Interspecific strategic effects of mobility in predator-prey systems

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- 2. The Model
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We assume both prey and predators adjust their activities to maximize their per capita growth rates.

Prey

Predators

mobile strategy

mobile (active) strategy

sessile strategy

sessile (ambush)

The Model

The underlying predator-prey dynamics is modelled by Gausetype interactions (Gause, 1934; Hofbauer and Sigmund, 1998) as follows. If x is the prey population size (i.e. prey density) and y is predator density, then

$$\dot{x} = xf(x) - yF(x)$$
(1)
$$\dot{y} = -\mu y + cyF(x).$$

Here f(x) is the per capita growth rate of the prey population in the absence of predators and F(x) is the number of prey killed by one predator per unit time.

The death rate of predators is μ and c is the conversion factor.

- α_{MM} Foraging efficiencies of the mobile predator chasing mobile prey
- α_{MS} Foraging efficiencies of the mobile predator chasing sessile prey
- α_{SM} Foraging efficiencies of a sessile predator catching mobile prey
- α_{SS} Foraging efficiencies of a sessile predator catching sessile prey



$\begin{array}{ccc} & & & & & & \\ Prey & & & & \\ Predator & & & \\ Sessile & & & \\ \end{array} \begin{bmatrix} \alpha_{MM} & \alpha_{MS} \\ \alpha_{SM} & \alpha_{SS} \end{bmatrix}$

Table 1. Predator foraging efficiency.

 θ_x and θ_y are the current mobile proportions in the system.

The expected number of prey killed per unit time by a randomly selected predator, $F(x; \theta_x, \theta_y)$, is given by

$$F(x;\theta_x,\theta_y) = \theta_y \left[F_{MM}(\theta_x x) + F_{MS}((1-\theta_x)x) \right] + (1-\theta_y) \left[F_{SM}(\theta_x x) + F_{SS}((1-\theta_x)x) \right]$$

where F_{AB} is the functional response of a predator using strategy A when the prey population uses strategy B with $A, B \in \{M, S\}$.

The Model

Both prey and predators use game theory-based strategies to maximize their per capita population growth rates.

Replicator equation:

$$\dot{\theta}_x = u\theta_x \left(1 - \theta_x\right) \left[-\frac{F_{MM}(\theta_x x)\theta_y y}{\theta_x x} - \frac{F_{SM}(\theta_x x)(1 - \theta_y) y}{\theta_x x} + \frac{F_{MS}((1 - \theta_x)x)\theta_y y}{(1 - \theta_x)x} + \frac{F_{SS}((1 - \theta_x)x)(1 - \theta_y) y}{(1 - \theta_x)x} \right]$$

$$\dot{\theta}_y = v\theta_y \left(1 - \theta_y\right) \left[c \left(F_{MM}(\theta_x x) + F_{MS}((1 - \theta_x)x)\right) - c \left(F_{SM}(\theta_x x) + F_{SS}((1 - \theta_x)x)\right) \right].$$

The Model

Both prey and predators use game theory-based strategies to maximize their per capita population growth rates.

Smoothed best response:

$$\dot{\theta}_{x} = \frac{\theta_{x} \exp m \left(-\frac{F_{MM}(\theta_{x}x)\theta_{y}y}{\theta_{x}x} - \frac{F_{SM}(\theta_{x}x)(1-\theta_{y})y}{\theta_{x}x}\right)}{\theta_{x}x} \right) }{\theta_{x} \exp m \left(-\frac{F_{MM}(\theta_{x}x)\theta_{y}y}{\theta_{x}x} - \frac{F_{SM}(\theta_{x}x)(1-\theta_{y})y}{\theta_{x}x}\right) + (1-\theta_{x})\exp m \left(-\frac{F_{MS}((1-\theta_{x})x)\theta_{y}y}{(1-\theta_{x})x} - \frac{F_{SS}((1-\theta_{x})x)(1-\theta_{y})y}{(1-\theta_{x})x}\right)}{\theta_{y} \exp nc(F_{MM}(\theta_{x}x) + F_{MS}((1-\theta_{x})x))} - \theta_{y}.$$

 $\dot{\theta}_x$

 $\dot{\theta}_y$

For the L-V model, substituting $F_{AB}(x) = \alpha_{AB}x$, yields the following population density and (replicator) strategy dynamics, respectively.

$$\begin{aligned} \dot{x} &= rx\left(1 - \frac{x}{K}\right) - \alpha_{MM}\theta_{y}y\theta_{x}x - \alpha_{MS}\theta_{y}y\left(1 - \theta_{x}\right)x \\ &- \alpha_{SM}\left(1 - \theta_{y}\right)y\theta_{x}x - \alpha_{SS}\left(1 - \theta_{y}\right)y\left(1 - \theta_{x}\right)x \\ \dot{y} &= -\mu y + c\alpha_{MM}\theta_{y}y\theta_{x}x + c\alpha_{MS}\theta_{y}y\left(1 - \theta_{x}\right)x \\ &+ c\alpha_{SM}\left(1 - \theta_{y}\right)y\theta_{x}x + c\alpha_{SS}\left(1 - \theta_{y}\right)y\left(1 - \theta_{x}\right)x \end{aligned}$$
$$= u\theta_{x}\left(1 - \theta_{x}\right)\left(-\alpha_{MM}\theta_{y}y - \alpha_{SM}\left(1 - \theta_{y}\right)y + \alpha_{MS}\theta_{y}y + \alpha_{SS}\left(1 - \theta_{y}\right)y\right) \\ = v\theta_{y}\left(1 - \theta_{y}\right)\left(c\alpha_{MM}\theta_{x}x + c\alpha_{MS}\left(1 - \theta_{x}\right)x - c\alpha_{SM}\theta_{x}x - c\alpha_{SS}\left(1 - \theta_{x}\right)x\right) \end{aligned}$$

The effect of a dominated strategy

Suppose a mobile predator has higher foraging efficiency than a sessile predator independent of the strategy of the prey

 $\alpha_{MM} > \alpha_{SM}$ $\alpha_{MS} > \alpha_{SS}$

Suppose $\alpha_{MM} > \alpha_{MS}$.

The proportion of mobile predators is increasing

$$\dot{\theta}_y \ge 0$$

and will evolve to 1 unless *x* evolves to 0.



 $\dot{\theta}_x$ is always negative (if $\alpha_{MS} < \alpha_{MM}$)

 θ_y is sufficiently close to 1

or

y evolves to 0

x evolves to the carrying capacity K

 $\begin{array}{l} \theta_y \text{ is sufficiently close to } 1\\ \dot{\theta}_x \text{ is always negative (if } \alpha_{MS} < \alpha_{MM}) \\ & & \\ & \\ & \\ \theta_x = 0 \text{ and } \theta_y = 1 \\ & \\ & \\ & \\ \dot{x} = rx \left(1 - \frac{x}{K}\right) - \alpha_{MS} yx \\ & \\ \dot{y} = -\mu y + c \alpha_{MS} yx. \end{array}$

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$$\dot{x} = rx\left(1 - \frac{x}{K}\right) - \alpha_{MS}yx$$
$$\dot{y} = -\mu y + c\alpha_{MS}yx.$$

The system has an interior equilibrium

$$(x,y) \;=\; \left(rac{\mu}{c lpha_{MS}}, rac{r(c lpha_{MS}K-\mu)}{c lpha_{MS}^2 K}
ight)$$

if and only if $\frac{\mu}{cK} < \alpha_{MS}$.

The interior equilibrium is globally asymptotically stable if it exists; otherwise,

$$(x, y)$$
 evolves to $(K, 0)$.



Replicator equation $\frac{\mu}{cK} < \alpha_{MS}$



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Using the smoothed best response dynamics, we have

$$\dot{\theta}_x = \frac{\theta_x \exp m(-\alpha_{MM}\theta_y y - \alpha_{SM}(1 - \theta_y)y)}{\theta_x \exp m(-\alpha_{MM}\theta_y y - \alpha_{SM}(1 - \theta_y)y) + (1 - \theta_x) \exp m(-\alpha_{MS}\theta_y y - \alpha_{SS}(1 - \theta_y)y)} - \theta_x \dot{\theta}_y = \frac{\theta_y \exp n(c\alpha_{MM}\theta_x x + c\alpha_{MS}(1 - \theta_x)x)}{\theta_y \exp n(c\alpha_{MM}\theta_x x + c\alpha_{MS}(1 - \theta_x)x) + (1 - \theta_y) \exp n(c\alpha_{SM}\theta_x x + c\alpha_{SS}(1 - \theta_x)x)} - \theta_y.$$



Smoothed best response strategy dynamics $\frac{\mu}{cK} < \alpha_{MS}$



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No dominated strategy

We assume that

α_{MS} and α_{SM} are both larger than α_{MM} and α_{SS}

The predator forages more efficiently if it has the opposite strategy as its prey.

Prey are better able to avoid predation if they have the same strategy as the predator.

Nash equilibrium for the frequency-dependent evolutionary game between predators and prey is given by

$$\theta_x^* = \frac{\alpha_{MS} - \alpha_{SS}}{\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}} \text{ and } \theta_y^* = \frac{\alpha_{SM} - \alpha_{SS}}{\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}}$$

(No individual can increase its fitness by altering its strategy.)

With the mobile proportions fixed at their nash equilibrium values, the population dynamics becomes

$$\dot{x} = rx\left(1 - \frac{x}{K}\right) - \alpha^* xy$$
$$\dot{y} = -\mu y + c\alpha^* xy$$

where

$$\alpha^* \equiv \frac{\alpha_{MS} \alpha_{SM} - \alpha_{SS} \alpha_{MM}}{\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}} > 0$$



The equilibrium of most interest now is one where both strategic behaviors are present for each species

$$E_1 = (x^*, y^*, \theta_x^*, \theta_y^*),$$

where

$$\begin{aligned} x^* &= \frac{\mu}{c\alpha^*}, \\ y^* &= \frac{r}{\alpha^*} \left(1 - \frac{\mu}{cK\alpha^*} \right), \\ \theta^*_x &= \frac{\alpha_{MS} - \alpha_{SS}}{\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}}, \\ \theta^*_y &= \frac{\alpha_{SM} - \alpha_{SS}}{\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}}, \end{aligned}$$

 E_1 exists if and only if $\frac{1}{\alpha^*} < \frac{cK}{\mu}$.



Trajectories of the LV system with strategy dynamics given by the replicator equation when E_1 exists.



Trajectories of the LV system with strategy dynamics given by the replicator equation when E_1 exists.



Trajectories of the LV system with strategy dynamics given by the smoothed best response equation when E_1 exists.



Trajectories of the LV system with strategy dynamics given by the smoothed best response equation when E_1 exists.

The coupled system may also have equilibria where both predators and prey are present but the populations adopt pure strategies.

$$E_{2} = \left(\frac{\mu}{c\alpha_{SM}}, \frac{r\left(-\mu + c\alpha_{SM}K\right)}{c\alpha_{SM}^{2}K}, 1, 0\right)$$

$$E_{3} = \left(\frac{\mu}{c\alpha_{MS}}, \frac{r\left(-\mu + c\alpha_{MS}K\right)}{c\alpha_{MS}^{2}K}, 0, 1\right)$$

$$E_{4} = \left(\frac{\mu}{c\alpha_{SS}}, \frac{r\left(-\mu + c\alpha_{SS}K\right)}{c\alpha_{SS}^{2}K}, 0, 0\right)$$

$$E_{5} = \left(\frac{\mu}{c\alpha_{MM}}, \frac{r\left(-\mu + c\alpha_{MM}K\right)}{c\alpha_{MM}^{2}K}, 1, 1\right)$$

 E_2, E_3, E_4, E_5 are unstable if they exist.

The only other equilibria of the coupled system are when the predator population is extinct, and the prey population is at carrying capacity.

$$E_{6} = (K, 0, \hat{\theta}_{x}, 0),$$

$$E_{7} = (K, 0, \hat{\theta}_{x}, 1),$$

$$E_{8} = (K, 0, \frac{\alpha_{MS} - \alpha_{SS}}{\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}}, \hat{\theta}_{y})$$

where $\hat{\theta}_x$ and $\hat{\theta}_y$ take any value between 0 and 1. Finally, there is also the trivial equilibrium with no prey and predators $E_9 = (0, 0, \hat{\theta}_x, \hat{\theta}_y).$

An alternative Lotka-Volterra model and the global stability of E_1

 $x_1 = \theta_x x \qquad \qquad x_2 = (1 - \theta_x) x$ $y_1 = \theta_y y \qquad \qquad y_2 = (1 - \theta_y) y$

$$\begin{aligned} \dot{x}_{1} &= rx_{1}\left(1 - \frac{x_{1} + x_{2}}{K}\right) - \alpha_{SM}x_{1}y_{2} - \alpha_{MM}x_{1}y_{1} + \frac{x_{1}x_{2}(u-1)(\alpha_{MS}y_{1} + \alpha_{SS}y_{2} - \alpha_{SM}y_{2} - \alpha_{MM}y_{1})}{x_{1} + x_{2}} \\ \dot{x}_{2} &= rx_{2}\left(1 - \frac{x_{1} + x_{2}}{K}\right) - \alpha_{MS}x_{2}y_{1} - \alpha_{SS}x_{2}y_{2} + \frac{x_{1}x_{2}(u-1)(\alpha_{SM}y_{2} + \alpha_{MM}y_{1} - \alpha_{MS}y_{1} - \alpha_{SS}y_{2})}{x_{1} + x_{2}} \\ \dot{y}_{1} &= -\mu y_{1} + c\alpha_{MS}x_{2}y_{1} + c\alpha_{MM}x_{1}y_{1} + \frac{y_{1}y_{2}(v-1)(c\alpha_{MS}x_{2} + c\alpha_{MM}x_{1} - c\alpha_{SS}x_{2})}{y_{1} + y_{2}} \\ \dot{y}_{2} &= -\mu y_{2} + c\alpha_{SM}x_{1}y_{2} + c\alpha_{SS}x_{2}y_{2} + \frac{y_{1}y_{2}(v-1)(c\alpha_{SM}x_{1} + c\alpha_{SS}x_{2} - c\alpha_{MS}x_{2} - c\alpha_{MM}x_{1})}{y_{1} + y_{2}}. \end{aligned}$$

$$\begin{aligned} x_1^* &= \frac{\mu(\alpha_{MS} - \alpha_{SS})}{c(\alpha_{MS}\alpha_{SM} - \alpha_{SS}\alpha_{MM})}, \\ x_2^* &= \frac{\mu(\alpha_{SM} - \alpha_{MM})}{c(\alpha_{MS}\alpha_{SM} - \alpha_{SS}\alpha_{MM})}, \\ y_1^* &= \frac{r(\alpha_{SM} - \alpha_{SS})(-\mu\alpha_{MS} + \alpha_{MS}c\alpha_{SM}K + \mu\alpha_{MM} - c\alpha_{SS}K\alpha_{MM} - \mu\alpha_{SM} + \mu\alpha_{SS})}{Kc(\alpha_{MS}\alpha_{SM} - \alpha_{SS}\alpha_{MM})^2}, \\ y_2^* &= \frac{r(-\alpha_{MM} + \alpha_{MS})(-\mu\alpha_{MS} + \alpha_{MS}c\alpha_{SM}K + \mu\alpha_{MM} - c\alpha_{SS}K\alpha_{MM} - \mu\alpha_{SM} + \mu\alpha_{SS})}{Kc(\alpha_{MS}\alpha_{SM} - \alpha_{SS}\alpha_{MM})^2}, \end{aligned}$$

In particular, in the special case that u = v = 1, we obtain

$$\dot{x}_1 = rx_1 \left(1 - \frac{x_1 + x_2}{K}\right) - \alpha_{SM} x_1 y_2 - \alpha_{MM} x_1 y_1 \dot{x}_2 = rx_2 \left(1 - \frac{x_1 + x_2}{K}\right) - \alpha_{MS} x_2 y_1 - \alpha_{SS} x_2 y_2 \dot{y}_1 = -\mu y_1 + c\alpha_{MS} x_2 y_1 + c\alpha_{MM} x_1 y_1 \dot{y}_2 = -\mu y_2 + c\alpha_{SM} x_1 y_2 + c\alpha_{SS} x_2 y_2.$$

The global asymptotic stability of the system can the be shown by considering the following Lyapunov function

$$V = c \left(x_1 - x_1^* - x_1^* \ln \left(\frac{x_1}{x_1^*} \right) \right) + c \left(x_2 - x_2^* - x_2^* \ln \left(\frac{x_2}{x_2^*} \right) \right) + \left(y_1 - y_1^* - y_1^* \ln \left(\frac{y_1}{y_1^*} \right) \right) \\ + \left(y_2 - y_2^* - y_2^* \ln \left(\frac{y_2}{y_2^*} \right) \right),$$

The derivative of V is obtained as

$$\dot{V} = c \left(x_1 - x_1^*\right) \left(r \left(1 - \frac{x}{K}\right) - \alpha_{SM} y_2 - \alpha_{MM} y_1\right) + c \left(x_2 - x_2^*\right) \left(r \left(1 - \frac{x}{K}\right) - \alpha_{MS} y_1 - \alpha_{SS} y_2\right) \\ + \left(y_1 - y_1^*\right) \left(-\mu y_1 + c \alpha_{MS} x_2 + c \alpha_{MM} x_1\right) + \left(y_2 - y_2^*\right) \left(-\mu + c \alpha_{SM} x_1 + c \alpha_{SS} x_2\right) \\ = -\frac{cr}{K} \left(x_1 - x_1^* + x_2 - x_2^*\right)^2.$$

Thus, $\dot{V} < 0$ unless $x = x^*$.

The trajectory must converge to an invariant subset of

$$\{(x_1, x_2, y_1, y_2) \mid x_1 + x_2 = x^* \text{ and } y \ge 0\}.$$

It can be shown that E_1 is globally asymptotically stable since all these trajectories converge to E_1 .

Let *M* be the maximal invariant subset

$$\{(x_1, x_2, y_1, y_2) \mid x_1 + x_2 = x^* \text{ and } y \ge 0\}$$

 $\dot{y} = \mu(y^* - y)$

Thus, (x_1, x_2, y_1, y_2) converges to an invariant subset of

$$\{(x_1, x_2, y_1, y_2) \mid x_1 + x_2 = x^* \text{ and } y_1 + y_2 = y^*\}.$$

$$\dot{x} = \dot{y} = 0 \text{ yields}$$

$$r\left(1 - \frac{x^*}{K}\right) \left(1 - \frac{\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}}{\alpha_{MS}\alpha_{SM} - \alpha_{SS}\alpha_{MM}} \left(\begin{array}{c}\alpha_{SS} + \theta_x \left(\alpha_{MS} - \alpha_{SS}\right) + \theta_y \left(\alpha_{SM} - \alpha_{SS}\right) \\ + \theta_x \theta_y \left(\alpha_{MM} - \alpha_{MS} - \alpha_{SM} + \alpha_{SS}\right)\end{array}\right)\right) = 0$$

which simplifies to

$$\left(\alpha_{SS} - \alpha_{SM} + \theta_y \left(\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}\right)\right) \left(\alpha_{SS} - \alpha_{MS} + \theta_x \left(\alpha_{SM} - \alpha_{MM} + \alpha_{MS} - \alpha_{SS}\right)\right) = 0.$$

That is, either $\theta_x = \theta_x^*$ or $\theta_y = \theta_y^*$ $\theta_x = \theta_x^*$ \longrightarrow $x_1 = x_1^*$ and $x_2 = x_2^*$, and $\dot{\theta}_x = 0$ $y_1 = \theta_y y = y_1^*$ and $y_2 = y_2^*$ Similarly, $\theta_y = \theta_y^*$ $x_1 = x_1^*$ and $x_2 = x_2^*$

Thus, every trajectory that is initially in the interior of M converges to E_1 .

In this section, we consider the game between the prey species and predator species with Holling type II functional (behavioral) response, i.e. $F(x) = \frac{\alpha x}{1+\alpha hx}$.

$$\begin{aligned} \dot{x} &= rx\left(1 - \frac{x}{K}\right) - \frac{\alpha_{SM}\theta_x x(1 - \theta_y)y}{\alpha_{SM}\theta_x xh + 1} - \frac{\alpha_{MS}(1 - \theta_x)x\theta_y y}{\alpha_{MS}(1 - \theta_x)xh + 1} - \frac{\alpha_{MM}\theta_x x\theta_y y}{\alpha_{MM}\theta_x xh + 1} - \frac{\alpha_{SS}(1 - \theta_x)x(1 - \theta_y)y}{\alpha_{SS}(1 - \theta_x)xh + 1} \\ \dot{y} &= -\mu y + \frac{c\alpha_{SM}\theta_x x(1 - \theta_y)y}{\alpha_{SM}\theta_x xh + 1} + \frac{c\alpha_{MS}(1 - \theta_x)x\theta_y y}{\alpha_{MS}(1 - \theta_x)xh + 1} + \frac{c\alpha_{MM}\theta_x x\theta_y y}{\alpha_{MM}\theta_x xh + 1} + \frac{c\alpha_{SS}(1 - \theta_x)x(1 - \theta_y)y}{\alpha_{SS}(1 - \theta_x)xh + 1} \\ \dot{\theta}_x &= u\theta_x \left(1 - \theta_x\right) \left(\frac{\alpha_{MS}\theta_y y}{\alpha_{MS}(1 - \theta_x)xh + 1} - \frac{\alpha_{SM}(1 - \theta_y)y}{\alpha_{SM}\theta_x xh + 1} + \frac{\alpha_{SS}(1 - \theta_y)y}{\alpha_{SS}(1 - \theta_x)xh + 1} - \frac{\alpha_{MM}\theta_y y}{\alpha_{MM}\theta_x xh + 1}\right) \\ \dot{\theta}_y &= v\theta_y \left(1 - \theta_y\right) \left(\frac{c\alpha_{MS}(1 - \theta_x)x}{\alpha_{MS}(1 - \theta_x)xh + 1} - \frac{c\alpha_{SM}\theta_x x}{\alpha_{SM}\theta_x xh + 1} + \frac{c\alpha_{MM}\theta_x x}{\alpha_{MM}\theta_x xh + 1} - \frac{c\alpha_{SS}(1 - \theta_x)x}{\alpha_{SS}(1 - \theta_x)xh + 1}\right) \end{aligned}$$

We assume $\alpha_{MM} = 0$ and $\alpha_{SS} = 0$

Then the population density and strategy dynamics becomes

$$\begin{aligned} \dot{x} &= rx\left(1 - \frac{x}{K}\right) - \frac{\alpha_{SM}\theta_x x(1 - \theta_y)y}{\alpha_{SM}\theta_x xh + 1} - \frac{\alpha_{MS}(1 - \theta_x)x\theta_y y}{\alpha_{MS}(1 - \theta_x)xh + 1} \\ \dot{y} &= -\mu y + \frac{c\alpha_{SM}\theta_x x(1 - \theta_y)y}{\alpha_{SM}\theta_x xh + 1} + \frac{c\alpha_{MS}(1 - \theta_x)x\theta_y y}{\alpha_{MS}(1 - \theta_x)xh + 1} \\ \dot{\theta}_x &= u\theta_x \left(1 - \theta_x\right) \left(\frac{\alpha_{MS}\theta_y y}{\alpha_{MS}(1 - \theta_x)xh + 1} - \frac{\alpha_{SM}(1 - \theta_y)y}{\alpha_{SM}\theta_x xh + 1}\right) \\ \dot{\theta}_y &= v\theta_y \left(1 - \theta_y\right) \left(\frac{c\alpha_{MS}(1 - \theta_x)x}{\alpha_{MS}(1 - \theta_x)xh + 1} - \frac{c\alpha_{SM}\theta_x x}{\alpha_{SM}\theta_x xh + 1}\right) \end{aligned}$$

The interior equilibrium is given by $E_{h1} = (x_h^*, y_h^*, \theta_{xh}^*, \theta_{yh}^*)$ where

$$\begin{aligned} x_h^* &= \frac{\mu}{\alpha^*(c-\mu h)} \\ y_h^* &= \frac{rc}{\alpha^*(c-\mu h)} \left(1 - \frac{\mu}{(c-\mu h)K\alpha^*} \right) \\ \theta_{xh}^* &= \frac{\alpha_{MS}}{\alpha_{MS} + \alpha_{SM}} \\ \theta_{yh}^* &= \frac{\alpha_{SM}}{\alpha_{MS} + \alpha_{SM}}. \end{aligned}$$

Moreover, with the mobile proportions fixed at the equilibrium values,

$$(heta^*_{xh}, heta^*_{yh})$$

the population dynamics becomes

$$\dot{x} = rx\left(1 - \frac{x}{K}\right) - \frac{\alpha^* xy}{\alpha^* xh + 1}$$
$$\dot{y} = -\mu y + \frac{c\alpha^* xy}{\alpha^* xh + 1}.$$

That is, equilibrium E_{h1} is biologically feasible if and only if the following conditions hold:

$$(\mathcal{H}1) \ \mu h < c \text{ and } \frac{1}{\alpha^*} < \frac{(c-\mu h)K}{\mu}.$$

The system has biologically feasible pure strategic equilibrium $E_{h2} = \left(\frac{\mu}{\alpha_{SM}(c-\mu h)}, \frac{rc}{\alpha_{SM}(c-\mu h)} \left(1 - \frac{\mu}{(c-\mu h)K\alpha_{SM}}\right), 1, 0\right)$ if and only if the following conditions hold.

 $(\mathcal{H}2) \ \mu h < c \text{ and } \frac{\mu}{(c-\mu h)K} < \alpha_{SM}.$

Under the condition

$$(\mathcal{H}3) \ \mu \ h < c \ \text{and} \ \frac{\mu}{(c-\mu h)K} < \alpha \ _{MS}.$$
$$E_{h3} = \left(\frac{\mu}{\alpha \ _{MS}(c-\mu h)}, \ \frac{rc}{\alpha_{MS}(c-\mu h)} \left(1 - \frac{\mu}{(c-\mu h)K\alpha_{MS}}\right), \ 0, \ 1\right) \text{ exists as}$$
a biologically feasible pure strategic equilibrium.

The system has three equilibria where the predator population goes to extinction and the prey population reaches carrying capacity.

These equilibria are

$$E_{h4} = (K, 0, \hat{\theta}_x, 0),$$
$$E_{h5} = (K, 0, \hat{\theta}_x, 1)$$

and

$$E_{h6} = (K, 0, \frac{\alpha_{MS}}{\alpha_{MS} + \alpha_{SM}}, \hat{\theta}_y),$$

where $\hat{\theta}_x$ and $\hat{\theta}_y$ take any value between 0 and 1.

In addition, the system has a trivial equilibrium point

$$E_{h7} = (0, 0, \hat{\theta}_x, \hat{\theta}_y)$$

where both prey and predator populations are extinct.

 E_{h1} is always unstable.

 E_{h2} and E_{h3} , are both unstable.

The trivial equilibrium E_{h7} , which corresponds to the extinction of both the prey and predator species is also unstable.



Trajectories of the RM system with strategy dynamics given by the replicator equation when E_{h1} exists.



Trajectories of the RM system with strategy dynamics given by the replicator equation when E_{h1} exists.



Trajectories of the RM system with strategy dynamics given by the replicator equation when E_{h1} exists.





Interestingly, the situation changes when the strategy dynamics is given by the smoothed best response:

$$\dot{\theta}_x = \frac{\theta_x e^{-\frac{m\alpha}{\alpha} \frac{SM(1-\theta-y)y}{SM\theta_x x h+1}}}{\theta_x e^{-\frac{m\alpha}{\alpha} \frac{SM(1-\theta-y)y}{SM\theta_x x h+1}} + (1-\theta_x) e^{-\frac{m\alpha}{\alpha} \frac{MS\theta_y y}{MS^{(1-\theta_x)x h+1}}} - \theta_x$$

$$\dot{\theta}_{y} = \frac{\theta_{y} \mathrm{e}^{\frac{n c \alpha}{\alpha} \frac{M S^{(1-\theta_{x})x}}{M S^{(1-\theta_{x})xh+1}}}}{\theta_{y} \mathrm{e}^{\frac{n c \alpha}{\alpha} \frac{M S^{(1-\theta_{x})x}}{M S^{(1-\theta_{x})xh+1}} + (1-\theta_{y}) \mathrm{e}^{\frac{n c \alpha}{\alpha} \frac{S M^{\theta_{x}x}}{S M^{\theta_{x}xh+1}}} - \theta_{y}.$$

 E_{h1} is still unstable.



Trajectories of the RM system with strategy dynamics given by the smoothed best response when E_{h1} exists.



Trajectories of the RM system with strategy dynamics given by the smoothed best response when E_{h1} exists.

Conclusions

We investigate the dynamics of a predator-prey system with the assumption that both prey and predators use game theory-based strategies to maximize their per capita population growth rates.

The predators adjust their strategies in order to catch more prey per unit time, while the prey, on the other hand, adjust their reactions to minimize the chances of being caught.



Numerical simulation results indicate that, for some parameter values, the system has chaotic behavior.

Our investigation reveals the relationship between the game theory-based reactions of prey and predators, and their population changes.



Thank you!