



## Seasonally Limited Host Supply Generates Microparasite Population Cycles

CHRISTOPHER J. DUGAW\*

Department of Mathematics,  
University of California Davis,  
One Shields Avenue,  
Davis,  
CA 95616,  
USA

*E-mail:* cjdugaw@ucdavis.edu

ALAN HASTINGS

Department of Environmental Science and Policy,  
University of California Davis,  
One Shields Avenue,  
Davis,  
CA 95616,  
USA

EVAN L. PREISSER

Graduate Group in Population Biology,  
University of California Davis,  
One Shields Avenue,  
Davis,  
CA 95616,  
USA

DONALD R. STRONG

Department of Evolution and Ecology and  
The Bodega Marine Laboratory,  
University of California Davis,  
One Shields Avenue,  
Davis,  
CA 95616,  
USA

1 Cycles in biological populations have been shown to arise from enemy–victim sys-  
2 tems, delayed density dependence, and maternal effects. In an initial effort to model  
3 the year-to-year dynamics of natural populations of entomopathogenic nematodes  
4 and their insect hosts, we find that a simple, nonlinear, mechanistic model pro-  
5 duces large amplitude, period two population cycles. The cycles are generated by

\*Author to whom correspondence should be addressed.

seasonal dynamics within semi-isolated populations independently of inter-annual feedback in host population numbers, which differs from previously studied mechanisms. The microparasites compete for a fixed number of host insect larvae. Many nematodes at the beginning of the year quickly eliminate the pool of small hosts, and few nematodes are produced for the subsequent year. Conversely, initially small nematode populations do not over-exploit the host population, so the surviving hosts grow to be large and produce many nematodes that survive to the following year.

© 2003 Published by Elsevier Ltd on behalf of Society for Mathematical Biology.

## 1. INTRODUCTION

Recent advances in the theory of nonlinear dynamics provide tools for viewing the complex behavior of biological populations (Hastings *et al.*, 1993; Hanski and Korpimäki, 1995; Higgins *et al.*, 1997; Grenfell *et al.*, 1998; Blasius *et al.*, 1999; Kendall *et al.*, 1999; Stenseth, 1999; Finkenstadt and Grenfell, 2000; Turchin and Ellner, 2000; Bjørnstad and Grenfell, 2001). Connections between the observed dynamics of natural populations (Higgins *et al.*, 1997) and ecological theory are developing rapidly, particularly in the areas of human epidemiology (Earn *et al.*, 2000) and the dynamics of small mammals (Stenseth, 1999). Cyclic behavior is perhaps the most well studied aspect of enemy–victim interactions, but the specific mechanisms producing these cycles have been the subject of debate (Kendall *et al.*, 1999). We model a common but little studied enemy–victim system where microparasite population cycles are driven by competition for the limited supply of host larvae within a single season.

Entomopathogenic nematodes are common, widespread enemies of soil insects (Hominick, 2002). Natural populations are extremely patchy in space (Stuart and Gaugler, 1994) and can cause extremely high mortalities to natural populations of hosts (Strong, 1999), suggesting local over-exploitation of hosts (Strong, 2002). Local populations of these enemies wax and wane over runs of 3–5 years (D.R.S. and E.L.P., unpublished). Immature, soil dwelling, ‘infective juveniles’ of the entomopathogenic nematode (*Heterorhabditis marelatus*) search in the soil for host ghost moth (*Hepialus californicus*) caterpillars that feed upon the taproot of bush lupine (*Lupinus arboreus*) (Strong *et al.*, 1996). They enter and kill the insect host, reproduce inside, and emerge into the soil in a pulse when the cadaver’s resources are exhausted. Small, young hosts yield few infective juveniles, while large old caterpillars yield upward of a million (E.L.P., unpublished data). This means that hosts escaping infection early in the season can grow into a much more productive resource for late season infection by the nematode. Infective juveniles cycle back into the non-reproductive, non-feeding population in the soil when they exit the cadaver. The cycle can repeat several times in the wet soil of a winter growing season. The environment of this interaction is the few liters of soil around the taproot of a lupine rhizosphere. Several meters separate most taproots from one

1 another in the grassland matrix where lupines grow. Both nematodes and caterpillars  
2 remain near the taproot, rates of dispersal are low, and populations are virtually  
3 restricted to single rhizospheres within a growing season. In dry summer soil the  
4 nematodes are quiescent, and host attack rates decrease to virtually nil.

5 The nematode is less frequent in space than the host ghost moth, and rhizospheres  
6 without the enemy produce most of the ghost moth recruits (Strong *et al.*, 1996).  
7 Ghost moths are very strong flyers, and can lay eggs many hundreds of meters  
8 away from their natal rhizosphere (Wagner, 1985). Their great powers of disper-  
9 sal supplies rhizospheres with host recruits independently of their local nematode  
10 densities.

11 To understand the dynamics of natural populations of these potent natural ene-  
12 mies, we developed the simplest model consistent with their biology: a continuous  
13 time mechanistic model describing the within-year dynamics of *H. marelatus* with  
14 ghost moth caterpillars in single lupine rhizospheres. The within-year model is  
15 then used to generate a discrete time model of between-year dynamics (Briggs and  
16 Godfray, 1996; Roberts and Heesterbeek, 1998; Gamarra and Sole, 2002) for nat-  
17 ural populations that exist in a highly seasonal environment. Fenton *et al.* (2000)  
18 modeled entomopathogenic nematode dynamics for biological control of glass-  
19 house insect pests with an approach featuring aseasonal dynamics and uniformly  
20 sized hosts, finding instability. In a second paper, they focused upon transient  
21 dynamics appropriate to short-term aseasonal biological control of host pests, and  
22 included host stage structure and a constant nematode development period (Fenton  
23 *et al.*, 2001). Using parameters derived from field and laboratory data, we model  
24 long-term seasonal dynamics of natural populations of nematodes with univoltine  
25 hosts, and include effects of host size and host immigration. While our model is  
26 based on the interactions between a particular microparasite and its host, it should  
27 be broadly applicable to organisms with seasonally limited resources.

## 28 2. MODEL

29 We modeled the population dynamics of nematodes and their hosts within a sin-  
30 gle rhizosphere. The infective juveniles search randomly through the wet, winter  
31 soil of the rhizosphere to find and infect hosts at a rate proportional to the product  
32 of the nematode and host densities (Strong *et al.*, 1999). In nature, most hosts are  
33 killed by a single hermaphroditic infective juvenile; thus, we model single infec-  
34 tions of hosts. Infection causes the death of the host within hours, removes a nema-  
35 tode from the soil-dwelling population, and adds an infected host cadaver.

36 Caterpillars hatch almost simultaneously and grow at similar rates; the host  
37 cohort is comprised of similar-sized caterpillars. A time-dependent (within the  
38 season) number  $\Lambda(t)$  of infective juvenile nematodes emerge simultaneously after  
39 a fixed period of time,  $\tau$ , from the host cadaver. Early in the growing season, hosts  
40 are small and produce few infective juveniles. By the end of the season, hosts are

large and can produce as many as a million infective juveniles. We denote the density of free-living, soil-dwelling nematodes by  $N$  and the density of hosts by  $H$ . Nematodes emerging from a cadaver at time  $t$  came from an infection event  $\tau$  time units earlier; emergence from cadavers increases the nematode density in the soil at the rate  $\beta H(t - \tau)N(t - \tau)\Lambda(t - \tau)$ , where  $\beta$  is infectivity, the rate at which nematodes find and infect hosts. The dynamics of the hosts are described by two terms: death from nematodes, and density-independent deaths at a rate  $k_H$ . The density of soil-dwelling nematodes decrease owing to entrance into hosts and to density-independent deaths of nematodes at a rate  $k_N$ . Density of soil-dwelling nematodes increases owing to emergence from hosts. The model for wet, growing season dynamics becomes

$$\begin{aligned} \frac{dH(t)}{dt} &= -\beta H(t)N(t) - k_H H(t) \\ \frac{dN(t)}{dt} &= -\beta H(t)N(t) - k_N N(t) + \beta H(t - \tau)N(t - \tau)\Lambda(t - \tau), \end{aligned} \quad (1)$$

which we solve over the period  $t = 0$  to  $T$ , where  $T$  is the (fixed) length of the wet season. During the dry season, when the nematodes are quiescent and do not infect hosts, we assume that nematodes experience only density-independent mortality. Nematodes that are inside host cadavers at the end of the dry season can remain in the cadaver for extended periods of time, where mortality rates are lower than in the soil (Kaya and Gaugler, 1993). Because of this, over-summer survival probabilities are higher for these nematodes. We assume that all of the nematodes have either emerged from hosts or died by the end of the dry season. A solution of equation (1) yields the density of nematodes outside  $N^k(T)$  of host cadavers at the end of wet season. The density of nematodes inside of hosts at the end of the wet season is  $N_i^k(T) = \int_{T-\tau}^T \beta H(\sigma)N(\sigma)\Lambda(\sigma)d\sigma$ . The dry-season submodel then yields the density of the nematodes at the beginning of the next wet season,  $N^{k+1}(0) = \lambda_o N^k(T) + \lambda_i N_i^k(T)$ , where  $\lambda_o$  and  $\lambda_i$  are, respectively, the survival probabilities of nematodes that began the dry season inside and outside of hosts. Returning to the wet season submodel [equations (1)], we study year-to-year dynamics by taking as initial conditions the resulting number of nematodes after the dry season  $N^{k+1}(0)$  and the assumed fixed number of host eggs per bush,  $H(0)$ . We assume that hosts emerge simultaneously so the initial interval for the hosts required by the delay term is uniformly zero. Because the only delay term is a proportional to the hosts, the initial nematode interval is not needed.

### 3. METHODS

We numerically integrated the wet-season delay differential equations using Euler's method, calculated the dry-season dynamics, and recorded the number of

Table 1. Estimates of infectivity,  $\beta$ , in different sized arenas from published sources and our data, measured in (rhizosphere volume/day =  $20 \times \pi 50^2 \text{ cm}^3/\text{day}$ ).

Nematode species	Host species	Soil vol. (cc)	Mean $\beta$ (rhiz. vol./day)	Reference
<i>Heterorhabditis</i> spp.	<i>G. mellonella</i>	25	$1.25 \times 10^{-5}$	Westerman (1998)
<i>Heterorhabditis</i> spp.	<i>Otiiorhynchus sulcatus</i>	25	$2.91 \times 10^{-6}$	Westerman (1998)
<i>H. bacteriophora</i>	<i>G. mellonella</i>	50	$1.26 \times 10^{-6}$	Campbell <i>et al.</i> (1999)
<i>Heterorhabditis</i> spp.	<i>G. mellonella</i>	15	$1.39 \times 10^{-4}$	Koppenhöfer <i>et al.</i> (1996)
<i>H. marelatus</i>	<i>G. mellonella</i>	10	$2.10 \times 10^{-6}$	Strong (unpublished data)
<i>H. marelatus</i>	<i>G. mellonella</i>	100	$7.33 \times 10^{-6}$	Strong (unpublished data)
<i>H. marelatus</i>	<i>G. mellonella</i>	1000	$2.50 \times 10^{-6}$	Strong (unpublished data)

1 hosts killed during each wet season. We tested the stability of the numerical method  
 2 by decreasing the time step size. To understand the full range of the dynamics, we  
 3 computed a bifurcation diagram numerically, by varying the infectivity,  $\beta$ , iterating  
 4 the return map for 1000 generations to eliminate transients, and then plotting the  
 5 next 100 points to describe the attractor as a function of the parameters. The com-  
 6 plexity of the year-to-year map makes any analytic treatment of bifurcation impos-  
 7 sible, but we numerically generated the time one return map from the continuous,  
 8 within-year dynamics in the regions of parameter space where the dynamics are  
 9 qualitatively different. We used several different initial conditions to assure that  
 10 all attractors had been found. We used  $\beta$  as the bifurcation parameter, because the  
 11 qualitative dynamics are most sensitive to this parameter (see Discussion).

12 To focus on the most relevant dynamical behavior, we estimated infectivity,  $\beta$ ,  
 13 from laboratory experiments (Table 1). We exactly solved equations (1), for  $\beta$   
 14 in terms of the initial density of nematodes  $N_0$ , the number of successful infect-  
 15 ing nematodes  $I$ , the time in days of the experiment  $T$ , and  $V_e/V_r$ , the ratio of  
 16 the experimental volume to the rhizosphere volume ( $V_e \approx 20 \times \pi 50^2 \text{ cm}^3$ ). We  
 17 solved the equations for short periods too brief for either nematode reproduction  
 18 or significant nematode mortality,  $k_N = 0$  and  $I = N_0 - N$ . Nematode mortality  
 19 is low in the pasteurized soil of experiments (Koppenhöfer *et al.*, 1996). Although  
 20 multiple infections are quite rare in nature, multiple infections are common in the  
 21 laboratory experiments for estimation of  $\beta$ ; thus, the number of hosts available for  
 22 infection,  $H$ , is constant. The resulting equation for infectivity,  $\beta$ , is

$$23 \quad \beta = \frac{1}{HT} \cdot \ln \left( \frac{N_0}{N_0 - I} \right) \cdot \frac{V_e}{V_r}. \quad (2)$$

24 Our estimates based on equation (2) and previous estimates of  $\beta$  are presented  
 25 with the appropriate units in Table 1. However, we believe these are overesti-  
 26 mates of natural values of this parameter (see the Discussion). Nematode mortality  
 27  $k_N$ , and host productivity growth rate  $r$  are taken from the literature Strong (2002),

Wagner (1985). The maximum productivity of nematodes from hosts,  $\Lambda_{\max}$ , equaled the mean productivity of host larvae near pupation. To determine the influence of within-season growth of hosts on the dynamics, we compared unvarying  $\Lambda$  with a saturating function in which productivity of nematodes increased as hosts grew in size, i.e.  $\Lambda(t) = \min(Ce^{rt}, \Lambda_{\max})$ . This function closely matches data on host growth (Wagner, 1985). To test the robustness of our results, several solutions were calculated with small variations in the parameters. A saturating type II functional response for nematode attack was used as well, but did not qualitatively change results.

#### 4. RESULTS

The qualitative result of our model is a series of large amplitude, period-two cycles of the nematode population, yielding high and very low numbers for a large range of parameter values (Fig. 1). The wide range over which cycles occur critically depends on the assumption of host growth during the season. Were hosts not to grow in size (i.e., constant  $\Lambda$ ) the high-low cycles would be restricted to a much smaller range of parameters. The cycles will persist for the low immigration rates [corresponding to results for equilibria of difference equations in Karlin and McGregor (1972)] that we have observed within seasons in the field. Host mortality in the model tracks the two-cycle nematode population dynamics, with almost 100% mortality for rhizospheres in which the initial nematode population is high at the beginning of the growing season and much lower mortality for rhizospheres in which the nematode population is initially low.

For very low values of nematode infectivity  $\beta$ , the nematodes become extinct. As  $\beta$  increases there is a nematode population steady state that quickly undergoes a series period doubling bifurcation, leading to chaos (Fig. 2). These initial bifurcations are explained by classic work dating back to May and Oster (1976). They showed that these dynamics are ‘generic to any curve with a hump whose steepness can be parametrically tuned’. Integrating the within year dynamics leads to a year-to-year map for the nematodes that has a single hump. The parameter  $\beta$  plays the role of the reproduction parameter in the Ricker and logistic maps by controlling the steepness (see Figs. 3 and 4). The complex dynamics quickly lock in on a period two cycle, because the hump in the discrete map moves to nematode population numbers below one. As  $\beta$  increases further the system returns to an equilibrium. This equilibrium persists, but approaches zero.

#### 5. DISCUSSION

Competition for hosts is the indirect mechanism of density dependence leading to cycles in nematode numbers in our model. Because hosts grow larger during the season, density dependence is stronger; elimination of small, relatively unproductive hosts early in the season reduces the availability of large, highly productive

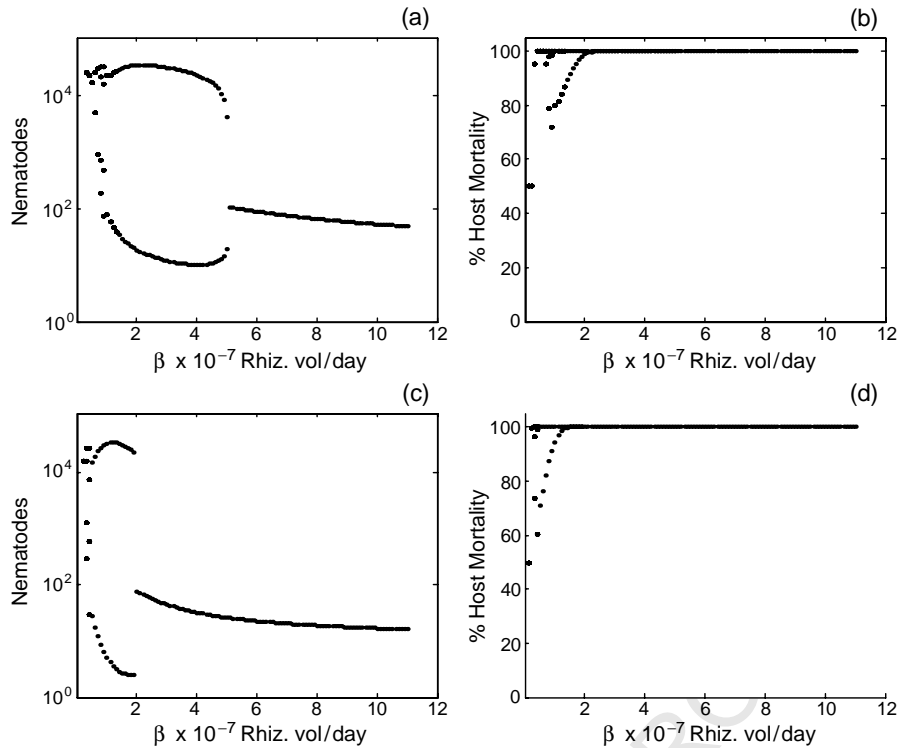


Figure 1. Model results showing dependence of nematode population and annual host mortality on nematode infectivity,  $\beta$ , and host growth. Panels (a) and (b) show nematode population at the start of the wet season, and total annual host mortality with host size included. Panels (c) and (d) are analogous, but with no host size effect. For both simulations host mortality  $k_H = 0.0001/\text{days}$ , nematode mortality  $k_N = 0.063/\text{days}$ , and nematode generation time  $\tau = 35$  days. The length of the wet season was 160 days, and dry-season survivorship for soil-dwelling nematodes was  $10^{-6}$  and  $10^{-3}$  for nematodes inside cadavers. In (a) and (b) maximal host productivity was  $\Lambda_{\max} = 800\,000$  nematodes per host,  $r = 0.09/\text{days}$ ,  $C = 10000$ , and in (c) and (d) host productivity was held constant at  $\Lambda = 800\,000$ . Notice that the cycles exist for a much smaller range of parameters.

1 hosts late in the season (Briggs *et al.*, 1999). Timing of infections plays a role  
 2 in the density dependence, because the earlier an infection occurs, the longer the  
 3 infective juveniles produced from the infection are exposed to high mortality in  
 4 the soil. This is why cycles can occur when there is no host growth or nearly  
 5 100% host mortality every year. Cycles found in previous models for hosts with  
 6 non-overlapping generations depend on an inter-annual feedback between host and  
 7 enemy populations (Mills and Getz, 1996). The cycles we find in the model are  
 8 different, because host recruitment in the following year is independent of local  
 9 interaction between enemy and victim. In our model we have assumed the time  
 10 for nematodes to emerge from an infected host,  $\tau$ , is constant. However, this time  
 11 delay is shorter for smaller hosts (E.L.P., unpublished data). A large nematode pop-  
 12 ulation early in the wet season will thus more quickly reduce the host population,



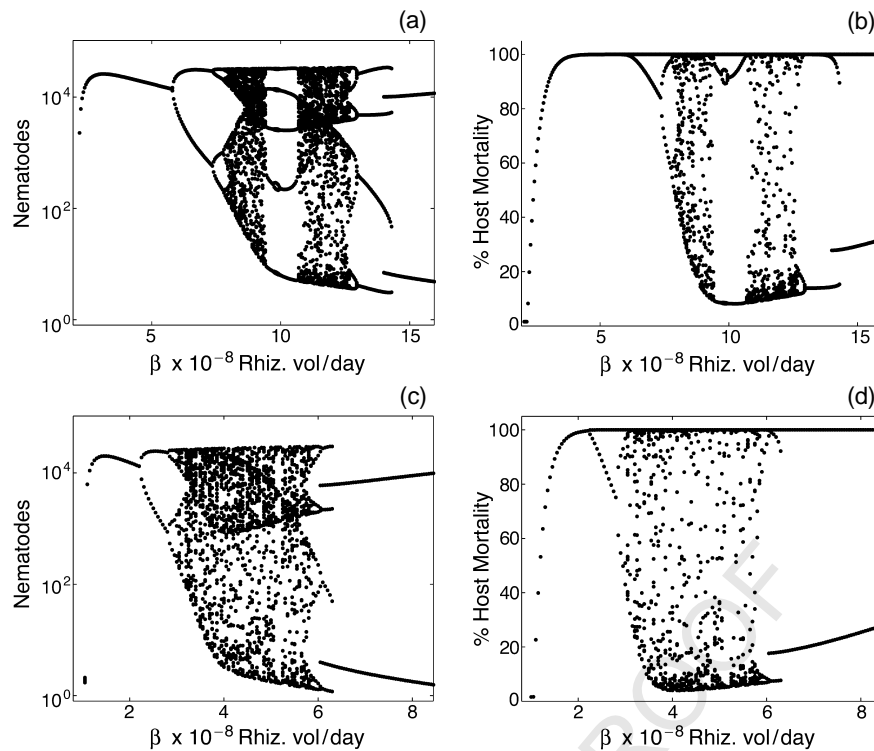


Figure 2. Detail of model results showing dependence of nematode population and annual host mortality on nematode infectivity ( $\beta$ ) and host growth for very small values of infectivity. Panels (a) and (b) show nematode population at the start of the wet season, and total annual host mortality with host size included. Panels (c) and (d) are analogous, but with no host size effect. Parameters are the same as in Fig. 1.

leading to a greater likelihood of cycles. These dynamics should apply to many enemy–victim systems in which subsidies are important to dynamics, such as those in which the feedback of predation to prey reproduction occurs on multiple scales; hosts are locally depleted while new hosts arrive from a larger population that is not exposed to the enemy (Polis and Strong, 1996).

Our model produces microparasite cycles over a range of values of infectivity,  $\beta$ , that is slightly lower than our and other estimates of this parameter. However, we believe with Fenton *et al.* (2000) that previous estimates are too high for natural settings. First, the natural host ghost moths are not as vulnerable to the nematodes as the wax worms (*Galleria mellonella*) used in most estimates of  $\beta$  (Strong *et al.*, 1996). Second, experiments have been conducted in small enclosed arenas of homogeneous soil that concentrate the waste gasses produced by hosts that nematodes use to find hosts. Consequently, more nematodes infect hosts, leading to artificially high estimates of  $\beta$ . Our preliminary work outside of containers suggests that realistic field estimates of  $\beta$  will prove to be lower than current laboratory estimates. Additionally, mathematical results suggest that stochasticity and

1  
2  
3  
4  
5  
6  
7  
8  
9  
10  
11  
12  
13  
14  
15  
16



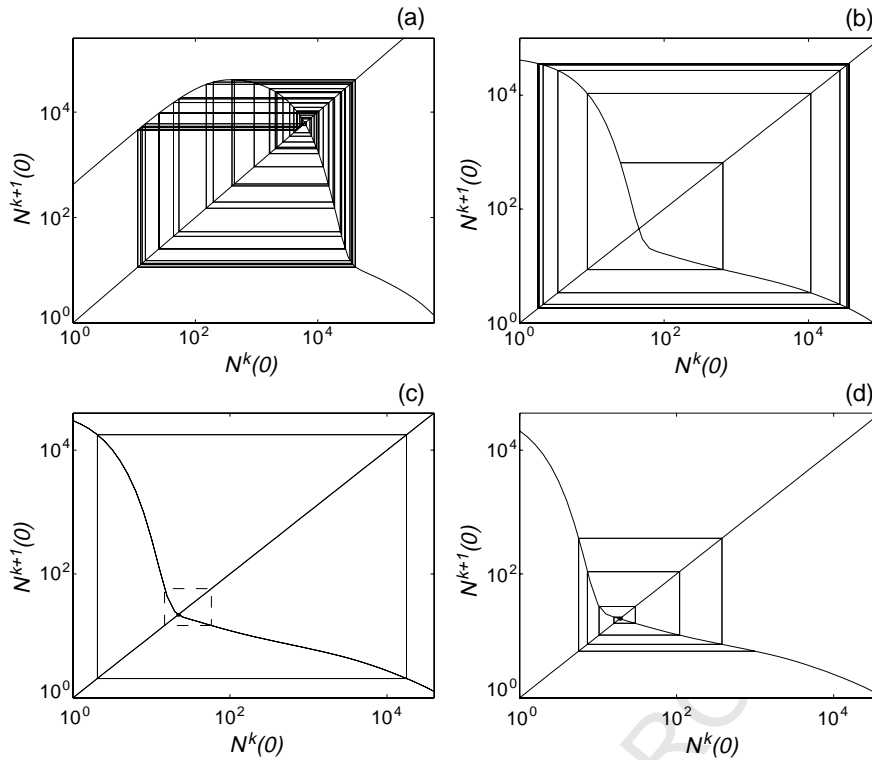


Figure 3. Map that takes the nematode population size at the start of the wet season to the nematode population size at the start of the wet season the following year, with varying host size. Maps are plotted on a log–log scale for clarity; therefore, the origin is excluded. Panel (a) shows the chaotic dynamics in nematode population with  $\beta = 8 \times 10^{-8}$ . Panel (b) shows the period two dynamics of map when  $\beta = 4.5 \times 10^{-7}$ . The chaotic dynamics are no longer present because the hump has moved to the range where the nematode population is less than one. Panel (c) shows the coexistence of a stable equilibrium and a stable two cycle (solid line); the unstable cycle is indicated with a dashed line,  $\beta = 4.9 \times 10^{-7}$ . These dynamics exist for a very narrow region parameter space when the  $N^{k+1}(0) = N^k(0)$  line crosses through the region of the map where the slope quickly changes from being greater in absolute value than 1 to being less than 1 in absolute value. Panel (d) shows the stable equilibrium dynamics with  $\beta = 6 \times 10^{-7}$ .

1 a lack of spatial homogeneity extend the range of parameters over which cycles  
2 occur [e.g. Billings and Schwartz (2002)].

3 Entomopathogenic nematodes persist in nature, while published models find their  
4 dynamics unstable. We present the simplest mechanistic model consistent with  
5 their biology in an attempt to understand the dynamics of the interaction with hosts  
6 of these widespread, important, but little known natural enemies. The extreme  
7 cycles that we found in the model appear in previous modeling of these enemies,  
8 which addressed their use in biological control of mushroom pests (Fenton *et al.*,  
9 2000). These cycles bode instability, and suggest that there may be high rates of  
10 local extinction in nature. The model cycles have directed our attention to mea-

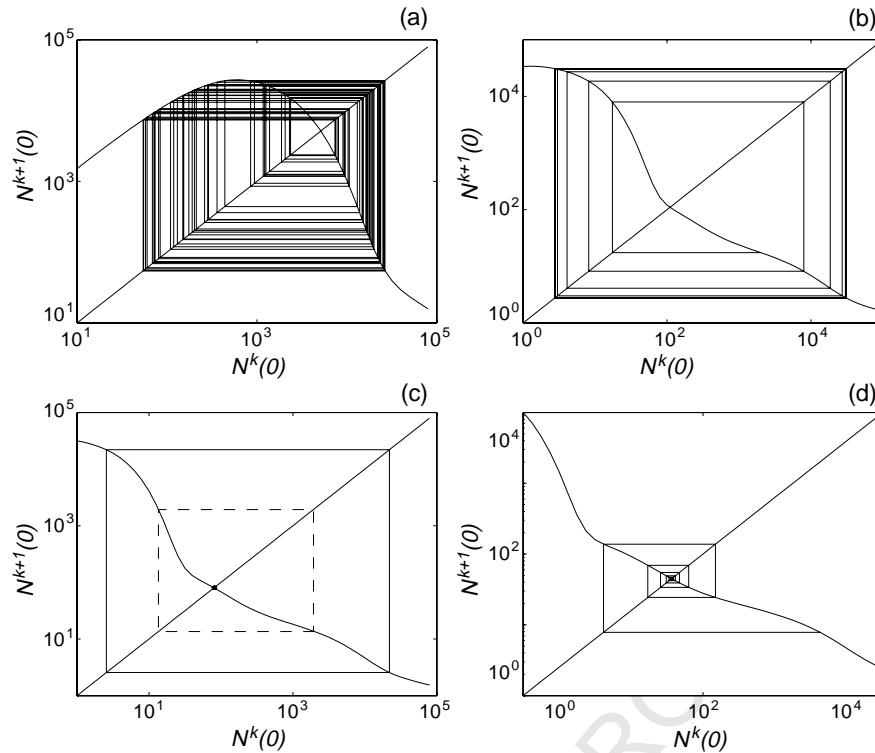


Figure 4. Map that takes the nematode population size at the start of the wet season to the nematode population size at the start of the wet season the following year, with no host size effect. Maps are plotted on a log–log scale for clarity; therefore, the origin is excluded. Panel (a) shows the chaotic dynamics in nematode population with  $\beta = 3.5 \times 10^{-8}$ . Panel (b) shows the period two dynamics of map when  $\beta = 1.5 \times 10^{-7}$ . The chaotic dynamics are no longer present because the hump has moved to the range where the nematode population is less than 1. Panel (c) shows the coexistence of a stable equilibrium and a stable two cycle (solid line); the unstable cycle is indicated with a dashed line,  $\beta = 1.9 \times 10^{-7}$ . These dynamics exist for a very narrow region parameter space when the  $N^{k+1}(0) = N^k(0)$  line crosses through the region of the map where the slope quickly transitions from being greater in absolute value than 1 to being less than 1 in absolute value. Panel (d) shows the stable equilibrium dynamics with  $\beta = 3 \times 10^{-7}$ .

suring time series of infective juvenile densities in the soil. We have documented 1  
 colonization of new lupine rhizospheres, and suggest the reasonable hypothesis 2  
 that low rates of nematode immigration counter local extinctions, as first suggested 3  
 for host–parasitoid interactions by [Nicholson and Bailey \(1935\)](#) and as in classical 4  
 metapopulation theory [e.g. [Hastings and Harrison \(1994\)](#), [Hanski \(1999\)](#)]. 5

#### ACKNOWLEDGEMENT

This research was supported by NSF Grant 0315289.

## REFERENCES

- 1
- 2 Billings, L. and I. B. Schwartz (2002). Exciting chaos with noise: unexpected dynamics in  
3 epidemic outbreaks. *J. Math. Biol.* **44**, 31–48.
- 4 Bjørnstad, O. N. and B. T. Grenfell (2001). Noisy clockwork: time series analysis of pop-  
5 ulation fluctuations in animals. *Science* **293**, 638–643.
- 6 Blasius, B., A. Huppert and L. Stone (1999). Complex dynamics and phase synchroniza-  
7 tion in spatially extended ecological systems. *Nature* **399**, 354–359.
- 8 Briggs, C. J. and H. C. J. Godfray (1996). The dynamics of insect-pathogen interactions in  
9 seasonal environments. *Theor. Popul. Biol.* **50**, 149–177.
- 10 Briggs, C. J., R. M. Nisbet and W. W. Murdoch (1999). Delayed feedback and multiple  
11 attractors in a host-parasitoid system. *J. Math. Biol.* **38**, 317–345.
- 12 Campbell, J. F., A. M. Koppenhöfer, H. K. Kaya and B. Chinnasri (1999). Are there tem-  
13 porarily non-infectious dauer stages in entomopathogenic nematode populations: a test  
14 of the phased infectivity hypothesis. *Parasitology* **118**, 499–508.
- 15 Earn, D. J. D., P. Rohani, B. M. Bolker and B. T. Grenfell (2000). A simple model for  
16 complex dynamical transitions in epidemics. *Science* **287**, 667–670.
- 17 Fenton, A., R. Norman, J. P. Fairbairn and P. J. Hudson (2000). Modelling the efficacy of  
18 entomopathogenic nematodes in the regulation of invertebrate pests in glasshouse crops.  
19 *J. Appl. Ecol.* **37**, 309–320.
- 20 Fenton, A., R. Norman, J. P. Fairbairn and P. J. Hudson (2001). Evaluating the efficacy of  
21 entomopathogenic nematodes for the biological control of crop pests: a nonequilibrium  
22 approach. *Am. Nat.* **158**, 408–425.
- 23 Finkenstadt, B. F. and B. T. Grenfell (2000). Time series modelling of childhood diseases:  
24 a dynamical systems approach. *J. R. Statist. Soc. Ser. C-Appl. Stat.* **49**, 187–205.
- 25 Gamarra, J. G. P. and R. V. Sole (2002). Biomass-diversity responses and spatial depen-  
26 dencies in disturbed tallgrass prairies. *J. Theor. Biol.* **215**, 469–480.
- 27 Grenfell, B. T., K. Wilson, B. F. Finkenstaedt, T. N. Coulson, S. Murray, S. D. Albon,  
28 J. M. Pemberton, T. H. Clutton-Brock and M. J. Crawley (1998). Noise and determinism  
29 in synchronized sheep dynamics. *Nature* **394**, 674–677.
- 30 Hanski, I. A. (1999). *Metapopulation Ecology*, Oxford Series in Ecology and Evolution,  
31 Oxford, New York: Oxford University Press.
- 32 Hanski, I. A. and E. Korpimäki (1995). Microtine rodent dynamics in northern Europe:  
33 parameterized models for the predator–prey interaction. *Ecology* **76**, 840–850.
- 34 Hastings, A. and S. Harrison (1994). Metapopulation dynamics and genetics. *Annu. Rev.*  
35 *Ecol. Syst.* **25**, 167–188.
- 36 Hastings, A., C. L. Hom, S. Ellner, P. Turchin and H. C. J. Godfray (1993). Chaos in  
37 ecology: is mother nature a strange attractor? *Annu. Rev. Ecol. Syst.* **24**, 1–33.
- 38 Higgins, K., A. Hastings, J. N. Sarvela and L. W. Botsford (1997). Stochastic dynam-  
39 ics and deterministic skeletons: population behavior of dungeness crab. *Science* **276**,  
40 1431–1435.
- 41 Hominick, W. M. (2002). Biogeography, in *Entomopathogenic Nematology*, R. Gau-  
42 gler (Ed.), New York: CABI Publishing, pp. 115–145.
- 43 Karlin, S. and J. McGregor (1972). Polymorphisms for genetic and ecological systems with  
44 weak coupling. *Theor. Popul. Biol.* **3**, 210–238.
- 45 Kaya, H. K. and R. Gaugler (1993). Entomopathogenic nematodes, in *Annual Review of*  
46 *Entomology*, Palo Alto: Annual Reviews Inc., pp. 181–206.

- Kendall, B. E., C. J. Briggs, W. W. Murdoch, P. Turchin, S. P. Ellner, E. McCauley, R. M. Nisbet and S. N. Wood (1999). Why do populations cycle? A synthesis of statistical and mechanistic modeling approaches. *Ecology* **80**, 1789–1805.
- Koppenhöfer, A. M., B. A. Jaffee, A. E. Muldoon, D. R. Strong and H. K. Kaya (1996). Effect of nematode-trapping fungi on an entomopathogenic nematode originating from the same field site in California. *J. Invertebr. Pathol.* **68**, 246–252.
- May, R. M. and G. F. Oster (1976). Bifurcations and dynamic complexity in simple ecological models. *Am. Naturalist* **110**, 573–599.
- Mills, N. J. and W. M. Getz (1996). Modelling the biological control of insect pests: a review of host–parasitoid models. *Ecolo. Modelling* **92**, 121–143.
- Nicholson, A. J. and V. A. Bailey (1935). The balance of animal populations. *Proc. Zoological Soc. London* **3**, 551–598.
- Polis, G. A. and D. R. Strong (1996). Food web complexity and community dynamics. *Am. Nat.* **147**, 813–846.
- Roberts, M. G. and J. A. P. Heesterbeek (1998). A simple parasite model with complicated dynamics. *J. Math. Biol.* **37**, 272–290.
- Stenseth, N. C. (1999). Population cycles in voles and lemmings: density dependence and phase dependence in a stochastic world. *Oikos* **87**, 427–461.
- Strong, D. R. (1999). Predator control in terrestrial ecosystems: the underground food chain of bush lupine, in *Herbivores, Between Plants and Predators*, H. Olf, V. K. Brown and R. H. Drent (Eds), Oxford: Blackwell Science, pp. 577–602.
- Strong, D. R. (2002). Populations or entomopathogenic nematodes in food webs, in *Entomopathogenic Nematology*, Chapter 11, R. Gaugler (Ed.), NY: CABI Publishing.
- Strong, D. R., H. K. Kaya, A. V. Whipple, A. L. Child, S. Kraig, M. Bondonno, K. Dyer and J. L. Maron (1996). Entomopathogenic nematodes: natural enemies of root-feeding caterpillars on bush lupine. *Oecologia* **108**, 167–173.
- Strong, D. R., A. V. Whipple, A. L. Child and B. Dennis (1999). Model selection for a subterranean trophic cascade: root-feeding caterpillars and entomopathogenic nematodes. *Ecology (Washington DC)* **80**, 2750–2761.
- Stuart, R. J. and R. Gaugler (1994). Patchiness in populations of entomopathogenic nematodes. *J. Invertebr. Pathol.* **64**, 39–45.
- Turchin, P. and S. P. Ellner (2000). Living on the edge of chaos: population dynamics of fennoscandian voles. *Ecology* **81**, 3099–3116.
- Wagner D. L. (1985). The biosystematics of *Hepialus* F. s. lato, with special emphasis on the californicus-hectoides species group. PhD Thesis, U.C. Berkeley.
- Westerman, P. R. (1998). Penetration of the entomopathogenic nematode heterorhabditis spp into host insects at 9 and 20 degrees c. *J. Invertebr. Pathol.* **72**, 197–205.

Received 27 August 2002 and accepted 24 September 2003 38