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3 Abstract

Through their interactions, their activities, and even their mere presence, organisms change the environment for themselves and others. This "niche construction" process becomes particularly interesting when it creates evolutionary feedback, whereby selective pressures are altered in response to environmental change. Here we consider how niche construction influences the evolution of cooperation, which has been a longstanding challenge to evolutionary theory. We simulate populations of individuals that cooperatively produce a public good that permits increased growth in a stressful environment and investigate how local- and global-scale niche construction affects the ability of these populations to resist invasion by non-producing cheats. We find that niche construction profoundly impacts the evolution of cooperation by creating new opportunities for adaptation. Cooperators are able to escape subversion by cheats as long as niche construction clears these paths of adaptation. This work provides a crucial step towards understanding how evolution occurs in complex environments like those found in nature.

#### 20 Introduction

- Cooperative behaviors are common across all branches of the tree of life. Insects divide labor within their colonies, plants and soil bacteria exchange essential nutrients, birds care for others' young, and the trillions of cells in the human body restrain their growth and coordinate to provide vital functions. Each instance of cooperation presents an evolutionary challenge: How can individuals that sacrifice their own well-being to help others avoid subversion by those that do not? Over time, we would expect these defectors to rise in abundance at the expense of others, eventually driving cooperators—and perhaps the entire population—to extinction. Several factors can defer this potential tragedy of the commons (Hamilton, 1964; Hardin, 1968; Nowak, 2006; West et al., 2007b). For example, cooperators must benefit more from the cooperative act than others. This can occur when cooperators are clustered together in spatially-structured populations 33 (Fletcher and Doebeli, 2009; Nadell et al., 2010; Kuzdzal-Fick et al., 2011) or when cooperators use communication (Brown and Johnstone, 2001; Darch et al., 2012) or other cues (Sinervo et al., 2006; Gardner and West, 2010; Veelders et al., 2010) to cooperate conditionally with kin. Interestingly, cooperation can also be bolstered by genetic linkage with self-benefitting traits (Foster et al., 2004; Dandekar et al., 2012; Asfahl et al., 2015), setting the stage for an "adaptive race" in which cooperators and defectors vie for the first highly-beneficial non-social adaptation (Waite and Shou, 2012; Morgan et al., 2012).
- Using a model of public goods production, Hankshaw and Kerr (2015) recently

showed that in spatially-structured populations, cooperators gain a substantial leg up on defectors in an adaptive race when the cooperative behavior increases local population density, thus increasing the likelihood of acquiring beneficial non-social mutations. By hitchhiking along with these adaptations, cooperators can rapidly become more abundant. Nevertheless, this advantage is fleeting. As soon as the opportunities for adaptation are exhausted, cooperators are once again at a disadvantage against defectors. However, cooperation can be maintained indefinitely when frequent periodic environmental changes produce a stream of non-social adaptive opportunities. Although natural organisms typically find themselves in changing environments, cooperators may not be able to rely on the the environment to provide sufficient adaptive opportunities for their long-term survival. Here, we extend the model presented in Hankshaw and Kerr (2015) to explore whether niche construction feedbacks can act as a continual source of adaptive opportunity that allows cooperation to persist. Populations alter their local environment based on the presence of different non-social adaptations. Frequency-dependent selection on these adaptations creates an ecoevolutionary feedback, which increasingly favors the adaptations present in each population. This has several potential benefits for cooperators. First, because populations of cooperators are larger, they are more likely to gain

niche at a higher rate than defectors. Finally, because each population con-

mutations that enable them to track their changing environment. Similarly,

this difference in size means that larger cooperator populations "export" their

6 structs a unique niche, the threat of immigrant defectors may be diminished

by maladaptation to a cooperator population's niche.

By their very nature, public goods benefit populations by making their environment more hospitable. For example, bacteria produce extracellular products that find soluble iron (Griffin et al., 2004), digest large proteins (Diggle et al., 2007; Darch et al., 2012), and reduce the risk of predation (Cosson et al., 2002), 71 among many others (West et al., 2007a). While many studies have explored how the environment affects the evolution of cooperative bahviors, relatively few have examined how those behaviors affect the environment and the resulting change in evolutionary trajectories. Lehmann (2007) demonstrated analyti-75 cally that when niche construction act benefits future generations, cooperation 76 is favored due to reduced competition among kin. When rate-benefitting and yield-benefitting altruistic acts co-evolve, Van Dyken and Wade (2012) showed that "reciprocal niche construction", where the feedbacks produced by one act benefitted the other, can lead to increased selection for both traits. While 80 these studies have focused on the niche constructing effects of cooperation, we instead focus our attention here on how niche construction enables cooperators to escape defection by hitchhiking along with non-social traits.

#### Materials and Methods

We develop a computational model to observe the evolution of public goods cooperation in a spatially-structured metapopulation of populations. As described below, populations grow to carrying capacity, mutate, and migrate to neighboring patches. During this process, populations adapt to their local environments. The environments are, in turn, modified by the presence of these adapted individuals, allowing each population to construct a unique niche along its evolutionary trajectory. Model parameters and their values are listed in Table 1.

#### 93 Individuals and Fitness

Each individual has a genotype of length L+1. A binary allele at the first locus determines whether or not the individual is a cooperator, which carries cost c. Note that we refer to all individuals with allelic state 0 at this locus as a "defector", regardless of origin. The remaining L loci are  $stress\ loci$ , and are each occupied by a 0 or an integer from the set  $A = \{1, \dots, a_{max}\}$ , where  $a_{max}$ is the number of possible alleles. These alleles represent adaptations to the 99 environment, and the number of loci determines the number of possible adap-100 tations. All non-zero alleles carry fitness benefit  $\delta$ . Organisms also influence 101 their environment, which can feed back to influence selection. We model this 102 as a form of frequency dependent selection. Specifically, the selective value 103 of stress allele a at locus i increases with the proportion of the population 104 that has allele a-1 (modulo  $a_{max}$ ) at locus i-1. The slope of this increase 105 is  $\epsilon$  (which gauges the intensity of niche construction). As a consequence of 106 this form of frequency dependence, genotypes with sequentially increasing al-107 lelic states will tend to evolve. Because mutations are random, as described 108 below, each population will evolve sequences that start with different allelic 109 states. These different sequences represent the unique niches constructed by 110

populations. Under this model, the fitness of an individual with genotype g is:

$$W_g = z + a_{g,1}c + \delta \sum_{l=2}^{L+1} I_A(a_{g,l}) + \epsilon \sum_{h=1}^{N} I_{a_{h,1}}(a_{g,1}) + \epsilon \sum_{l=2}^{L} n(a_{g,l})$$

where  $a_{g,l}$  represents the allelic state of genotype g at locus l, z is a baseline fitness, L is the number of stress loci, N is the population size at that patch, and c is the cost of the cooperative allele.  $I_x(y)$  indicates whether the allelic state y matches allelic state x (1) or not (0).  $n(a_{g,l})$  is the number of individuals in the population with allelic state at the previous locus equal to one less than that at the focal locus  $a_{g,l}$ , or:

$$n(a_{g,l}) = \sum_{h=1}^{N} I_{a_{g,l}} (1 + a_{h,l-1} (\bmod a_{max}))$$

## 19 Population Growth

If p is the proportion of cooperators in a population at the beginning of a growth cycle, then that population reaches the following size during the growth phase:

$$S(p) = S_{min} + p(S_{max} - S_{min})$$

Therefore, a population composed entirely of defectors reaches size  $S_{min}$ , while one composed entirely of cooperators reaches size  $S_{max}$  (with  $S_{max} \geq S_{min}$ ).

The function S(p) gauges the benefit of public good production, as population size increases linearly with the proportion of cooperators. During growth, competition occurs. Consider an arbitrary genotype g. Let  $n_g$  be the number of individuals with genotype g, and let  $W_g$  be the fitness of genotype g(see equation [1]). The composition of genotypes after population growth is multinomial with parameters S(p) and  $\{\pi_1, \pi_2, \dots, \pi_{|G|}\}$ , where:

$$\pi_g = \frac{n_g W_g}{\sum_{i=1}^G n_i W_i}$$

Thus,  $\pi_g$  is the probability that an individual in the population after growth has genotype g (such that  $\sum \pi_g = 1$ ). G represents the set of all  $(a_{max} + 1)^L$  genotypes.

#### 134 Mutation

For simplicity, we apply mutation after population growth. These mutations occur independently at each locus and result in an allelic state change. At the binary cooperation locus, mutations flip the allelic state at rate  $\mu_c$ , causing cooperators to become defectors and vice versa. Mutations at a stress locus cause a new allelic state to be chosen at random from the set  $\{0\} \cup A$ . These mutation occur at each stress locus at rate  $\mu_s$ . Therefore, the probability that genotype q mutates into genotype q' is given by:

$$\tau_{g \to g'} = \mu_s^{H_s(g,\ g')} (1 - \mu_s)^{\{L - H_s(g,\ g')\}} \mu_c^{H_p(g,\ g')} (1 - \mu_c)^{\{1 - H_p(g,\ g')\}}$$

where  $H_s(g, g')$  and  $H_p(g, g')$  are the Hamming distances between genotypes g and g' at the stress loci and cooperation locus, respectively. The Hamming distance is the number of loci at which allelic states differ. Because there is no inherent relationship among alleles, each of the  $a_{max} + 1$  alleles is equally likely to arise via mutation at a given locus.

#### 147 Migration and Metapopulation Structure

The metapopulation consists of  $N^2$  patches arranged in a  $N \times N$  lattice. After mutation, individuals emigrate to an adjacent patch with probability m. This adjacent patch is randomly chosen with uniform probability from the source patch's Moore neighborhood, which is composed of the nearest 8 patches on the lattice. Because the metapopulation lattice has boundaries, patches located on an edge have smaller neighborhoods.

#### 154 Initialization and Simulation

Metapopulations are initiated in a state that follows the onset of an environ-155 mental stress. First, populations are seeded at each patch with cooperator 156 proportion  $p_0$  and grown to density  $S(p_0)$ . Stress is then introduced by sub-157 jecting the population to a bottleneck. The number of survivors with each 158 genotype q is sampled from a binomial distribution, where the number of tri-159 als is  $n_g$ . The probability of success is  $\mu_t$ , which represents the likelihood that 160 a mutation occurs that enables survival. Because individuals have not yet 161 adapted to this new stress, the allelic state of each genotype is set to 0 at each 162

stress locus ( $\forall g \in G, l \in \{2, ..., L+1\}$ :  $a_{g,l} = 0$ ). Following initialization, simulations are run for T cycles, where each cycle consists of growth, mutation, and migration. After migration, populations are thinned to allow for growth in the next cycle. The number of survivors for each genotype g is sampled from a binomial distribution, where the number of trials is  $n_g$  and the probability of success is d.

#### 169 Source Code and Software Environment

The simulation software and configurations for the experiments reported are available online (Us, 2015). Simulations used Python 3.4.0, NumPy 1.9.1, Pandas 0.15.2 (McKinney, 2010), NetworkX 1.9.1 (Hagberg *et al.*, 2008). Data analyses were performed with R 3.1.3 (R Core Team, 2015).

#### 174 Results

175 results...

### 76 Discussion

- summary of results
- similarities/differences from previous work
- Schwilk and Kerr (2002)

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• future primacy/recency
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- Laland et al. (1996)
- Lehmann (2007)
- public goods as niche construction
- Host symbiont many instances of cooperation occur among pathogens.
- $\bullet$  future QS or other environmental sensing
- Facultative cooperation
- Rodrigues (2012)
- Dumas and Kümmerli (2010)
- Kümmerli and Brown (2010)
- Darch/Diggle
- QS?
- Environmental Sensing?
- 193 Niche construction and selective feedbacks
- $^{194}$  Niche construction and other social interactions # Acknowledgments
- PRFB
- BEACON
- Google
- Organizers?

# $_{199}$ Figures

## 200 Tables

Table 1: Model parameters and their value.

Parameter	Description	Base Value
$\overline{L}$	Number of Stress Loci	3
$a_{max}$	Number of alleles	4
δ	Fitness benefit, nonzero alleles	0.5
$\epsilon$	Fitness benefit, sequential alleles	TODO
c	Production Cost	0.1
z	Baseline fitness	1
$S_{min}$	Minimum Population Size	800
$S_{max}$	Maximum Population Size	2000
$\mu_s$	Mutation Rate (Stress)	$10^{-5}$
$\mu_c$	Mutation Rate (Cooperation)	$10^{-5}$
$\mu_t$	Mutation Rate (Tolerance to New Stress)	$10^{-5}$
m	Migration Rate	0.05
d	Population Dilution Factor	0.1
$N^2$	Number of Metapopulation Sites	625
$p_0$	Initial Cooperator Proportion	0.5
<i>T</i>	Number of Simulation Cycles	TODO

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