Spatial structure undermines parasite suppression by gene drive cargo

James J Bull[1], **Christopher H Remien**[2], **Richard Gomulkiewicz**[3], **and Stephen M Krone**[4]

[1]**Department of Biological Sciences, University of Idaho, Moscow, ID, USA**
[2]**Department of Mathematics, University of Idaho, Moscow, ID, USA**
[3]**School of Biological Sciences, Washington State University, Pullman, WA, USA**
[4]**Department of Mathematics, University of Idaho, Moscow, ID, USA**

Corresponding author:
J J Bull[1]

Email address: bull@utexas.edu

ABSTRACT

Gene drives may be used in two ways to suppress vectored diseases. Both involve engineering the drive to spread in the vector population. One has the drive directly depress vector numbers, possibly to extinction. The other leaves intact the vector population but suppresses the disease agent during its interaction with the vector. This second application may use a drive engineered to carry a genetic cargo that suppresses the disease agent. An advantage of the second application is that it is far less likely to select vector resistance to block the drive, but the disease agent may instead evolve resistance to the inhibitory cargo. However, gene drives are expected to spread so fast and attain such high coverage in the vector population that, if disease agent can evolve resistance only gradually, disease eradication may be feasible. Here we use simple models to show that spatial structure in the vector population can greatly facilitate persistence and evolution of resistance by the disease agent. We suggest simple approaches to avoid some types of spatial structure, but others may be intrinsic to the populations being challenged and difficult to overcome.

INTRODUCTION

Genetic engineering has advanced to the point that it is not only possible to introduce arbitrary, massive changes into the genomes of countless species, but it is also possible to engineer changes that rapidly sweep throughout an entire species. The rapid sweeps are enabled by a class of agents called gene drives that function on the principle of biasing transmission in gametes or offspring survival (Hamilton, 1967; Lyttle, 1977; Burt, 2003; Deredec et al., 2008; Gould, 2008; Gould et al., 2008; Report, 2016). Perhaps the most insidious use of a gene drive is one that suppresses population numbers and even potentially drives the population extinct. A more benign form of gene drive is one that sweeps without causing much harm. Any gene drive may be harnessed with additional genetic material (i.e., 'effector gene' or simply 'cargo') that is carried along with the drive as it spreads (Sandler and Novitski, 1957; Gould, 2008; Gould et al., 2008; Gantz et al., 2015). A harmless drive equipped with a cargo provides a fast and simple means of genetically transforming a population, potentially endowing that population with properties that meet social goals without harming the species.

The application of gene drives is limited in a few important ways. They require species with largely obligate sexual reproduction and moderate to high rates of outcrossing. Furthermore, drives that impair fitness are highly susceptible to evolution of resistance (Burt, 2003; Noble et al., 2017; Unckless et al., 2017; Bull et al., 2019). For these reasons, some applications are most amenable to species modification with harmless gene drives carrying genetic cargo. One such application is the use of gene drives to transform disease vectors so that the disease agent (henceforth 'parasite') can no longer be transmitted: the parasite cannot be targeted with a gene drive, but its vector can. Such a gene drive can be designed to have little effect on the vector yet completely block the parasite (Sandler and Novitski, 1957; Burt, 2003; Gantz et al., 2015).

Current technology using CRISPR appears good enough to allow a gene drive to achieve wide coverage of a population (Gantz and Bier, 2015; Hammond et al., 2016; Godfray et al., 2017; Noble et al., 2017; Kyrou et al., 2018). Thus, an inhibitory cargo should be able to eradicate many types of parasite, provided that the cargo cannot be overcome by single mutations in the parasite population. One potential limitation of this approach is that even slight fitness costs of cargo carriage will ultimately lead to a decay of cargo in the vector population, but the decay should often be slow enough to allow parasite suppression for tens to hundreds of generations (Beaghton et al., 2017) – still potentially enough for eradication.

Here we suggest another possible basis of cargo failure, spatial structure in the host population combined with imperfect gene drive coverage/expression. If parasite movement is limited, even small areas of incomplete suppression may allow parasite persistence that serve as nuclei for evolution of parasite resistance. We offer simple models of the sensitivity of parasite persistence and resistance evolution under spatial structure to gauge the plausibility of parasite escape from gene drive control. Our approach potentially applies to any widespread genetic modification of a population, not just gene drives.

BACKGROUND

In this section, we offer a biological framework for the problem addressed in the models section. This framework is easily explained at an intuitive level and helps anticipate the models.

Mass action

Gene drives have traditionally been modeled and understood in the context of well-mixed populations (Prout, 1953; Bruck, 1957; Hamilton, 1967; Burt, 2003; Godfray et al., 2017). The gene drive gains its advantage from heterozygotes, whereby the non-drive allele of a heterozygote is replaced with the drive allele during reproduction. Heterozygote frequency is enhanced with outcrossing (mass action), depressed with inbreeding and some other types of assortative mating. With mass action, an efficient gene drive can spread from low frequencies to near-fixation in close to 10 generations (Burt, 2003; Godfray et al., 2017; Beaghton et al., 2017).

The evolution of a gene drive and its associated cargo can be divided into two phases. The first phase encompasses the short-lived spread of the drive. Although gene drives are potentially highly efficient, various types of fitness effects, imperfections in the drive mechanics and variation in the host population can limit the final coverage of the drive (Deredec et al., 2008; Godfray et al., 2017; Beaghton et al., 2017). Once the drive has spread to its limit, phase two sets in, whereby evolution proceeds according to fitness effects on the host. Any fitness cost stemming from the drive allele or its genetic cargo now begins to select a population reversal toward loss of the cargo and/or drive, favoring alleles resistant to the drive or cargo-free drive states. The speed of this reversal depends heavily on fitness costs and on the initial frequencies of the different parties; it is typically much slower than the spread of the drive (Beaghton et al., 2017). In the long term, a genetic cargo with any fitness cost will be lost. The social benefits of the cargo must therefore be manifest in a time frame compatible with its expected duration.

Spatial structure

Gene drive spread requires reproduction. The spread will thus follow the conduits of reproductive connections in the host population, which may well have a strong spatial component – as when individuals mate with neighbors (North et al., 2013; Beaghton et al., 2016; Tanaka et al., 2017). Any genetic variation that arises in the gene drive or cargo, such as mutations that delete or down-regulate the cargo, will be propagated along those conduits and expand accordingly, leading to spatial structure in parasite inhibition (Beaghton et al., 2017). Even more simply, for purely dynamical reasons, the drive may fail to reach isolated pockets of the population (North et al., 2013). In turn, spatial structure of a genetically variable inhibitor will often mean that different locations of the parasite experience different levels of inhibition. With spatial structure, even small regions of reduced parasite suppression may enable parasite persistence which then facilitate parasite evolution of resistance to the inhibitor.

Pre-existing genetic variation in the host population may also affect gene drive spread and cargo expression (e.g., Drury et al., 2017). For example, a gene drive that avoids harming the host will need to insert into a genomic region that can be disrupted with little or no fitness effects; such a region may thus not be strongly selected to conserve sequences and may be variable across the host population, blocking gene drive spread in some regions. Cargo gene expression may likewise be affected by the genome in

98 which it resides, and geographic variation in genomic content may lead to geographic variation in cargo
99 expression.

100 Our perspective is to investigate the consequences of spatial structure in an inhibitor of the parasite.
101 The details of structure will typically be implementation-specific, but an appreciation for the importance
102 of spatial structure when it exists may be a requisite for successful application of a gene drive cargo.

ANALYSIS

Mathematical models

Our model is most easily applied to an asexual pest/parasite infecting a single host species. Although
not specifically modeled, our problem may be extended in spirit to a parasite transmitted between two
host species, as to *Plasmodium* transmitted by a mosquito to humans and back to mosquitoes; in this
case, the gene drive is introduced into the mosquito to block *Plasmodium* reproduction and transmission.
Our models merely omit the second host, but the effects they reveal apply to this case, subject to some
conditions mentioned in the Discussion.

The social goal is to suppress parasite reproduction with a genetic cargo in the host. To keep the
problem simple, we assume that a gene drive and its cargo have already swept through the host species.
(The actual process of gene drive evolution is thus ignored, and indeed, non-gene-drive methods of cargo
infusion may also be used to achieve this end (Okamoto et al., 2014).) In any one host individual, parasite
inhibition by the cargo occurs in one of three states: (i) full inhibition, (ii) partial inhibition, or (iii) no
inhibition. Partial inhibition would result from weak expression of the cargo in the host; no inhibition
would result from loss of the cargo from the gene drive or resistance to the gene drive itself, such that the
host individual lacks the gene drive and its cargo altogether. The formulation of the model is trivially
extended to multiple states of partial inhibition.

The model counts numbers of parasites in each type of host. Host type merely translates into parasite
reproduction. Notation is

x_h: The relative frequency of hosts of type h

b_{gh}: The fecundity of a parasite genotype g in hosts of type h

n_{gh}: The current number of parasites of genotype g in hosts of type h

where parasite genotype $g \in \{0, 1, \ldots, M\}$ and host type $h \in \{0, 1, \ldots, H\}$. Here, host type 0 indicates
hosts with no cargo (hence no parasite suppression), and larger values of h correspond to increased levels
of suppression, with H denoting the number of types of cargo-carrying hosts differing in some aspect
of parasite suppression; parasite genotype 0 is the wild type (with no protection against the cargo), and
larger values of g correspond to mutant strains with increased levels of resistance to the cargo.

To approximate the separation of phases between rapid gene drive spread and the subsequent effect
of cargo on the parasite, we let the x_h and b_{gh} be constant in time, so the only changes are in parasite
numbers. Time is discrete. For biological reasons, in the case of three host types, we also assume that
the fecundity of parasite genotype 0 satisfies $b_{00} > 1 > b_{01} > b_{02}$, which means that the parasite has
negative growth in all host types except the one lacking cargo.

To set the stage for a structured host population, we suppose that hosts are clustered in *patches*
of similar host types, patch types designated by subscript h. ('Host' and 'patch' are used somewhat
interchangeably below, but 'patch' merely helps convey structure.) Patches could result, for example,
from limited host migration and gene drive spread through the host population in a manner that follows
host population structure. The clustering of hosts and the consequent movement of parasites between
patches determines the extent to which structure is experienced by the parasites. To establish a mass-action
baseline, adult parasites reproduce and release all progeny into a random pool, from which they settle
into each of the $H + 1$ patch types at frequencies $x_0, x_1, \ldots, x_H$. Spatial structure is modeled indirectly by
assuming that a fraction α of the progeny from each patch type remains in the same patch type without
entering the random pool. This process is fundamentally the same as migration in standard population
genetics problems (Crow and Kimura, 1970). Our formulation is different, in that α describes a lack of
movement from the natal site instead of movement between patches/populations; our formulation leads
naturally to calculating the null case of mass action ($\alpha = 0$), whereby the efficacy of a cargo is maximal
and is presumably the default expectation in gene drive applications.

We first consider a simple system of linear dynamical equations. Let

$$N_g = \sum_{h=0}^{H} n_{gh} b_{gh} \tag{1}$$

denote the overall progeny output of strain g across all environments. Using primes to indicate one generation hence,

$$n'_{gh} = \alpha n_{gh} b_{gh} + (1-\alpha) N_g x_h \quad . \tag{2}$$

The joint dynamics for genotype g across all host types can be written as a matrix projection recursion (Caswell, 2006). Dropping the genotype subscript g, the recursion has form

$$\mathbf{n}' = \mathbf{M}\mathbf{n} \tag{3}$$

where $\mathbf{n}$ and $\mathbf{n}'$ are $(H+1)$-dimensional column vectors with hth element n_h and n'_h, respectively. The $(H+1) \times (H+1)$ matrix $\mathbf{M}$ has diagonal elements $M_{hh} = b_h[\alpha + (1-\alpha)x_h]$ and off-diagonal elements $M_{hi} = b_i(1-\alpha)x_h$, $i \neq h$. In words, matrix element M_{hi} describes the rate that individuals who originate in patch type i contribute to the abundance in patch type h at the next time step. The densities of the genotype g at any time t, $\mathbf{n}(t)$, can be computed as $\mathbf{n}(t) = \mathbf{M}^t \mathbf{n}(0)$ (Caswell, 2006).

With a sufficiently large b_{ij} and $n_{ij} > 0$, these equations allow the persistence of genotype i in patch j even when $x_j = 0$. This effect occurs because, once a patch is populated, it may continue in the absence of further immigration. Because we interpret x_j as the fraction of hosts of type j, we require $x_j > 0$ for any host type that harbors parasites. This condition is further imposed when enforcing a carrying capacity (see below).

These equations allow unlimited population growth which can cause unrealistic dynamics at high densities, so we sometimes choose to study a related model that imposes density regulation. We limit our description of this version to the case of two patches and two strains for simplicity.

Introducing density regulation

To understand whether parasites persist or die out, dynamics at low population densities (characterized by innate fecundities) should be sufficient to distinguish the two outcomes, at least deterministically. However, when considering evolution of alternative genotypes in a persisting population, density regulation may be important, especially because a small portion of the environment with cargo-free hosts may be limited in how many parasites it can support – a small patch may allow parasite persistence but have little impact on parasite numbers across the entire environment. We thus introduce a simple form of patch-specific density regulation that will be used in some numerical trials of two patch types, 0 and 1.

Let K be the overall upper density limit for the environment and let $K_0 = Kx_0$, $K_1 = Kx_1$ denote the ceilings for patch 0 and 1, respectively. Let $\Pi_1 = \min\{n_{01}b_{01} + n_{11}b_{11}, K_1\}$ be total parasite progeny production emanating from patch 1, and $\Pi_0 = \min\{n_{00}b_{00} + n_{10}b_{10}, K_0\}$ be the total parasite progeny production from patch 0, each limited locally, without respect to regulation in the other patch type. The extension to more than two patch types is straightforward.

To decide how overall parasite production in a patch is divided between two genotypes when progeny output is limited by carrying capacity, let $p_{11} = n_{11}b_{11}/(n_{01}b_{01} + n_{11}b_{11})$ denote the fraction of offspring that are strain 1 within patch 1, and $p_{01} = 1 - p_{11}$ the fraction in patch 1 that are strain 0. Similarly, in patch 0 we have fractions $p_{10} = n_{10}b_{10}/(n_{00}b_{00} + n_{10}b_{10})$ and $p_{00} = 1 - p_{10}$. Accounting for mixing and assuming both genotypes are density-regulated the same, the total number of genotype 1 entering patch 1 is

$$\alpha p_{11}\Pi_1 + (1-\alpha)[p_{10}\Pi_0 + p_{11}\Pi_1]x_1 \tag{4}$$

and the total number of genotype 0 entering patch 1 is

$$\alpha p_{01}\Pi_1 + (1-\alpha)[p_{00}\Pi_0 + p_{01}\Pi_1]x_1 \quad . \tag{5}$$

Analogous equations apply to genotypes 0 and 1 entering patch 0. Note that mixing can lead to a temporary violation of local carrying capacities when patches are repopulated by adults, but the capacity limit is imposed again at the next round of reproduction.

Parasite persistence is facilitated by spatial structure

The growth or suppression of the parasite depends on whether the magnitude of the leading eigenvalue of $\mathbf{M}$ in eqn (3) exceeds 1 (density dependence can be ignored when addressing persistence). Although the characteristic equation is easily found, the leading eigenvalue (λ_{max}) of a genotype is tractable for arbitrary H only for the extremes of mass action ($\alpha = 0$), and complete parasite isolation among host types ($\alpha = 1$). For mass action,

$$\lambda_{max,g} = \sum_h x_h b_{gh} \quad . \tag{6}$$

As is well appreciated for mass action, parasite growth is just the weighted average fecundity across all host types. Small levels of weak suppression (i.e., low values of x_0 despite possibly high values of b_{00}) will not themselves enable parasite persistence except when parasite fecundity is extraordinarily high. When more than one genotype has an eigenvalue greater than 1, density dependence will determine which one prevails (see below).

With complete parasite separation across the host types ($\alpha = 1$),

$$\lambda_{max,g} = b_{g0} \tag{7}$$

(λ_{max} is associated with patch type 0 because cargo-free hosts are assumed to offer the highest fecundity of all patches, regardless of genotype). Here, the parasites inhabiting each host type have their own eigenvalue, and any host type with $b_{gh} > 1$ will allow parasites of genotype g to persist in that patch type (subject to competition among different parasite genotypes). In this extreme, the values of x_h no longer matter: even a small fraction of permissive hosts will allow the parasite to persist.

The question motivating our study is how sensitive parasite persistence is to α – the spatial structure parameter. As the eigenvalue for several patch/host types is unwieldy, we start by reducing the problem to just two patch types, $g = 0$ and 1 (fully permissive and fully blocking of wild-type); this reduction also simplifies patch type abundances: $x_0 + x_1 = 1$. For this case, the largest eigenvalue is

$$\lambda_{max,g} \quad = \frac{1}{2} \left\{ \alpha \left[b_{g1} x_0 + b_{g0}(1-x_0) \right] + b_{g1}(1-x_0) + b_{g0}x_0 \right\} \quad + \tag{8}$$

$$\frac{1}{2} \sqrt{ \left[b_{g1} \left((1-\alpha)(1-x_0) + \alpha \right) + b_{g0} \left((1-\alpha)x_0 + \alpha \right) \right]^2 - 4\alpha b_{g1} b_{g0} \left[(1-\alpha)(1-x_0) + (1-\alpha)x_0 + \alpha \right] } \quad .$$

The values of the eigenvalues for different fecundities in the two patch types are shown in Fig. 1, the panels differing in the α and x_0 values. (For reasons noted above, most emphasis should be placed on whether $\lambda_{max} > 1$, rather than its magnitude above 1.) It is clear that persistence is possible, though typically a large b_{00} value is required if b_{01} is small. But for small x_0, when b_{01} exceeds 1, persistence is possible despite low fecundity values in cargo-free hosts, b_{00}. In addition to the effect of fecundity, there are also effects of α and x_0; one interesting effect is that the isoclines are visually step-like except in the lower left panel. For those cases, fecundity on cargo-bearing hosts has little effect on the parasite growth rate until $b_{01} \approx \lambda_{max}$.

Most empirical interest is likely to be in the extreme case that the cargo completely suppresses wild-type parasite reproduction ($b_{01} = 0$), as that would be the goal of the engineer. Indeed, any gene drive release could be avoided until such an appropriate inhibitor was found. This case corresponds to the the the sliver defined by the vertical axes in Fig. 1. Furthermore, this case is highly tractable:

$$\lambda_{max,0} \big|_{b_{01}=0} \equiv \lambda_0 = b_{00} \left[x_0 + \alpha(1-x_0) \right] \quad . \tag{9}$$

Persistence in this case requires $\lambda_0 = b_{00} \left[x_0 + \alpha(1-x_0) \right] \geq 1$. This implies, for example, that persistence of the parasite is assured even when completely inhibited by the cargo as long as fecundity in its absence, b_{00}, exceeds $\left[x_0 + \alpha(1-x_0) \right]^{-1}$. Consideration of this minimum cargo-free fecundity (Fig. 2) shows that spatial structuring with α well below 1 (e.g., 0.5) enables parasite persistence for even rare cargo-free hosts, as long as the parasite can grow there moderately well (e.g., $b_{00} \geq 2$). The results are largely insensitive to the value of x_0 for $\alpha \geq 0.6$.

It is also easy to see that, when x_0 is small, as it should be if the engineering worked as expected, the effect of increasing α is approximately the same as increasing the frequency of permissive hosts (x_0) –

Figure 1. Contour plots of eigenvalues (eqn (8)) for different fecundities in the two patch types (b_{00}, b_{01}) for the wild-type parasite genotype ($g = 0$). Panel titles indicate the α and x_0 values; λ_{max} values are given on the contours.

both are mostly linear effects and of the same magnitude. Whereas x_0 is somewhat under the control of the engineer, α is not and has the potential to thwart parasite eradication. Thus, a very small x_0 could easily cause parasite population collapse in the mass action case, but $b_{00}\alpha > 1$ is sufficient for persistence no matter how small x_0. Persistence could be achieved by a cargo-free patch merely large enough that parasite progeny often did not disperse beyond the patch edges.

The ease of resistance evolution depends somewhat on ecology

At the population level, the evolution of resistance to a genetic cargo has many parallels with the evolution of resistance to antibiotics and pesticides. The latter problems are thoroughly studied, and it is well appreciated that resistance is especially prone to evolve under intermediate levels of drug/chemical

Figure 2. Minimum fecundity in cargo-free hosts required for parasite persistence when the cargo causes complete inhibition, ($b_{01} = 0$) from eqn (9)). The curves intersect $\alpha = 0$ at $b_{00} = \frac{1}{x_0}$. For $\alpha \geq 0.6$, there is little effect among these values of x_0 on the minimum fecundity. The horizontal line at $b_{00} = 1$ indicates the minimum fecundity required for the parasite to persist in the absence of cargo, $x_0 = 1$.

application (e.g., Gould and MacKenzie, 2002; Andersson and Hughes, 2012; Tabashnik and Gould, 2012; Neve et al., 2014; Gould et al., 2018). Inhibition by a gene drive cargo is different from chemicals in that the levels of inhibition over time are established at fixed, semi-permanent levels in the near term. They are also largely unchangeable, at least in the short term, should it be discovered that they are inadequate.

Of the many factors to consider, an important one is the mutational spectrum of resistance: a cargo for which simple, single mutations can allow parasite persistence seems doomed to fail, and intuition suffices for preliminary understanding, at least deterministically. Our interest instead lies in gradual evolution and the selection of weak resistance mutations. It might be hoped that cargo-based inhibitors can be found for which resistance mutations are impossible, but a more realistic hope is that inhibitors could be found for which resistance can evolve only gradually.

The evolution of resistance can be considered in two contexts. One is known as 'evolutionary rescue,' whereby the population is in decline and a resistance mutation potentially reverses the decline (Martin et al., 2013; Uecker et al., 2014; Hufbauer et al., 2015; Gomulkiewicz et al., 2017). The other context, the one addressed here, is resistance evolution in a persisting population. For our application, we imagine that the parasite is persisting because of spatial structure and would go extinct under mass action – the large majority of hosts inhibit parasite reproduction because of the cargo. We address how selection acts on a weakly resistant mutation, a mutation that is not necessarily sufficient to provide positive parasite growth from inhibitory hosts alone.

It might seem valid to evaluate long term resistance evolution from a comparison of eigenvalues of

wild-type and mutant growth, in which case the preceding figures could be used to infer evolution of alternative genotypes. However, we are considering resistance evolution in established populations at which density dependence is operating. When density dependence operates locally, as assumed here, it will have a different effect on parasite growth in cargo-free patches than in inhibitory patches. If the wild-type parasite cannot grow in inhibitory patches but the mutant can, the fecundities of both genotypes will be suppressed by density dependence in cargo-free patches but the mutant's fecundity in the inhibitory environment will be unaffected (at least while rare). Eigenvalue calculations do not include these fecundity modifications. Our model of resistance evolution uses the system of equations in (3) but with the density-dependent carrying capacity enforced as in eqns (4) and (5).

This model was evaluated numerically for different combinations of α, x_0, and genotype fecundities. We focused on the case of a wild-type (starting genotype) that was unable to grow in the inhibitory environment of cargo-bearing hosts ($b_{01} = 0$), and for which the mutant could grow in the inhibitory environment at some reduction in its ability to grow in the cargo-free environment (i.e., a trade-off was imposed between growth in the two patch types). The patterns across trials are qualitatively similar and easily comprehended (Fig. 3). Resistance could invariably evolve if it did not incur too much of a cost to growth in the cargo-free hosts.

Some 'ecological' patterns are evident. (1) A large effect of α – host patch fidelity – on evolution exists in some parameter ranges. Thus, to displace wild-type, the mutant is more sensitive to a reduction in cargo-free fecundity at larger α – until fecundity in cargo-bearing hosts is high enough that the mutant could sustain itself in those hosts alone ($b_{11} > 1$). This effect is evident when comparing lower portions of the top right and middle right panels in Fig. 3 and was seen in other trials (not shown). We interpret this effect as that higher values of α increasingly partition growth in the two patch types, so that any fitness loss in the permissive patch – the one sustaining the population – is increasingly penalized, but only up to the point that the mutant parasite can maintain itself in cargo-bearing hosts.

(2) Patch size x_0 also appears to have a large effect, but in the opposite direction as that of α: as x_0 increases, resistance evolves more easily (i.e., tolerates larger reductions in cargo-free fecundity, $b_{00} - b_{10}$). This effect is seen by comparing the first 3 panels of Fig. 3 which vary only x_0. Although not shown, the trend does not continue at high values of x_0, and the effect reverses. Indeed, mutants whose $b_{11} > 1$ are not favored at high x_0 values if they suffer too much cost in b_{10}. We do not pretend to grasp these patterns intuitively. One challenge in understanding these outcomes is that x_0 has effects at different steps of the life cycle: with higher x_0, the cargo-free patch has an increased carrying capacity and thus produces more progeny; some of these progeny return to patch 0 and the others go to the random pool. But the increase in x_0 reduces the number of random-pool progeny that land in patch 1 and thereby reap the benefits of resistance.

DISCUSSION

Genetic engineering has brought us to the brink of being able to introduce selfish genetic elements (known as gene drives) in countless species (Esvelt et al., 2014; Report, 2016; Collins, 2018). Although safety concerns and regulatory hurdles have so far prevented the release of engineered gene drives, it seems inevitable that they will be released and eventually become a standard intervention for pest management and disease control – if they work as expected. Much excitement is about using gene drives for population suppression (Hamilton, 1967; Lyttle, 1977; Burt, 2003; Godfray et al., 2017; Lambert et al., 2018), but an alternative use that may encounter fewer regulatory obstacles is for spreading genetic 'cargo' through a population (Sandler and Novitski, 1957; Gould, 2008; Gantz et al., 2015; Beaghton et al., 2017). In this latter application, the gene drive is a rapid and potentially harmless means of genetically engineering an entire population to carry a novel gene of interest. Thus, a mosquito that transmits *Plasmodium* might be targeted for eradication by a gene drive (Kyrou et al., 2018) or be targeted with a different gene drive to spread a cargo that express anti-*Plasmodium* antibodies that block *Plasmodium* reproduction in the mosquito (Gantz et al., 2015).

Our study addressed the latter type of application and even assumed that the gene drive cargo had already spread in the host (e.g., mosquito) population. If this inhibitor fully suppresses the parasite in any host individual expressing the cargo – if it operates as expected – how might parasites persist despite our efforts to eradicate? Our emphasis here is on the possible contribution of spatial structure in the host population to parasite escape. Even when the gene drive successfully spreads the cargo to most of the host population, spatial structure combined with imperfect gene drive spread may leave pockets of cargo-free

mosquitoes that allow the parasite to persist locally.

Our findings suggest that spatial structure in the host population can contribute to – indeed be sufficient to – enable parasite persistence against a cargo that would otherwise eradicate the parasite. Pockets of parasite persistence then foment the evolution of resistance to escape the cargo, unless resistance mutations cannot arise. The pockets need not be large, possibly representing a very small fraction of the range of the species targeted by the drive.

Our results follow other work suggesting another reason that cargo-bearing gene drives may fail to eradicate parasites: the cargo frequency will begin to decline as soon as the gene drive carrier has reached its zenith in the parasite population (Beaghton et al., 2017). Thus, independent of spatial structure, any cost to carrying the gene drive element or cargo will select a loss of those elements once the gene drive spread has ended. An additional problem facing cargo-bearing gene drives is that, at least for those using homing endonuclease technology, they must target genomic sequences that are not essential to the host and thus have few selective constraints. Weakly selected sequences can tolerate variation that would block a gene drive. So the types of cargo-bearing drives most likely to be implemented will target sites that are most capable of evolving resistance to the drive.

There are two requirements for parasite escape under the models studied here: spatial structure and genetic variation in cargo presence/expression that coincides with the spatial structure. Observation of a spatially structured population would indicate that the first requirement is met. Pre-existing genetic variation in the host population may even contribute to variation in cargo presence or expression, as when existing variation directly resists gene drive spread or affect cargo expression, and spatial structure of the variation would lead to both requirements being satisfied. But even if existing genetic variation does not affect gene drive evolution or cargo expression, any existing spatial structure may become a problem for genetic variation that evolves during gene drive spread. Gene drive spread can generate its own variation in cargo presence/expression by evolving as it follows demographic paths of reproduction in the host population (Beaghton et al., 2017). Such variation would be maintained by any spatial structure intrinsic to the host population.

The problem of parasite persistence and resistance evolution in response to a gene drive cargo has parallels with evolution of resistance in other contexts: antibiotic treatment of bacteria, use of chemical pesticides, and even genetically engineered 'Bt' crops. There is a widespread recognition that intermediate levels of pesticides and antibiotics favor the evolution of resistance (e.g., Gould and MacKenzie, 2002; Andersson and Hughes, 2012; Tabashnik and Gould, 2012; Neve et al., 2014; Gould et al., 2018). These problems are more often cast as stemming from temporal variation in dose rather than spatial variation, but the two types of heterogeneity are similar over the right time scales. In contrast, the practice of planting non-engineered strains among genetically engineered Bt crops to delay insect resistance evolution is explicitly one of destroying spatial structure (Tabashnik and Gould, 2012) and highlights the importance of spatial structure to the evolution of resistance. Gene drive cargo expression is presumably stable in time (for individual hosts and their dependants), at least for the short term, so parasite escape is primarily a problem of spatial structure rather than one of temporal structure.

The models leave many questions unanswered about the nature and magnitude of spatial structure required to enable parasite persistence. Indeed, it is the combination of spatial structure in conjunction with genetic variation that matters. The right kind of spatial structure might exist in one part of a species range, the appropriate genetic variation in another, yet the parasite be suppressed by the cargo. Furthermore, the permanence of spatial structure in the host population will depend heavily on host dispersal patterns, and it is the combined movement over the the life cycle of the parasite that determines the relevant structure. For parasites that alternate between two host species (e.g., mosquito-borne diseases of humans), a highly structured mosquito species will not enable parasite persistence if the second host – humans – is sufficiently mobile. Issues such as longevity of the spatially structured patches, plus averages and variances of host dispersal distances may need to be explored in the context of specific applications before understanding the potential for parasite escape.

Evolution of resistance is perhaps the ultimate concern. If cargo can be engineered to be resistance-proof, spatial structure will be only a temporary setback as additional interventions are implemented. Spatial structure will have the largest impact in facilitating evolution of resistance to cargo when resistance can evolve only gradually, in small steps. In this case, parasite eradication might well be achieved were it not for structure, but the structure provides the nucleus for gradual evolution of resistance that ultimately enables the parasite to maintain itself on cargo-bearing hosts. Stacking multiple inhibitors in the same

host individual (as proposed for malaria (Gantz et al., 2015)) may, in the ideal case, prevent stepwise resistance evolution. Here the concern is an evolutionary loss of part of the cargo, so that only single inhibitors operate in some hosts; spatial structure would then contribute to evolution of resistance.

It has been convenient to focus on cargo-free patches as the type of reservoir enabling parasite escape. An alternative – or additional – type of refuge to consider is patches of intermediate cargo expression, enabling the parasite to persist at some level and directly favoring resistance. Intermediate patches may occur with many levels of expression and may arise because of genetic background effects in the host species or may arise by mutation during the spread of the gene drive itself. Thus, if cargo expression is costly to the host, drives with reduced expression will spread even faster than drives with the original engineering. These mutant-drive cargoes will form their own spatial structure as they spread, which may then be maintained by intrinsic host structure. The engineering faces a delicate balance between cargo over-expression and cargo under-expression. Over-expression may impose a fitness cost that selects against the drive/cargo, whereas the under-expression risks facilitating parasite persistence and evolution of resistance. The effect of patch intermediacy on persistence may be evaluated for our 2-patch case in Fig. 1 merely by considering one of the two patches to be intermediate instead of extreme (e.g., b_{00} would be depressed or b_{01} would be greater than 0). In a sense, our two patch model describes a worst-case scenario for parasite persistence; we expect permissive conditions in the real world will be broader than our results suggest.

Our findings can be used to inform implementation practices most likely to succeed. In any implementation, inhibitory cargo should be chosen so that resistance evolution is difficult (i.e., requires multiple steps) or impossible, as inhibitors that can be overcome with single mutations seemed assured of eventual failure. Beyond that, there are a few design features that may facilitate parasite suppression and work to limit evolution of resistance.

- **Prevent emergence of spatially structured variation**. Existing spatially structured genetic variation in a wild species may be difficult to change, although inundating small areas with release of lab-reared strains may offer a temporary solution – as underlies the sterile insect technique (Klassen and Curtis, 2005; Dyck et al., 2005). However, spatial variation that arises from gene drive spread and evolution (Beaghton et al., 2017) may be reduced by gene drive release at multiple sites (North et al., 2013), especially sites of import, such as population centers: the different waves of advance will collide with other waves, reducing any spatial evolution from single release points. Releasing multiple, independent drives in the same population, as proposed to overcome resistance evolution within unstructured populations (Burt, 2003; Deredec et al., 2011), may limit the extent to which any area is completely free of cargo from at least one drive.

- **Target areas of incomplete coverage**. Following gene drive spread, areas can be assessed for cargo presence and expression. Regions identified as having inadequate coverage can be targeted for additional interventions to offset the limited effect on parasites.

- **Gene knockouts versus introduced transgenes**. Low levels of cargo expression may occur in patches, just as do cargo-free hosts. Inadequate expression of cargo may be even more conducive to resistance evolution than is cargo absence. A gene drive system that knocks out a non-essential host gene required for parasite reproduction/transmission may be less subject to intermediate expression than is a cargo transgene. Unfortunately, ablation of a non-essential host gene may carry larger fitness defects than does expression of a foreign transgene, and thus select resistance to the drive.

The models analyzed here depict spatial structure abstractly. As they suggest that spatial structure may seriously impede gene drive implementations using cargo – an otherwise promising use of gene drives – we recommend the study of explicit models of spatial structure to further evaluate the suggestions presented here. We expect spatial structure to generally provide parasite escape from inhibitory genetic cargo in the host population, but there are likely to be many details that affect the ease of escape and evolution of resistance. Indeed, it may be desirable to study models specific to the biology of an implementation before making critical decisions about engineering and sites of release (e.g., Eckhoff et al., 2017; Lambert et al., 2018; North et al., 2019).

ACKNOWLEDGMENTS

JJB and CHR acknowledge support from National Institutes of Health (R01GM 122079 and P20GM 104420). RG was supported by a WSU Honors College Distinguished Professorship and NSF Grant DEB 1354264. The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

REFERENCES

Andersson, D. I. and Hughes, D. (2012). Evolution of antibiotic resistance at non-lethal drug concentrations. *Drug Resistance Updates: Reviews and Commentaries in Antimicrobial and Anticancer Chemotherapy*, 15(3):162–172.

Beaghton, A., Beaghton, P. J., and Burt, A. (2016). Gene drive through a landscape: Reaction-diffusion models of population suppression and elimination by a sex ratio distorter. *Theoretical Population Biology*, 108:51–69.

Beaghton, A., Hammond, A., Nolan, T., Crisanti, A., Godfray, H. C. J., and Burt, A. (2017). Requirements for driving antipathogen effector genes into populations of disease vectors by homing. *Genetics*, 205(4):1587–1596.

Bruck, D. (1957). Male segregation ratio advantage as a factor in maintaining lethal alleles in wild populations of house mice. *Proceedings of the National Academy of Sciences of the United States of America*, 43(1):152–158.

Bull, J. J., Remien, C. H., and Krone, S. M. (2019). Gene-drive-mediated extinction is thwarted by population structure and evolution of sib mating. *Evolution, Medicine, and Public Health*, 2019(1):66–81.

Burt, A. (2003). Site-specific selfish genes as tools for the control and genetic engineering of natural populations. *Proceedings. Biological Sciences / The Royal Society*, 270(1518):921–928.

Caswell, H. (2006). *Matrix population models*, volume 3. Wiley Online Library.

Collins, J. P. (2018). Gene drives in our future: challenges of and opportunities for using a self-sustaining technology in pest and vector management. *BMC proceedings*, 12(Suppl 8):9.

Crow, J. F. and Kimura, M. (1970). *An introduction to population genetics theory*. Burgess Pub. Co.

Deredec, A., Burt, A., and Godfray, H. C. J. (2008). The population genetics of using homing endonuclease genes in vector and pest management. *Genetics*, 179(4):2013–2026.

Deredec, A., Godfray, H. C. J., and Burt, A. (2011). Requirements for effective malaria control with homing endonuclease genes. *Proceedings of the National Academy of Sciences of the United States of America*, 108(43):E874–880.

Drury, D. W., Dapper, A. L., Siniard, D. J., Zentner, G. E., and Wade, M. J. (2017). CRISPR/Cas9 gene drives in genetically variable and nonrandomly mating wild populations. *Science Advances*, 3:e1601910.

Dyck, V. A., Hendrichs, J., and Robinson, A. S. (2005). *Sterile Insect Technique*. Springer-Verlag.

Eckhoff, P. A., Wenger, E. A., Godfray, H. C. J., and Burt, A. (2017). Impact of mosquito gene drive on malaria elimination in a computational model with explicit spatial and temporal dynamics. *Proceedings of the National Academy of Sciences*, 114(2):E255–E264.

Esvelt, K. M., Smidler, A. L., Catteruccia, F., and Church, G. M. (2014). Concerning RNA-guided gene drives for the alteration of wild populations. *eLife*, page e03401.

Gantz, V. M. and Bier, E. (2015). Genome editing. The mutagenic chain reaction: a method for converting heterozygous to homozygous mutations. *Science (New York, N.Y.)*, 348(6233):442–444.

Gantz, V. M., Jasinskiene, N., Tatarenkova, O., Fazekas, A., Macias, V. M., Bier, E., and James, A. A. (2015). Highly efficient Cas9-mediated gene drive for population modification of the malaria vector mosquito Anopheles stephensi. *Proceedings of the National Academy of Sciences of the United States of America*, 112(49):E6736–6743.

Godfray, H. C. J., North, A., and Burt, A. (2017). How driving endonuclease genes can be used to combat pests and disease vectors. *BMC biology*, 15(1):81.

Gomulkiewicz, R., Krone, S. M., and Remien, C. H. (2017). Evolution and the duration of a doomed population. *Evolutionary Applications*, 10(5):471–484.

Gould, F. (2008). Broadening the application of evolutionarily based genetic pest management. *Evolution*, 62(2):500–510.

Gould, F., Brown, Z. S., and Kuzma, J. (2018). Wicked evolution: Can we address the sociobiological dilemma of pesticide resistance? *Science (New York, N.Y.)*, 360(6390):728–732.

Gould, F., Huang, Y., Legros, M., and Lloyd, A. L. (2008). A killer-rescue system for self-limiting gene drive of anti-pathogen constructs. *Proceedings. Biological Sciences / The Royal Society*, 275(1653):2823–2829.

Gould, I. M. and MacKenzie, F. M. (2002). Antibiotic exposure as a risk factor for emergence of resistance: the influence of concentration. *Journal of Applied Microbiology*, 92 Suppl:78S–84S.

Hamilton, W. D. (1967). Extraordinary sex ratios. A sex-ratio theory for sex linkage and inbreeding has new implications in cytogenetics and entomology. *Science (New York, N.Y.)*, 156(3774):477–488.

Hammond, A., Galizi, R., Kyrou, K., Simoni, A., Siniscalchi, C., Katsanos, D., Gribble, M., Baker, D., Marois, E., Russell, S., Burt, A., Windbichler, N., Crisanti, A., and Nolan, T. (2016). A CRISPR-Cas9 gene drive system targeting female reproduction in the malaria mosquito vector Anopheles gambiae. *Nature Biotechnology*, 34(1):78–83.

Hufbauer, R. A., Szűcs, M., Kasyon, E., Youngberg, C., Koontz, M. J., Richards, C., Tuff, T., and Melbourne, B. A. (2015). Three types of rescue can avert extinction in a changing environment. *Proceedings of the National Academy of Sciences of the United States of America*, 112(33):10557–10562.

Klassen, W. and Curtis, C. F. (2005). History of the sterile insect technique. In Dyck, V. A., Hendrichs, J., and Robinson, A. S., editors, *Sterile Insect Technique*, pages 3–36. Springer Netherlands.

Kyrou, K., Hammond, A. M., Galizi, R., Kranjc, N., Burt, A., Beaghton, A. K., Nolan, T., and Crisanti, A. (2018). A CRISPR-Cas9 gene drive targeting doublesex causes complete population suppression in caged Anopheles gambiae mosquitoes. *Nature Biotechnology*.

Lambert, B., North, A., Burt, A., and Godfray, H. C. J. (2018). The use of driving endonuclease genes to suppress mosquito vectors of malaria in temporally variable environments. *Malaria Journal*, 17(1):154.

Lyttle, T. W. (1977). Experimental population genetics of meiotic drive systems. I. Pseudo-Y chromosomal drive as a means of eliminating cage populations of Drosophila melanogaster. *Genetics*, 86(2 Pt. 1):413–445.

Martin, G., Aguilée, R., Ramsayer, J., Kaltz, O., and Ronce, O. (2013). The probability of evolutionary rescue: towards a quantitative comparison between theory and evolution experiments. *Philosophical Transactions of the Royal Society of London. Series B, Biological Sciences*, 368(1610):20120088.

Neve, P., Busi, R., Renton, M., and Vila-Aiub, M. M. (2014). Expanding the eco-evolutionary context of herbicide resistance research. *Pest Management Science*, 70(9):1385–1393.

Noble, C., Olejarz, J., Esvelt, K. M., Church, G. M., and Nowak, M. A. (2017). Evolutionary dynamics of CRISPR gene drives. *Science Advances*, 3(4):e1601964.

North, A., Burt, A., Godfray, H. C. J., and Buckley, Y. (2013). Modelling the spatial spread of a homing endonuclease gene in a mosquito population. *The Journal of Applied Ecology*, 50(5):1216–1225.

North, A. R., Burt, A., and Godfray, H. C. J. (2019). Modelling the potential of genetic control of malaria mosquitoes at national scale. *BMC biology*, 17(1):26.

Okamoto, K. W., Robert, M. A., Gould, F., and Lloyd, A. L. (2014). Feasible introgression of an anti-pathogen transgene into an urban mosquito population without using gene-drive. *PLoS neglected tropical diseases*, 8(7):e2827.

Prout, T. (1953). Some effects of variations in the segregation ratio and of selection on the frequency of alleles under random mating. *Acta Genetica Et Statistica Medica*, 4(2-3):148–151.

Report, N. A. (2016). Gene drive research in non-human organisms: Recommendations for responsible conduct.

Sandler, L. and Novitski, E. (1957). Meotic drive as an evolutionary force. *The American Naturalist*, 91(857):105–110.

Tabashnik, B. E. and Gould, F. (2012). Delaying corn rootworm resistance to Bt corn. *Journal of Economic Entomology*, 105(3):767–776.

Tanaka, H., Stone, H. A., and Nelson, D. R. (2017). Spatial gene drives and pushed genetic waves. *Proceedings of the National Academy of Sciences of the United States of America*, 114(32):8452–8457.

Uecker, H., Otto, S. P., and Hermisson, J. (2014). Evolutionary rescue in structured populations. *The American Naturalist*, 183(1):E17–35.

508 Unckless, R. L., Clark, A. G., and Messer, P. W. (2017). Evolution of Resistance Against CRISPR/Cas9
509   Gene Drive. *Genetics*, 205(2):827–841.

Figure 3

Figure 3. Parasites even partially resistant to gene drive cargo are favored if the cost to cargo-free fecundity is not too severe. Each panel represents evolution of wild-type versus mutant alleles under a the set of parameter values given in the title; wild-type fecundity in cargo-bearing hosts (b_{01}) is zero in all panels. Each dot represents a different mutant allele whose fecundities (b_{10}, b_{11}) are given by its coordinates. Green indicates that the wild-type was in a strong majority at the end of the trial, blue that the mutant was in a strong majority, and an intermediate color indicates that both alleles were moderately common. In the absence of competition from the mutant, the wild-type is capable of persisting for all conditions tested; its growth rate when rare is given as λ_0 in the panel, from eqn (9). Trials were run for 1000 generations in which both alleles started equally abundant in both patch types. Carrying capacity was 10^6 for all trials shown here.