

The Ecological Effects of Trait Variation in a u -Predator, v -Prey System

Sam Fleischer, Pablo Chavarria

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Advisors

- ▶ Dr. Jing Li
Assistant Professor, CSU Northridge
Department of Mathematics
- ▶ Dr. Casey terHorst
Assistant Professor, CSU Northridge
Biology Department

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Observations

- ▶ Predator/Prey interactions are prevalent in nature
 - ▶ Crab vs. gastropod
 - ▶ Protist vs. bacteria
- ▶ There is trait variation within species
 - ▶ Thickness of plant cuticula
 - ▶ Strength of gastropod shell
- ▶ Incorporating trait variation provides richer dynamics than classical Lotka-Volterra models

Lotka-Volterra

The Ecological Effects
of Trait Variation in a
 u -Predator, v -Prey
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Chavarria

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Motivation

Model Formulation

Preliminary Results

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The Model

Case 1: $u = 1, v = 1$

Equilibria Analysis

Stability Analysis

Special Case:

$M^* = N^* = 0$

Special Case:

$M^* = 0, N^* = K$

$$\frac{dN}{dt} = N(b - aM)$$
$$\frac{dM}{dt} = M(aeN - d)$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $M \equiv$ Predator Density

Parameters

- ▶ $a \equiv$ Attack rate
- ▶ $b \equiv$ Prey birth rate
- ▶ $e \equiv$ Efficiency
- ▶ $d \equiv$ Predator death rate

$$\frac{dN}{dt} = N(b - aM)$$
$$\frac{dM}{dt} = M(aeN - d)$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $M \equiv$ Predator Density

Parameters

- ▶ $a \equiv$ Attack rate \leftarrow *No variation!*
- ▶ $b \equiv$ Prey birth rate
- ▶ $e \equiv$ Efficiency
- ▶ $d \equiv$ Predator death rate

$$a(m) = \alpha \exp \left[-\frac{(m - \theta)^2}{2\tau^2} \right]$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $M \equiv$ Predator Density
- ▶ $m \equiv$ Predator Character (Trait Value)

Parameters

- ▶ $\alpha \equiv$ Maximum attack rate
- ▶ $\theta \equiv$ Optimal trait value
- ▶ $\tau \equiv$ Specialization Constant

$$a(m) = \alpha \exp \left[-\frac{(m - \theta)^2}{2\tau^2} \right]$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $M \equiv$ Predator Density
- ▶ $m \equiv$ Predator Character (Trait Value)

Parameters

- ▶ $\alpha \equiv$ Maximum attack rate
- ▶ $\theta \equiv$ Optimal trait value \longleftarrow *No variation!*
- ▶ $\tau \equiv$ Specialization Constant

Our Extension

$$a(m, n) = \alpha \exp \left[-\frac{(m - n - \theta)^2}{2\tau^2} \right]$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $M \equiv$ Predator Density
- ▶ $n \equiv$ Prey Character (Trait Value)
- ▶ $m \equiv$ Predator Character (Trait Value)

Parameters

- ▶ $\alpha \equiv$ Maximum attack rate
- ▶ $\theta \equiv$ Optimal trait difference
- ▶ $\tau \equiv$ Specialization Constant

Our Extension

Distribution Assumptions

- ▶ Trait values are **normally distributed** over the populations

$$p(n, \bar{n}) = \frac{1}{\sqrt{2\pi\beta^2}} \exp \left[-\frac{(n - \bar{n})^2}{2\beta^2} \right]$$

$$p(m, \bar{m}) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left[-\frac{(m - \bar{m})^2}{2\sigma^2} \right]$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $\bar{n} \equiv$ Mean Prey Character
- ▶ $M \equiv$ Predator Density
- ▶ $\bar{m} \equiv$ Mean Predator Character

Parameters

- ▶ $\beta^2 \equiv$ Prey Trait Variance
- ▶ $\sigma^2 \equiv$ Predator Trait Variance

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Our Extension

Average Attack Rate

$$\begin{aligned}\bar{a}(\bar{m}, \bar{n}) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} a(m, n) \cdot p(m, \bar{m}) \cdot p(n, \bar{n}) dm dn \\ &= \frac{\alpha \tau}{\sqrt{\sigma^2 + \beta^2 + \tau^2}} \exp \left[-\frac{(\bar{m} - \bar{n} - \theta)^2}{2(\sigma^2 + \beta^2 + \tau^2)} \right]\end{aligned}$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $\bar{n} \equiv$ Mean Prey Character
- ▶ $M \equiv$ Predator Density
- ▶ $\bar{m} \equiv$ Mean Predator Character

Parameters

- ▶ $\beta^2 \equiv$ Prey Trait Variance
- ▶ $\sigma^2 \equiv$ Predator Trait Variance
- ▶ $\alpha \equiv$ Maximum attack rate
- ▶ $\theta \equiv$ Optimal trait difference
- ▶ $\tau \equiv$ Specialization Constant

Our Extension

Fitness Assumptions

- ▶ Prey experiences logistic growth in absence of predator
- ▶ Predator experiences exponential decay in absence of prey

$$Y(m, n, M, N) = r \left(1 - \frac{N}{K} \right) - Ma(m, n)$$

$$W(m, n, N) = eNa(m, n) - d$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $n \equiv$ Prey Character
- ▶ $M \equiv$ Predator Density
- ▶ $m \equiv$ Predator Character

Parameters

- ▶ $r \equiv$ Intrinsic Prey Growth Rate
- ▶ $K \equiv$ Prey Carrying Capacity
- ▶ $d \equiv$ Predator Death Rate
- ▶ $e \equiv$ Efficiency

Our Extension

Average Fitness

$$\begin{aligned}\overline{Y}(\overline{m}, \overline{n}, M, N) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} Y(m, n, M, N) \cdot p(m, \overline{m}) \cdot p(n, \overline{n}) dm dn \\ &= r \left(1 - \frac{N}{K} \right) - M \overline{a}(\overline{m}, \overline{n})\end{aligned}$$

$$\begin{aligned}\overline{W}(\overline{m}, \overline{n}, N) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} W(m, n, N) \cdot p(m, \overline{m}) \cdot p(n, \overline{n}) dm dn \\ &= e N \overline{a}(\overline{m}, \overline{n}) - d\end{aligned}$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $\overline{n} \equiv$ Mean Prey Character
- ▶ $M \equiv$ Predator Density
- ▶ $\overline{m} \equiv$ Mean Predator Character

Parameters

- ▶ $r \equiv$ Intrinsic Prey Growth Rate
- ▶ $K \equiv$ Prey Carrying Capacity
- ▶ $d \equiv$ Predator Death Rate
- ▶ $e \equiv$ Efficiency

Our Extension

Ecological Component of the Model

$$\frac{dN}{dt} = N \cdot \bar{Y}(\bar{m}, \bar{n}, M, N) = N \left[r \left(1 - \frac{N}{K} \right) - M \bar{a}(\bar{m}, \bar{n}) \right]$$

$$\frac{dM}{dt} = M \cdot \bar{W}(\bar{m}, \bar{n}, N) = M [eN\bar{a}(\bar{m}, \bar{n}) - d]$$

Variables

- ▶ $N \equiv$ Prey Density
- ▶ $\bar{n} \equiv$ Mean Prey Character
- ▶ $M \equiv$ Predator Density
- ▶ $\bar{m} \equiv$ Mean Predator Character

Parameters

- ▶ $r \equiv$ Intrinsic Prey Growth Rate
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The Ecological Effects of Trait Variation in a u -Predator, v -Prey System (draft)

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Start Date: March 2014 - - Today's Date: March 3, 2015

Let $M_i(t)$ be the density of the i^{th} predator species, and let $N_j(t)$ be the density of the j^{th} prey species. Let $\overline{m}_i(t)$ be the mean of a single quantitative trait in the i^{th} predator species, and let $\overline{n}_j(t)$ be the mean of a single quantitative trait in the j^{th} prey species.

Suppose the traits are normally distributed, with σ_i^2 as the constant variance of the i^{th} predator species, and with β_j^2 as the

constant variance of the j^{th} prey species.

$$p(m_i, \bar{m}_i) = \frac{1}{\sqrt{2\pi\sigma_i^2}} \exp \left[-\frac{(m_i - \bar{m}_i)^2}{2\sigma_i^2} \right]$$

$$p(n_j, \bar{n}_j) = \frac{1}{\sqrt{2\pi\beta_j^2}} \exp \left[-\frac{(n_j - \bar{n}_j)^2}{2\beta_j^2} \right]$$

All of the species' phenotypic variances have a genetic and environment component,

$$\sigma_i^2 = \sigma_{Gi}^2 + \sigma_{Ei}^2$$

$$\beta_j^2 = \beta_{Gj}^2 + \beta_{Ej}^2$$

Let $a_{ij}(m_i, n_j)$ be the attack rate of an individual predator from species i on an individual prey from species j . Supposing the attack rate is optimal at α_{ij} when the predator's trait and prey's trait are at an optimal difference θ_{ij} , and decreases in a Gaussian manner as the trait's diverge from that difference, then

$$a_{ij}(m_i, n_j) = \alpha_{ij} \exp \left[-\frac{(m_i - n_j - \theta_{ij})^2}{2\tau_{ij}^2} \right]$$

where τ_{ij} determines how phenotypically specialized a predator individual of species i must be to use a prey individual of species j . Let $\overline{a_{ij}}(\overline{m_i}, \overline{n_j})$ be the mean attack rate of predator species i on prey species j . Thus,

$$\begin{aligned}\overline{a_{ij}}(\overline{m_i}, \overline{n_j}) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} a_{ij}(m_i, n_j) \cdot p(m_i, \overline{m_i}) \cdot p(n_j, \overline{n_j}) dm_i dn_j \\ &= \frac{\alpha_{ij} \tau_{ij}}{\sqrt{\sigma_i^2 + \beta_j^2 + \tau_{ij}^2}} \exp \left[-\frac{(\overline{m_i} - \overline{n_j} - \theta_{ij})^2}{2(\sigma_i^2 + \beta_j^2 + \tau_{ij}^2)} \right]\end{aligned}$$

Let u be the number of predator species, and let v be the number of prey species. If predators have a linear functional response, convert the consumed prey into offspring with efficiencies e_{ij} , and experience a per-capita mortality rate d_i , then the fitness of a predator with phenotype m_i is

$$W_i(m_i, [N]_1^v, [n]_1^v) = \sum_{j=1}^v (e_{ij} a_{ij}(m_i, n_j) N_j) - d_i$$

and thus the mean fitness of the i^{th} predator population is

$$\begin{aligned}
 \overline{W}_i(\overline{m}_i, [N]_1^v, [\overline{n}]_1^v) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} W_i(m_i, [N]_1^v, [n]_1^v) p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left(\sum_{j=1}^v (e_{ij} a_{ij}(m_i, n_j) N_j) - d_i \right) p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \sum_{j=1}^v e_{ij} a_{ij}(m_i, n_j) N_j p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &\quad - \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} d_i p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &= \sum_{j=1}^v (e_{ij} \overline{a}_{ij}(\overline{m}_i, \overline{n}_j) N_j) - d_i
 \end{aligned}$$

In the absence of the predators, each prey experience logistic growth with intrinsic growth rates r_j and carrying capacities K_j . Thus the fitness of a prey with phenotype n_j is

$$Y_j(N_j, n_j, [M]_1^u, [m]_1^u) = r_j \left(1 - \frac{N_j}{K_j} \right) - \sum_{i=1}^u (a_{ij}(m_i, n_j) M_i)$$

and thus the mean fitness of the j^{th} prey population is

$$\begin{aligned}
 \overline{Y}_j(N_j, \overline{n}_j, [M]_1^u, [\overline{m}]_1^u) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} Y_j(N_j, n_j, [M]_1^u, [m]_1^u) p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left(r_j \left(1 - \frac{N_j}{K_j} \right) - \sum_{i=1}^u (a_{ij}(m_i, n_j) M_i) \right) p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} r_j \left(1 - \frac{N_j}{K_j} \right) p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &\quad - \sum_{i=1}^u M_i \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} a_{ij}(m_i, n_j) p(m_i, \overline{m}_i) p(n_j, \overline{n}_j) dm_i dn_j \\
 &= r_j \left(1 - \frac{N_j}{K_j} \right) - \sum_{i=1}^u \overline{a}_{ij}(\overline{m}_i, \overline{n}_j) M_i
 \end{aligned}$$

So the ecological dynamics of the model (population densities) are given by

$$\begin{cases} \frac{dM_i}{dt} = M_i \overline{W}_i(\overline{m}_i, [N]_1^v, [\overline{n}]_1^v) \\ \frac{dN_j}{dt} = N_j \overline{Y}_j(N_j, \overline{n}_j, [M]_1^u, [\overline{m}]_1^u) \end{cases} \quad (1)$$

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Equilibria Analysis

Stability Analysis

Special Case:

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Special Case:
 $M^* = 0, N^* = K$

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$$\begin{cases} f_1 = \frac{dM}{dt} = M\bar{W}(\bar{m}, N, \bar{n}) \\ f_2 = \frac{dN}{dt} = N\bar{Y}(N, \bar{n}, M, \bar{m}) \\ f_3 = \frac{d\bar{m}}{dt} = \sigma_G^2 \frac{\partial \bar{W}}{\partial \bar{m}} \\ f_4 = \frac{d\bar{n}}{dt} = \beta_G^2 \frac{\partial \bar{Y}}{\partial \bar{n}} \end{cases} \quad (3)$$

where

$$\bar{W}(\bar{m}, N, \bar{n}) = e\bar{a}(\bar{m}, \bar{n})N - d$$

$$\bar{Y}(N, \bar{n}, M, \bar{m}) = r \left(1 - \frac{N}{K} \right) - \bar{a}(\bar{m}, \bar{n})M$$

$$\frac{\partial \bar{W}}{\partial \bar{m}} = \frac{e\alpha\tau N(\theta + \bar{n} - \bar{m})}{(\sigma^2 + \beta^2 + \tau^2)^{3/2}} \exp \left[-\frac{(\bar{m} - \bar{n} - \theta)^2}{2(\sigma^2 + \beta^2 + \tau^2)} \right]$$

$$\frac{\partial \bar{Y}}{\partial \bar{n}} = \frac{\alpha\tau M(\theta + \bar{n} - \bar{m})}{(\sigma^2 + \beta^2 + \tau^2)^{3/2}} \exp \left[-\frac{(\bar{m} - \bar{n} - \theta)^2}{2(\sigma^2 + \beta^2 + \tau^2)} \right]$$

$$f_3 = 0 \implies \bar{m} - \bar{n} = \theta \text{ or } N = 0 \quad (4)$$

$$f_4 = 0 \implies \bar{m} - \bar{n} = \theta \text{ or } M = 0 \quad (5)$$

$$f_1 = 0 \implies M = 0 \text{ or } N = \frac{d\sqrt{\sigma^2 + \beta^2 + \tau^2}}{e\alpha\tau} \exp\left[\frac{(\bar{m} - \bar{n} - \theta)^2}{2(\sigma^2 + \beta^2 + \tau^2)}\right] \quad (6)$$

$$f_2 = 0 \implies N = 0 \text{ or } M = \frac{r\sqrt{\sigma^2 + \beta^2 + \tau^2}}{\alpha\tau} \left(1 - \frac{N}{K}\right) \exp\left[\frac{(\bar{m} - \bar{n} - \theta)^2}{2(\sigma^2 + \beta^2 + \tau^2)}\right] \quad (7)$$

Clearly, $M = N = 0$ satisfies the equilibrium conditions. (7) is satisfied by $N = 0$, which, by (6), implies $M = 0$. This is intuitive because the predator can only survive if there is prey.

On the other hand, (6) is satisfied by $M = 0$, which, by (7), implies either $N = 0$ or $N = K$. This is intuitive because the prey can reach equilibrium at its carrying capacity.

For coexistence equilibria (represented by M^* and N^*), let

$\bar{m} - \bar{n} = \theta$. Then

$$\begin{cases} N^* = \frac{d\sqrt{\sigma^2 + \beta^2 + \tau^2}}{e\alpha\tau} \\ M^* = \frac{r\sqrt{\sigma^2 + \beta^2 + \tau^2}}{\alpha\tau} \left(1 - \frac{N^*}{K}\right) \end{cases}$$

Thus coexistence equilibria can be reached with the above values of N^* and M^* and any values \bar{m} and \bar{n} so long as $\bar{m} - \bar{n} = \theta$.

For local stability around the various equilibria

$E^* = (M^*, N^*, \bar{m}^*, \bar{n}^*)$, we find the Jacobian matrix:

$$J^* = J|_{E^*} = \begin{pmatrix} \left. \frac{\partial f_1}{\partial M} \right|_{E^*} & \left. \frac{\partial f_1}{\partial N} \right|_{E^*} & \left. \frac{\partial f_1}{\partial \bar{m}} \right|_{E^*} & \left. \frac{\partial f_1}{\partial \bar{n}} \right|_{E^*} \\ \left. \frac{\partial f_2}{\partial M} \right|_{E^*} & \left. \frac{\partial f_2}{\partial N} \right|_{E^*} & \left. \frac{\partial f_2}{\partial \bar{m}} \right|_{E^*} & \left. \frac{\partial f_2}{\partial \bar{n}} \right|_{E^*} \\ \left. \frac{\partial f_3}{\partial M} \right|_{E^*} & \left. \frac{\partial f_3}{\partial N} \right|_{E^*} & \left. \frac{\partial f_3}{\partial \bar{m}} \right|_{E^*} & \left. \frac{\partial f_3}{\partial \bar{n}} \right|_{E^*} \\ \left. \frac{\partial f_4}{\partial M} \right|_{E^*} & \left. \frac{\partial f_4}{\partial N} \right|_{E^*} & \left. \frac{\partial f_4}{\partial \bar{m}} \right|_{E^*} & \left. \frac{\partial f_4}{\partial \bar{n}} \right|_{E^*} \end{pmatrix}$$

The conditions for stability of E^* are equivalent to the conditions by which all roots of the characteristic polynomial of J^* have non-positive real parts (i.e. the Routh-Hurwitz criterion). First, we must calculate the partial derivatives.

$$\frac{\partial f_1}{\partial M} = \overline{W}(\overline{m}, N, \overline{n})$$

$$\frac{\partial f_1}{\partial N} = e\overline{a}(\overline{m}, \overline{n}) \cdot M$$

$$\frac{\partial f_1}{\partial \overline{m}} = \frac{e\overline{a}(\overline{m}, \overline{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot M \cdot N \cdot (\theta + \overline{n} - \overline{m})$$

$$\frac{\partial f_1}{\partial \overline{n}} = \frac{e\overline{a}(\overline{m}, \overline{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot M \cdot N \cdot (\overline{m} - \overline{n} - \theta)$$

$$\frac{\partial f_2}{\partial M} = -\overline{a}(\overline{m}, \overline{n}) \cdot N$$

$$\frac{\partial f_2}{\partial N} = \overline{Y}(N, \overline{n}, M, \overline{m}) - \frac{Nr}{K}$$

$$\frac{\partial f_2}{\partial \overline{m}} = \frac{\overline{a}(\overline{m}, \overline{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot M \cdot N \cdot (\overline{m} - \overline{n} - \theta)$$

$$\frac{\partial f_2}{\partial \overline{n}} = \frac{\overline{a}(\overline{m}, \overline{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot M \cdot N \cdot (\theta + \overline{n} - \overline{m})$$

$$\frac{\partial f_3}{\partial M} = 0$$

$$\frac{\partial f_3}{\partial N} = \frac{\sigma_G^2 e \bar{a}(\bar{m}, \bar{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot (\theta + \bar{n} - \bar{m})$$

$$\frac{\partial f_3}{\partial \bar{m}} = \frac{\sigma_G^2 e \bar{a}(\bar{m}, \bar{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot N \left(\frac{(\bar{m} - \bar{n} - \theta)^2}{\sigma^2 + \beta^2 + \tau^2} - 1 \right)$$

$$\frac{\partial f_3}{\partial \bar{n}} = \frac{\sigma_G^2 e \bar{a}(\bar{m}, \bar{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot N \left(1 - \frac{(\bar{m} - \bar{n} - \theta)^2}{\sigma^2 + \beta^2 + \tau^2} \right)$$

$$\frac{\partial f_4}{\partial M} = \frac{\beta_G^2 \bar{a}(\bar{m}, \bar{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot (\bar{n} - \bar{m})$$

$$\frac{\partial f_4}{\partial N} = 0$$

$$\frac{\partial f_4}{\partial \bar{m}} = \frac{\beta_G^2 \bar{a}(\bar{m}, \bar{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot M \left(\frac{(\bar{m} - \bar{n} - \theta)^2}{\sigma^2 + \beta^2 + \tau^2} - 1 \right)$$

$$\frac{\partial f_4}{\partial \bar{n}} = \frac{\beta_G^2 \bar{a}(\bar{m}, \bar{n})}{\sigma^2 + \beta^2 + \tau^2} \cdot M \left(1 - \frac{(\bar{m} - \bar{n} - \theta)^2}{\sigma^2 + \beta^2 + \tau^2} \right)$$

$E^* = (0, 0, \bar{m}^*, \bar{n}^*)$ where \bar{m}^* and \bar{n}^* are arbitrary values. Then

$$J^* = J|_{E^*} = \begin{pmatrix} -d & 0 & 0 & 0 \\ 0 & r & 0 & 0 \\ 0 & 0 & \frac{\sigma_G^2 e \bar{a}(\bar{m}^*, \bar{n}^*)}{\sigma^2 + \beta^2 + \tau^2} \cdot (\theta + \bar{n}^* - \bar{m}^*) & 0 \\ \frac{\beta_G^2 \bar{a}(\bar{m}^*, \bar{n}^*)}{\sigma^2 + \beta^2 + \tau^2} \cdot (\theta + \bar{n}^* - \bar{m}^*) & 0 & 0 & 0 \end{pmatrix}$$

Since J^* is a lower-triangular matrix, its eigenvalues are its diagonal entries: $-d$, r , 0 , and 0 . Since r is positive, this equilibrium is locally unstable.

$E^* = (0, K, \bar{m}^*, \bar{n}^*)$ where \bar{m}^* and \bar{n}^* are arbitrary values. Then

$$J^* = J|_{E^*} = \begin{pmatrix} e \bar{a}(\bar{m}^*, \bar{n}^*) K - d & 0 & 0 & 0 \\ 0 & -r & 0 & 0 \\ 0 & j_{32} & j_{33} & j_{34} \\ j_{41} & 0 & 0 & 0 \end{pmatrix}$$

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where

$$j_{32} = \frac{\sigma_G^2 e \bar{a}(\bar{m}^*, \bar{n}^*)}{\sigma^2 + \beta^2 + \tau^2} \cdot (\theta + \bar{n}^* - \bar{m}^*)$$

$$j_{33} = \frac{\sigma_G^2 e \bar{a}(\bar{m}^*, \bar{n}^*)}{\sigma^2 + \beta^2 + \tau^2} \cdot K \left(\frac{(\bar{m}^* - \bar{n}^* - \theta)^2}{\sigma^2 + \beta^2 + \tau^2} - 1 \right)$$

$$j_{34} = \frac{\sigma_G^2 e \bar{a}(\bar{m}^*, \bar{n}^*)}{\sigma^2 + \beta^2 + \tau^2} \cdot K \left(1 - \frac{(\bar{m}^* - \bar{n}^* - \theta)^2}{\sigma^2 + \beta^2 + \tau^2} \right)$$

$$j_{41} = \frac{\beta_G^2 \bar{a}(\bar{m}^*, \bar{n}^*)}{\sigma^2 + \beta^2 + \tau^2} \cdot (\theta + \bar{n}^* - \bar{m}^*)$$

By reordering the variables $E^{**} = (M^*, N^*, \bar{n}^*, \bar{m}^*)$, we can force J^* to be a lower-triangular matrix, and hence it's eigenvalues are its diagonal entries:

$$J^{**} = J|_{E^{**}} = \begin{pmatrix} e \bar{a}(\bar{m}^*, \bar{n}^*) K - d & 0 & 0 & 0 \\ 0 & -r & 0 & 0 \\ j_{41} & 0 & 0 & 0 \\ 0 & j_{32} & j_{34} & j_{33} \end{pmatrix}$$

Thus the eigenvalues are $e\bar{a}(\bar{m}, \bar{n})K - d$, $-r$, 0 , and j_{33} . Thus E^* is stable if the following hold:

$$d > e\bar{a}(\bar{m}^*, \bar{n}^*)K, \quad \text{and}$$

$$(\bar{m}^* - \bar{n}^* - \theta)^2 < \sigma^2 + \beta^2 + \tau^2$$

E^* is unstable if either of the above fails.

$E^* = (M^*, N^*, \mu^*, \mu^*)$ where μ^* is an arbitrary value. Then

$$J^* = J|_{E^*} = \begin{pmatrix} 0 & er\left(1 - \frac{N^*}{K}\right) & 0 & 0 \\ -\frac{d}{e} & -\frac{rN^*}{K} & 0 & 0 \\ 0 & 0 & -\frac{d\sigma_G^2}{\sigma^2 + \beta^2 + \tau^2} & \frac{d\sigma_G^2}{\sigma^2 + \beta^2 + \tau^2} \\ 0 & 0 & -\frac{r\beta_G^2\left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} & \frac{r\beta_G^2\left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} \end{pmatrix}$$

The characteristic polynomial is

$$P(\lambda) = |\lambda I - J^*| = \begin{vmatrix} \lambda - er \left(1 - \frac{N^*}{K}\right) & 0 & 0 \\ \frac{d}{e} & \lambda + \frac{rN^*}{K} & 0 \\ 0 & 0 & \lambda + \frac{d\sigma_G^2}{\sigma^2 + \beta^2 + \tau^2} - \frac{r\beta_G^2 \left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} \end{vmatrix}$$

Thus,

$$P(\lambda) = \begin{vmatrix} \lambda - er \left(1 - \frac{N^*}{K}\right) \\ \frac{d}{e} & \lambda + \frac{rN^*}{K} \end{vmatrix} \cdot \begin{vmatrix} \lambda + \frac{d\sigma_G^2}{\sigma^2 + \beta^2 + \tau^2} - \frac{d\sigma_G^2}{\sigma^2 + \beta^2 + \tau^2} \\ \frac{r\beta_G^2 \left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} & \lambda - \frac{r\beta_G^2 \left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} \end{vmatrix}$$

$$= P_1(\lambda) \cdot P_2(\lambda)$$

Thus the zeros of $P(\lambda)$ are the zeros of both $P_1(\lambda)$ and $P_2(\lambda)$.

$$P_1(\lambda) = \lambda^2 + \frac{rN^*}{K}\lambda + rd\left(1 - \frac{N^*}{K}\right) = 0$$
$$\Rightarrow \lambda_{1,2} = \frac{1}{2} \left[-\frac{rN^*}{K} \pm \sqrt{\Delta} \right]$$

Where $\Delta = \left(\frac{rN^*}{K}\right)^2 - 4rd\left(1 - \frac{N^*}{K}\right)$. Since $N^* < K$, $\sqrt{\Delta} < \left|\frac{rN^*}{K}\right|$. Thus $\text{Re}(\lambda_{1,2}) < 0$.

$$P_2(\lambda) = \lambda^2 + \left(\frac{d\sigma_G^2 - r\beta^2\left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} \right) \lambda + \left(\frac{rd\sigma_G^2\beta_G^2\left(1 - \frac{N^*}{K}\right)}{(\sigma^2 + \beta^2 + \tau^2)^2} \right) = 0$$
$$\Rightarrow \lambda_{3,4} = \frac{1}{2} \left[- \left(\frac{d\sigma_G^2 - r\beta^2\left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} \right) \pm \sqrt{\Delta} \right]$$

Where

$$\Delta = \left(\frac{d\sigma_G^2 - r\beta^2 \left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} \right)^2 - \left(\frac{4rd\sigma_G^2\beta_G^2 \left(1 - \frac{N^*}{K}\right)}{(\sigma^2 + \beta^2 + \tau^2)^2} \right)$$

Again, since $N^* < K$, $\sqrt{\Delta} < \left| \left(\frac{d\sigma_G^2 - r\beta^2 \left(1 - \frac{N^*}{K}\right)}{\sigma^2 + \beta^2 + \tau^2} \right) \right|$. Thus

$\text{Re}(\lambda_{3,4}) < 0 \iff d\sigma_G^2 > r\beta_G^2 \left(1 - \frac{N^*}{K}\right)$. So the coexistence equilibrium is stable if

$$d\sigma_G^2 > r\beta_G^2 \left(1 - \frac{N^*}{K}\right)$$