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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 adaptation (Waite and Shou, 2012; Morgan et al., 2012).
- 42 Hankshaw and Kerr (2015) recently showed that in spatially-structured popula-

tions, cooperators can 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, the cooperative trait can rapidly rise in abundance. Nevertheless, this advantage is fleeting. As soon as the opportunities for adaptation are exhausted, cooperators are once again at a disadvantage against defectors. However, Hankshaw and Kerr (2015) demonstrated that cooperation can be maintained indefinitely when frequent environmental changes produce a steady stream of non-social adaptive opportunities. Although organisms typically find themselves in dynamic environments, change might not occur at a rate that provides sufficient adaptive opportunities to ensure longterm cooperator survival. In this work, we demonstrate how cooperation can be maintained indefinitely by niche construction. We expand upon the model presented in Hankshaw and Kerr (2015) to allow populations to alter their local environment. As environments change, so too does selection. This creates an eco-evolutionary feedback whereby selection is altered based on current phenotypes, which changes the composition of phenotypes and their effects. The effect on selection can be either positive or negative, depending on whether the environmental change increases or decreases the fitness of the niche-constructing individual. Although niche construction occurs independently of cooperation in our model, the increase in density that results from public good production has a profound effect on how populations evolve in the presence of selective feedbacks. First, these populations exert greater influence on their environments, which allows them

- to benefit more from positive niche construction. And as environments change,
- either by negative niche construction or external influences, these larger pop-
- 69 ulations can adapt more quickly. We show that it is the combination of these
- <sub>70</sub> factors that allows cooperation to persist.

#### 71 Description of positive NC effects

Include Smax and m sweeps here, or in a later paragraph?

#### 73 Description of negative NC effects

74 TODO

#### 75 Stuff to be cut/integrated above

- 76 This niche construction creates feedback that increasingly favors the adapta-
- tions present in each population. We first examine how the intensity of these
- <sup>78</sup> feedbacks affects the hitchhiking process. Because the production of public
- 79 goods increases population density, populations containing cooperators will
- 80 exert a greater influence on their environment. As a result, these environ-
- ments will be more quickly brought to states where fitness is higher.
- As populations construct unique niches, they potentially decrease the threat
- of invasion from neighboring patches. This occurs when the traits that were
- 84 advantageous in an immigrant's home niche are maladaptive elsewhere. Be-
- 85 cause environmental change is influenced solely by non-social phenotypes in
- this model, this change of invasibility affects cooperators and defectors equally.

Here again, however, populations containing a greater number of cooperators may have an advantage. The greater number of individuals that emigrate from these larger populations allow them to "export" their niche—and thus reduce the fitness of neighboring competitors—at a higher rate. We explore whether the range expansion that this process enables provides additional opportunities for cooperation to hitchhike. Finally, we demonstrate how *negative* niche construction, where populations change their environment in ways that reduce fitness, can further support cooperation. Even though the niche construction process creates selective feedbacks, we would expect the magnitude of these feedbacks to decrease as populations evolve. Once individuals can no longer gain adaptations that compensate for the costs of cooperation, they are then outcompeted by noncooperators. However if populations construct their environment in a way 99 which decreases fitness, cooperation can still hitchhike when this change also 100

## <sub>2</sub> Materials and Methods

101

We build upon the model described in Hankshaw and Kerr (2015), in which cooperators and defectors compete and evolve in a spatially-structured metapopulation of populations. Each of these populations grows to carrying capacity,
mutates, and migrates to neighboring patches. During this process, populations adapt to their local environments. In our extended model, we allow
populations to modify their local environment, and these modifications feed

creates the opportunity to gain compensatory adaptations.

back to affect selection.

#### 10 Model Description

Our simulated environment consists of  $N^2$  patches arranged as an  $N \times N$ 111 lattice (see Table 1 for model parameters and their values), where each patch 112 supports a population of zero or more individuals. Each individual in the 113 population has a genotype, which is an ordered list of L + 1 integers (loci). 114 The first L loci are adaptive loci, and are each occupied by a 0 or an integer 115 from the set  $A = \{1, \ldots, a_{max}\}$ , where  $a_{max}$  is the number of potential alleles. 116 At each of these loci, the presence of a non-zero allele represents an adaptation 117 to the environment that confers fitness benefit  $\delta$ . A binary allele at locus L+1118 determines whether or not that individual is a cooperator. Individuals with 119 allelic state 1 at this locus are cooperators, carrying a cost c, while individuals 120 with allelic state 0 are defectors. When  $\delta \geq c$ , a minimally-adapted cooperator 121 recoups the cost of cooperation. Equation 1 defines function n(a, l), which 122 gives the number of individuals in the population with allelic state a at locus 123 l.  $I_x(y)$  indicates whether the allelic state y matches allelic state x (1) or not (0), and  $\gamma(i)$  is the genotype of individual i.

$$n(a,l) = \sum_{i \in P} I_{a_{g,l}}(a_{\gamma(i),l}) \tag{1}$$

Organisms also influence their environment, which, in turn, influences selection. We model this as a form of frequency dependent selection. Specifically, the selective value of adaptive allele a at locus l increases with the number

of individuals in the population that have allele a-1 at locus l-1 (note that we treat both adaptive loci and allelic states as circular, so the allelic 130 state at locus 1 is affected by locus L, and allele 1 is best preceded by allele 131  $a_{max}$ ). The slope of this increase is  $\epsilon$ , which specifies the intensity of niche con-132 struction. As a consequence of this form of frequency dependence, genotypes 133 with sequentially-increasing allelic states will tend to evolve. Because muta-134 tions are random, as described later, each population will evolve sequences 135 that start with different allelic states. These different sequences represent the 136 unique niches constructed by populations. Under this model, the fitness of an 137 individual with genotype q in population P is:

$$W_g = z + \delta \sum_{l=1}^{L} I_A(a_{g,l}) + \epsilon \sum_{l=1}^{L} n(\beta(a_{g,l}, a_{max}), \beta(l, L)) - ca_{g,L+1}$$
 (2)

where z is a baseline fitness, and  $a_{g,l}$  represents the allelic state of genotype g at locus l. The function  $I_A(a)$  indicates whether a given allelic state a is in A (i.e., it is non-zero), while the function  $\beta(x, x_{max})$  gives the value below some value x in the circular set  $\{1, \ldots, x_{max}\}$  (Equation 3).

$$\beta(x, x_{max}) = \{(x - 2 + x_{max}) \bmod x_{max}\} + 1$$
 (3)

Cooperators produce a public good that is equally accessible to all members of the population. This public good increases the carrying capacity at that patch, allowing the population to reach greater density. This benefit increases linearly with the proportion of cooperators. Thus, 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:

$$S(p) = S_{min} + p(S_{max} - S_{min}) \tag{4}$$

The function S(p) reflects the benefit of public good production. A population composed entirely of defectors reaches size  $S_{min}$ , while one composed entirely of cooperators reaches size  $S_{max}$  (with  $S_{max} \geq S_{min}$ ). During growth, individuals compete for inclusion in the resulting population. The composition of population P with cooperator proportion p after growth is multinomial with parameters and S(p) and  $\{\pi_1, \pi_2, \ldots, \pi_{|P|}\}$ , where:

$$\pi_i = \frac{W_{\gamma(i)}}{\sum_{j \in P} W_{\gamma(j)}} \tag{5}$$

2).  $\pi_i$  therefore reflects that an individual's ability to persist is proportional 156 to its fitness relative to others'. 157 For simplicity, we apply mutations after population growth. Mutations occur 158 independently at each locus and cause the allelic state to change. Mutations 159 occur at each adaptive locus at rate  $\mu_a$ , and cause a new allelic state to be 160 chosen at random from the set  $\{0\} \cup A$ . At the binary cooperation locus, 161 mutations occur at rate  $\mu_c$ . These mutations flip the allelic state, causing 162 cooperators to become defectors and vice versa. Therefore, the probability 163 that genotype g mutates into genotype g' is given by:

Here,  $W_{\gamma(i)}$  is the fitness of an individual i with genotype  $\gamma(i)$  (see Equation

$$\tau_{g \to g'} = \mu_a^{H_a(g, g')} (1 - \mu_a)^{\{L - H_a(g, g')\}} \mu_c^{H_c(g, g')} (1 - \mu_c)^{\{1 - H_c(g, g')\}}$$
(6)

where  $H_a(g, g')$  and  $H_c(g, g')$  are the Hamming distances between genotypes gand g' at the cooperation locus and adaptive loci, respectively. The Hamming 166 distance is the number of loci at which allelic states differ (Hamming, 1950). 167 Because we define no inherent relationship among alleles, each of the  $a_{max} + 1$ 168 allelic states is equally likely to arise via mutation at a given locus. 169 After mutation, individuals emigrate to an adjacent patch at rate m. The 170 destination patch is randomly chosen with uniform probability from the source 171 patch's Moore neighborhood, which is composed of the nearest 8 patches on the 172 lattice. Because the metapopulation lattice has boundaries, patches located 173 on an edge have smaller neighborhoods. 174 Metapopulations are initiated in a state that follows an environmental change. 175 First, populations are seeded at all patches with cooperator proportion  $p_0$  and 176 grown to density  $S(p_0)$ . An environmental challenge is then introduced, which 177 subjects the population to a bottleneck. For each individual, the probability 178 of survival is  $\mu_t$ , which represents the likelihood that a mutation occurs that 179 confers tolerance. Survivors are chosen by binomial sampling. Because indi-180 viduals have not yet adapted to this new environment, the allelic state of each 181 individual's genotype is set to 0 at each adaptive locus  $(\forall i \in P, l \in \{1, ..., L\})$ : 182  $a_{\gamma(i),l}=0$ ). Following initialization, simulations are run for T cycles, where 183 each discrete cycle consists of growth, mutation, and migration. At the end 184 of each cycle, populations are thinned to allow for growth in the next cycle.

185

The individuals that remain are chosen by binomial sampling, where each individual persists with probability d, regardless of allelic state.

#### Source Code and Software Environment

189 The simulation software and configurations for the experiments reported are

available online (Us, 2015). Simulations used Python 3.4.0, NumPy 1.9.1,

<sup>191</sup> Pandas 0.15.2 (McKinney, 2010), and NetworkX 1.9.1 (Hagberg et al., 2008).

Data analyses were performed with R 3.1.3 (R Core Team, 2015).

# 193 Results

#### Niche construction prolongs cooperator survival

 $^{195}$  "cooperator presence" greater in NC runs than with no-NC (GNH)

## Negative Niche construction helps

197 Runs with wraparound persist indefinitely, while those without do not.

#### $_{\scriptscriptstyle{198}}$ NC Prevents/Limits Defector invasion

199 NC Enables Cooperator Spread by "exporting" environ-

200 ment

#### Discussion

```
• summary of results
```

- similarities/differences from previous work
- Schwilk and Kerr (2002)
- -10.1073/pnas.0812644106
- public goods as niche construction
- future QS or other environmental sensing
- Facultative cooperation
- Rodrigues (2012)
- Dumas and Kümmerli (2010)
- Kümmerli and Brown (2010)
- Darch/Diggle
- QS?
- Environmental Sensing? (Koestler and Waters, 2014, Bernier et al. (2011))
- Negative Niche construction as a stragegy? would those that create this
  constant pressure (L=5, A=6) do better than those that do not (L=5,

A=5?

Niche construction and selective feedbacks Niche construction and other social interactions

#### $_{221}$ Public Goods

TODO: merge this in with the "Cooperative Niche Construction" section
TEST

By their very nature, public goods benefit populations by making their environ-224 ment more hospitable. For example, bacteria produce extracellular products 225 that find soluble iron (Griffin et al., 2004), digest large proteins (Diggle et al., 226 2007; Darch et al., 2012), and reduce the risk of predation (Cosson et al., 2002), 227 among many others (West et al., 2007a). While many studies have explored 228 how the environment affects the evolution of cooperative bahviors, relatively 229 few have examined how those behaviors affect the environment and the result-230 ing change in evolutionary trajectories. Lehmann (2007) demonstrated analyti-231 cally that when niche construction act benefits future generations, cooperation 232 is favored due to reduced competition among kin. When rate-benefitting and 233 yield-benefitting altruistic acts co-evolve, Van Dyken and Wade (2012) showed that "reciprocal niche construction", where the selective feedbacks produced 235 by one act benefitted the other, can lead to increased selection for both traits. 236 While these studies have focused on the niche constructing effects of cooperation, we instead focus our attention here on how niche construction enables 238 cooperators to escape defection by hitchhiking along with non-social traits.

#### 240 Primacy/Recency

In our model, alterations to the environment were immediately echoed by changes in selection. However, decoupling the timescales on which these processes occur can have substantial effects (Laland et al., 1996). By integrating past allelic states into Equation 2, we can begin to explore how the cumulative 244 effects of niche construction affect the creation of non-social adaptive opportunities and the benefits that they offer cooperation. Here, how these past 246 allelic states are integrated will play an important role. For example, when 247 the effects of earlier generations are weighted more heavily, the influence of 248 migration may be diminished. While this will reduce the threat of emigration 249 by defectors, cooperator populations will also be less effective at exporting 250 their niche. 251

### Cooperative Niche Construction

While our focus for this work has been on the eco-evolutionary feedbacks created by non-social traits, it would also be interesting to explore how this system 254 is affected by the timescale at which carrying capacity at a given patch is in-255 creased by public goods. In natural settings, a multitude of factors including 256 protein durability (Brown, 2007; Kümmerli and Brown, 2010), diffusion (Al-257 lison, 2005; Driscoll and Pepper, 2010), and resource availability (Zhang and 258 Rainey, 2013; Ghoul et al., 2014) influence both the rate and the degree to 259 which public goods alter the environment (and thereby selection). Lehmann 260 (2007) demonstrated that a cooperative, niche constructing behavior can be favored when it only affected selection for future generations, thus reducing the potential for competition among contemporary kin. The evolutionary inertia that this creates, however, may ultimately work against cooperators. When public good accumulates in the environment, cooperators must reduce their investment in production to remain competitive (Kümmerli and Brown, 2010).

TODO: wrap up. Facultative cooperation requires sensing.

### 268 Host-Symbiont

In many instances of cooperation, the environment is itself a biological entity, which can produce additional evolutionary feedbacks. As the host population 270 changes, so too will selection on their symbiont populations. Here, evolution-271 ary outcomes depend greatly on the degree of shared interest between the host 272 and symbiont. For example, the cooperative production of virulence factors by 273 the human pathogen P. aeruqinosa in lung infections is harmful to those with 274 cystic fibrosis (Harrison, 2007). Conversely, cooperative light production by A. 275 fischeri is vital for the survival of its host, the Hawaiian bobtail squid (Ruby, 276 1996). It was recently argued that incorporating the effects of niche construc-277 tion is critical for improving our understanding of viral evolution (Hamblin et 278 al., 2014) and evolution in co-infecting parasites (Hafer and Milinski, 2015). 279 Incorporating host dynamics, co-evolution, and the feedbacks that they pro-280 duce into models is likely to be equally important for gaining an understanding 281 of how cooperative behaviors evolve in these host-symbiont settings.

# Acknowledgments

• TODO: Organizers?

• TODO: lab comments

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# Figures Figures

- Figure 1
- Figure 1A

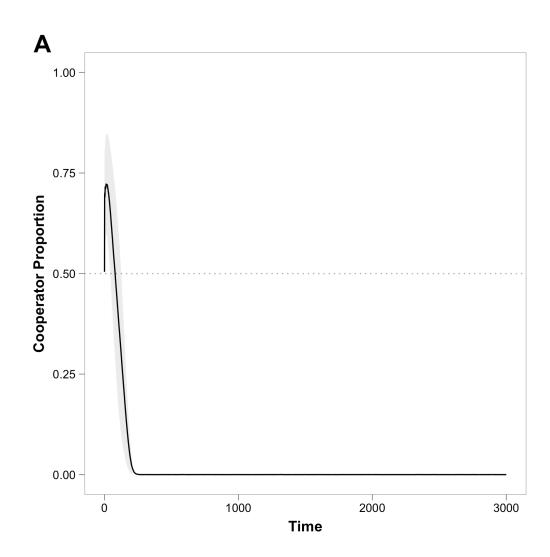


Figure 1: Proportion of cooperators over time when there are no opportunities for non-social adaptation

## Figure 1B

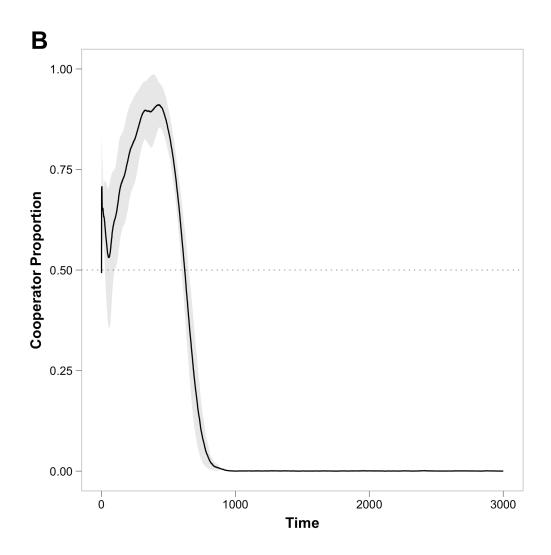


Figure 2: Proportion of cooperators over time with non-social adaptation (GNH)  $\,$ 

# Figure 1C

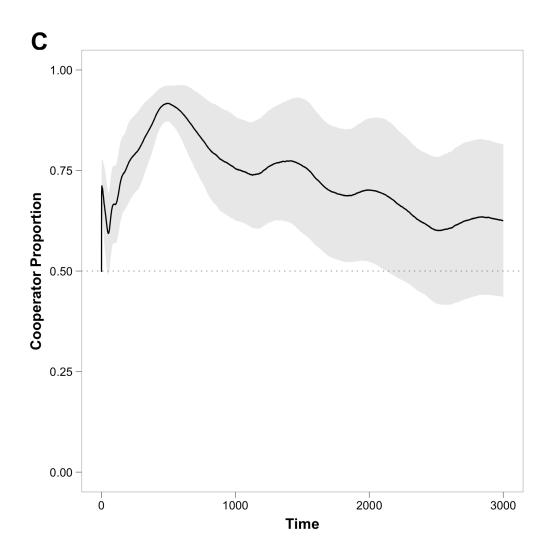


Figure 3: Proportion of cooperators over time with non-social adaptation and selective feedbacks  ${\bf r}$ 

# Figure 2

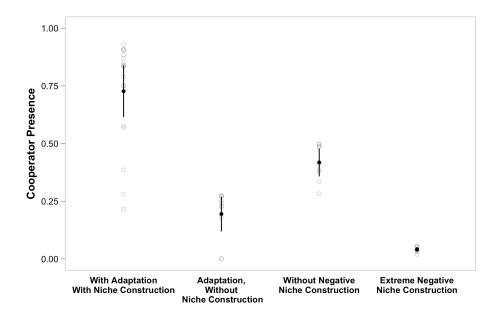


Figure 4: Cooperator Presence TODO

## Figure 3

Mean fitness over time for the treatments shown in Figure 2

#### <sup>302</sup> Figure 3A - Fitness for base case: niche construction

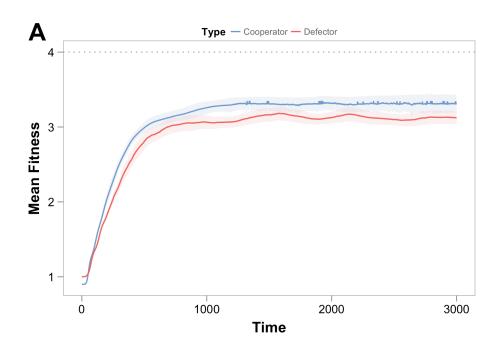


Figure 5: Grand mean Fitness of cooperators and defectors TODO

- Figure 3B Fitness with double delta, no epsilon
- Figure 3C Fitness with no negative niche construction (L=5, A=5)
- Figure 3D Fitness with extreme negative niche construction (L=1,  $^{306}$  A=6)

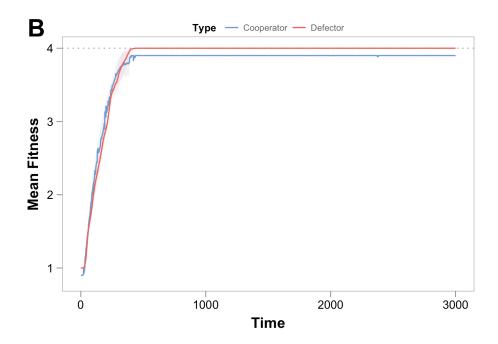


Figure 6: Grand mean Fitness of cooperators and defectors, double delta, no epsilon TODO

#### Figure 4

Cooperators invade from single population

# $_{309}$ Figure 5

310 Defectors are kept at bay

# Figure 6

Figure 6A - Effect of Migration Rate (m)

Figure 6B - Effect of Public Good Benefit (Smax-Smin)

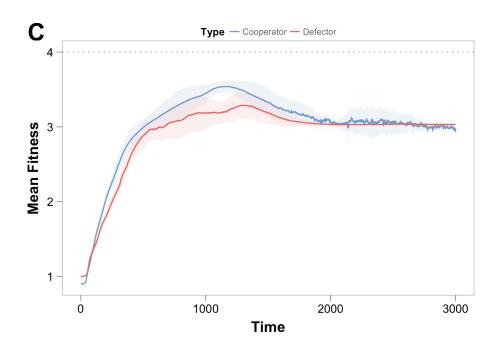


Figure 7: Grand mean Fitness of cooperators and defectors, no negative niche construction  $\operatorname{TODO}$ 

# Tables

Table 1: Model parameters and their value

Parameter	Description	Base Value
$N^2$	Number of metapopulation sites	625
L	Number of adaptive loci	5
$a_{max}$	Number of alleles	6
δ	Fitness benefit, nonzero alleles	0.3
c	Production cost	0.1
$\epsilon$	Fitness benefit, sequential alleles	0.00015
z	Baseline fitness	1

Parameter	Description	Base Value
$S_{min}$	Minimum population size	800
$S_{max}$	Maximum population size	2000
$\mu_a$	Mutation rate (adaptation)	$10^{-5}$
$\mu_c$	Mutation rate (cooperation)	$10^{-5}$
m	Migration rate	0.05
$p_0$	Initial cooperator proportion	0.5
$\mu_t$	Mutation rate (tolerance to new stress)	$10^{-5}$
T	Number of simulation cycles	1000
d	Population dilution factor	0.1

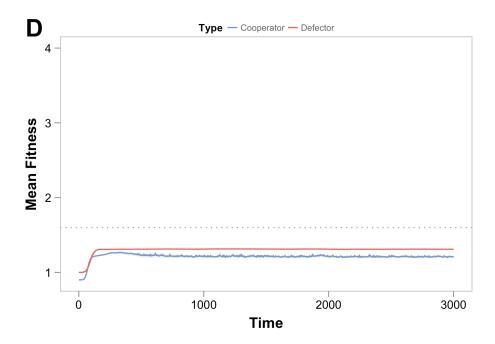


Figure 8: Grand mean Fitness of cooperators and defectors, extreme negative niche construction TODO

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