>Section</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>S1 Stability analysis of population model</td>
<td>S2</td>
</tr>
<tr>
<td>S2 Optimal year of second introduction</td>
<td>S4</td>
</tr>
<tr>
<td>S3 Sensitivity analyses</td>
<td>S5</td>
</tr>
<tr>
<td>S3.1 Strength of environmental stochasticity</td>
<td>S5</td>
</tr>
<tr>
<td>S3.2 Maximum reproductive rate</td>
<td>S7</td>
</tr>
<tr>
<td>S3.3 Introduction of two differing strains</td>
<td>S8</td>
</tr>
<tr>
<td>S3.3.1 Differing fecundity costs</td>
<td>S8</td>
</tr>
<tr>
<td>S3.3.2 Differing transmission failures</td>
<td>S8</td>
</tr>
<tr>
<td>S4 MATLAB code</td>
<td>S9</td>
</tr>
</tbody>
</table>
S1 Stability analysis of population model

In this section, we briefly find and analyze the equilibria of the model presented in Equations (1)-(6) of the main paper (i.e. the population model in the absence of Wolbachia-infected individuals and pheromone sources). This allows us to characterize the general properties of our model and analytically determine the values of the Allee threshold and carrying capacity. Figure 1 in the main text supplements this analysis by demonstrating the equilibrium structure for the Wolbachia invasion dynamics. For simplicity, we assume that $\gamma = 1$ in this analysis – as well as our parameterization – because finding the equilibria for $\gamma \neq 1$ is not tractable. However, we note that setting $\gamma = 1$ still allows us to achieve the desired qualitative properties of the population (e.g. a carrying capacity and Allee threshold). In this case, we can rewrite the full population model using Equation (6) in the main text as

$$N_{t+1} = \frac{mN_t^2S_0}{N_t + \theta + \alpha mN_t^2}$$  \hspace{1cm} (S1)

and we define the right hand side of the equation to be $q(N_t)$. To find the equilibria, we set $N_{t+1} = N_t = N^*$ and either $N^* = 0$ or $N^*$ are the solutions of

$$\alpha m(N^*)^2 + (1 - mS_0)N^* + \theta = 0.$$  \hspace{1cm} (S2)

Written explicitly, $N^* = 0$ and

$$N_{1,2}^* = \frac{mS_0 - 1 \pm \sqrt{(mS_0 - 1)^2 - 4\alpha m\theta}}{2\alpha m}$$  \hspace{1cm} (S3)

where the non-zero fixed points correspond to the Allee threshold and the carrying capacity which we denote as $N_1^*$ and $N_2^*$, respectively. In the absence of Allee effects ($\theta = 0$), there are only two equilibria: zero (or population extinction) and the carrying capacity which is given by

$$N_2^* = \frac{mS_0 - 1}{\alpha m}.$$  

Note that smaller values of $\alpha$ lead to larger carrying capacities. When $\theta > 0$, the non-zero fixed points exist and are distinct and non-negative if

$$(mS_0 - 1)^2 > 4\alpha m\theta,$$  \hspace{1cm} (S4)

and

$$4\alpha m\theta > 0.$$  \hspace{1cm} (S5)

We note that for sufficiently high $\theta$, condition (S4) does not hold; in this case, the reproductive rate is always less than one which leads to population extinction without other interventions. Because additional interventions would not be productive in such a case, when $\theta > 0$ we ensure that conditions (S4)-(S5) are always met in our parameterization.
We can determine the stability of all three fixed points using a standard linear stability analysis; i.e., we find $q'(N^*)$ (defined as the right hand side of Eqn. (1)) evaluated at $N^*$. It is straightforward to show that $q'(0) = 0$, and therefore $N^* = 0$ is always stable because we are in discrete time. For the non-zero fixed points (assuming that they are positive and distinct), we use Mathematica to simplify the expression for $q'(N^*_1)$ and we find that the stability of the Allee threshold ($N^*_1$) can be found by considering

$$q'(N^*_1) = 1 + \frac{\sqrt{(mS_0 - 1)^2 - 4\alpha m\theta}}{mS_0}.$$  \hspace{1cm} (S6)

From Eqns S4-S5, we know that the discriminant is positive. Additionally, $m$ and $S_0$ are positive so that

$$|q'(N^*_1)| = q'(N^*_1) > 1.$$ \hspace{1cm} (S7)

The last inequality holds from Equations S4-S5, and the Allee threshold is therefore unstable. Finally, we can find the stability of the carrying capacity ($N^*_2$) by considering

$$q'(N^*_2) = 1 - \frac{\sqrt{(mS_0 - 1)^2 - 4\alpha m\theta}}{mS_0}.$$ \hspace{1cm} (S8)

Again, the discriminant is positive and $mS_0 > 1$ (i.e., the population must have a maximum reproductive rate greater than one for the population to persist in the absence of interventions) so that the numerator is less than the denominator. Now,

$$|q'(N^*_2)| = q'(N^*_2) < 1.$$ \hspace{1cm} (S9)

Therefore, the carrying capacity is stable. This is also illustrated graphically in Figure 2 in the main text: if the population size is greater than $N^*_1$, the reproductive rate is greater than one and the population size therefore increases toward $N^*_2$. If the population size is greater than $N^*_2$, the reproductive rate of the population is less than one and therefore decreases back to $N^*_2$.

In this stability analysis, we assumed that $\gamma = 1$ for tractability. Although our analysis did not indicate the presence of oscillatory dynamics, it is possible that an alternative assumption for $\gamma$ could produce, for example, decaying transient oscillations toward the stable equilibrium. If oscillatory dynamics were indeed present, we conjecture that our findings would be not impacted. Our analyses demonstrate the conditions under which the transient decline in population size during the process of *Wolbachia* invasion (a pre-established phenomenon) will drive the population to extinction in the presence of Allee effects. This result should not depend on whether there are oscillatory dynamics when approaching (or repelling from) an equilibrium.

S3
The first column displays the probability of extinction (based on 500 stochastic realizations) when two strains of *Wolbachia*-infected individuals are introduced into the population for a given initial population size and strength of the Allee effect. The dashed gray curve in the first column corresponds to the Allee threshold for a given value of $\theta$. The second column displays the optimal mean number of years after the first strain is released to release the second strain. Gray regions indicate that a second release never led to extinction across all stochastic realizations. The first row implements demographic stochasticity only, while the second row implements both demographic and environmental stochasticity.

**S2 Optimal year of second introduction**

When considering the introduction of two strains of *Wolbachia*-infected individuals, we implemented an optimization scheme to numerically determine when the second strain should be introduced. For each realization, we find the smallest number of generations (following an initial *Wolbachia* introduction) such that an introduction of a second incompatible cytotype maximizes the transient decline in population size. This ensures that duration of a management operation based on the release of two incompatible cytotypes of *Wolbachia* is minimized while the impact is maximal.
Our goal is to determine the conditions under which a second strain successfully leads to population extinction; therefore, we summarize these findings by computing the mean optimal lag time across realizations that lead to extinction (Figure S1). When the initial population size lies above the Allee threshold, the optimal number of years between introductions becomes smaller as the strength of the Allee effect increases. However, for small Allee effects and initial population sizes above the Allee threshold (gray line in Figs. S1-AC), regions corresponding to a long time between introductions (i.e. the more yellow colors in Fig. S1-BD) are rarely able to lead to extinction (note the corresponding low extinction probabilities in Fig. S1-AC). Therefore, when two strains of Wolbachia are introduced there usually only needs to be a relatively small time (1-5 years) in between releases and this lag time decreases as the strength of the Allee effect increases.

S3 Sensitivity analyses

In this section we provide several sensitivity analyses of our parameterization. We note that these analyses are not exhaustive and more work should be done when considering these methods in a specific population. We provide sensitivity analyses on the standard deviation for environmental stochasticity as well as the value of the reproductive rate. To save on computational time, for these analyses we only consider the case of a single introduction of Wolbachia-infected individuals. However, given the consistency of these results with our analysis in the main text, we conjecture that the qualitative results in the presence of environmental stochasticity will not change. The final set of sensitivity analyses consider subsequent introductions of two incompatible Wolbachia cytotypes. In these simulations, we assume that the two cyotypes differ in fitness. Here, we only consider demographic stochasticity – again to save on computation time.

S3.1 Strength of environmental stochasticity

We implement environmental stochasticity into our model by assuming that at each generation, the per capita fecundity of the population is drawn from a normal distribution with mean \( m \) and variance \( \sigma^2 \). To ensure positive fecundities, we truncate the distribution to ensure \( m \geq 0 \). Measuring the variability in fecundity for actual populations is intrinsically difficult, so here we provide a sensitivity analysis of the magnitude for \( \sigma^2 \). We assume that the variance scales with the mean, and in the main text we assumed that they are equal. Here, we test two additional values for \( \sigma^2 \): \( \sigma^2 = 1 \) to demonstrate the role of stochasticity when there is little variability, and \( \sigma^2 = 100 \) to observe the effects of
stochasticity when the variability is much higher.

We determine the extinction probability when there is no introduction of Wolbachia-infected individuals as well as when there is a single introduction. The results are straightforward: more variability leads to a greater probability of extinction for a larger range of $\theta$ values (strength of the Allee effect) (Fig. S2). Consequently, the success of Wolbachia invasions designed to drive populations with weaker Allee effects to extinction increases as the variability of environmental stochasticity increases. However, the probability of extinction also increases with environmental stochasticity when there is no introduction of Wolbachia. Therefore, while an introduction of Wolbachia might drive a population with very strong Allee effects to extinction when variability is low, such an introduction might not be necessary when variability is high.

**Figure S2.** Demonstration of impact of the level of environmental stochasticity on the extinction probability as the initial population size and strength of the Allee effect ($\theta$) vary. (C) and (D) are identical to Fig 5D-E in the main text: the variance used in the implementation of environmental stochasticity is $\sigma^2 = 25$. (A) and (B) implement a much smaller variance ($\sigma^2 = 1$), and (E) and (F) assume that environmental stochasticity is much stronger ($\sigma^2 = 100$).
S3.2 Maximum reproductive rate

As discussed in the main text, lower reproductive rates only require a much smaller strength of the Allee effect ($\theta$) relative to $K$ to achieve population extinction following the introduction of Wolbachia. To explore how this translates to the stochastic version of the model, we replicate Fig. 5B in the main text (i.e. a single introduction with demographic stochasticity as the initial population size and $\theta$ relative to $K$ vary) using a smaller reproductive rate (i.e. $S_0$ is adjusted so that the maximum reproductive rate is 2 – as opposed to 3.75 which was used in Fig. 4 of the main text).

As predicted in the deterministic setting, Wolbachia introductions are much more successful for lower values of $\theta$ relative to $K$ when the reproductive rate is small (Fig. S3). However, as indicated by the gray line in Fig. S3 that depicts the Allee threshold, once $\theta/K \approx 0.5$ there is no longer an Allee threshold because the population reproductive rate is below one for almost all population sizes. In this case, the population will be driven to extinction without a Wolbachia introduction.

Figure S3. Results of the sensitivity analysis of the reproductive rate. (A) uses the parameters in the main text such that the maximum reproductive rate is 3.75. This panel assumes there is a single cytotype released. (B) assumes a lower maximum reproductive rate of 2. A single release is much more effective in the latter case; however, note that a value of $\theta/K \approx 0.5$ forces the population’s growth rate to be less than one for all initial population sizes (and therefore there is no Allee threshold). For values of $\theta/K$ above this, the population is driven to extinction independently of the Wolbachia introduction as a result of the strength of the Allee effect.
S3.3 Introduction of two differing strains

In the main text, we considered the plausibility of introducing two subsequent strains of *Wolbachia* into the population as a means of control. For simplicity, we assumed that both strains had identical effects on their host; that is, hosts infected with either strain had identical fecundity losses, rates of transmission failure, and survivorship. Here, we perform some sensitivity analyses of these parameters to assess the potential effects on the extinction probabilities when the two strains have different properties. However, we caution that this analysis is not exhaustive: due to computational restrictions, we limit the analysis to two examples which provides some insight to the parameter sensitivity. Consideration of the properties for particular *Wolbachia* strains would need to be thoroughly explored before practical implementation of these methods is considered.

For both sets of simulations below, it is important to note that the introduction sizes of the cytotype with different parameters from the main text require a different introduction size. This is because the invasion threshold varies as we vary $F_X$ and $\mu_X$; if the introduction size is not above the invasion threshold, that cytotype will die out and consequently the dynamics will not undergo the desired transient decline during the replacement process. For consistency, we choose introduction sizes that mirror that of main text. Specifically, the main text uses an introduction size so that the proportion of the population infected is 10%. This is $\sim 16\%$ above the invasion threshold. We therefore choose introductions that are also $\sim 16\%$ greater than the corresponding invasion thresholds.

S3.3.1 Differing fecundity costs

Here we assume that the two cytotypes differ in their fecundity costs. The first cytotype is assumed to have $F_X = 0.9$ and the second has $F_X = 0.95$. The results are displayed in Figure S4-B; the qualitative results do not change.

S3.3.2 Differing transmission failures

Here we assume that the two cytotypes differ in transmission failures. The first cytotype is assumed to have $\mu_X = 0.06$ and the second has $\mu_X = 0.03$. The results are displayed in Figure S4-C. Again, there is a minimal quantitative change and the qualitative results remain the same.
Figure S4. Results of the sensitivity analysis on introducing two cytotypes. (A) uses identical parameters for both cytotypes with the parameters in the main text, (B) assumes that the first introduced cytotype differs in its fecundity cost, and (C) assumes that the first cytotype differs in its transmission failure. In all, the qualitative results do not change.

S4 MATLAB code

Below we provide MATLAB code that performs a simulation of the model. In it, the option to include stochasticity is provided. Additionally, we provide the code to find the equilibria of the population model. All simulations in the main text and Supplement are extensions of this code. The first file is the main file for running a simulation:

```matlab
%% simulation

%% Input default parameters
p = sample_params();
invy = 100; % number of infected to introduce

% for deterministic define stoch as 0
% for stochastic define stoch as 1
stoch = 1;

% simulate
pops = model(p,invy,stoch);

% plot the population sizes of infected and uninfected
```
uninfect = pops(1,:);
infect = pops(2,:);

plot(uninfect, 'k', 'LineWidth', 1);
hold on
plot(infect, 'r', 'LineWidth', 1);
axis([0 75 0 1])
xlabel('Generation', 'FontSize', 12)
ylabel('Population size', 'FontSize', 12)

The second file contains all parameters:

function [p] = sample_params();

% Demographic parameters
p.lambda = 25;
p.S0 = 0.15;
p.alpha = 0.00002;
p.gamma = 1;

% Wolbachia parameters
p.mu_x = 0.03;
p.Fx = 0.95;
p.Hx = 0.05;

% Simulation parameters
p.num_gens = 75;
p.N_0 = 1000;

% Allee effect
p.theta = 500;

% Parameters related to stochasticity
p.sigma = 5;

For a given parameter set, this code will compute the fixed points (as specified in Equation (2) of the SI).
The final file, saved as “model.m” contains the model itself:

```matlab

function [pops] = model(p,invy,stoch);

% Input default parameters
p = sample_params();

% r is a vector of the equilibria
r = roots([p.lambda*p.alpha (1-p.lambda*p.S0) p.theta]);

The final file, saved as “model.m” contains the model itself:

function [pops] = model(p,invy,stoch);

% Main simulation file

% Initial population sizes
pops = zeros(2,p.num_gens); % rows are: [uninfected infected]
pops(1,1) = p.N_0;

% Assuming introduction happens at first generation
pops(2,1) = invy;

if stoch == 1
    % create a truncated normal distribution
    pd = makedist('Normal');
    pd.mu=p.lambda;
    pd.sigma=p.sigma;
    truncated_dist = truncate(pd,0,inf);
end

% Simulation

for i = 2:p.num_gens
    % compute proportion of females that are infected and uninfected
    if sum(pops(:,i-1)) == 0
        Fminus = 0;
        Fplus = 0;
    else
        Fminus = pops(1,i-1)/sum(pops(:,i-1));
        Fplus = pops(2,i-1)/sum(pops(:,i-1));
    end

end
```
end
% Since assuming 50:50 sex ratio
Mminus = Fminus;
Mplus = Fplus;

% compute "half steps" (proportion of viable offspring with cytotype X vs uninfected)
X_half = Fplus*(1 - p.mu_x)*p.Fx*(Mplus + Mminus);
W_half = Fplus*p.mu_x*p.Fx*(Mplus*p.Hx + Mminus) + Fminus*(Mplus*p.Hx + Mminus);

% proportion of offspring that are infected vs not infected
if (X_half + W_half == 0)
    uninfected_offspring = 0;
    infected_offspring = 0;
else
    uninfected_offspring = W_half/(X_half + W_half);
    infected_offspring = X_half/(X_half + W_half);
end

% total population
pop_tot = sum(pops(:,i-1));

% proportion of offspring that are viable
viable = X_half + W_half;
if pop_tot == 0 % population extinct, need this if statement in case theta = 0
    eggs = 0;
else
    if stoch == 0
        eggs = viable * pop_tot * p.lambda * (pop_tot / (pop_tot + p.theta));
    else
        % implement stochasticity here
        fecund = random(truncated_dist,1,1);
        eggs = poissrnd(viable * pop_tot * p.lambda * (pop_tot / (pop_tot + p.theta)));
    end
end

% emerging adults (includes negative density dependence)
if stoch == 0
    adults = eggs * p.S0/(1 + (p.alpha*eggs)^(p.gamma));
else
    % implement stochasticity here
    adults = binornd( eggs, p.S0/(1 + (p.alpha*eggs)^(p.gamma)));
end

% distribute into new timestep
pops(1,i) = adults*uninfected_offspring;
pops(2,i) = adults*infected_offspring;
end