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

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### Overview

- Motivation / Observations in Nature
- Model Formulation
- Preliminary Results
- Future Work

### Motivation / Observations in Nature

- Predator/Prey interactions are prevalent in nature
  - Crab vs. gastropod [Saloniemi, 1993]
  - Protist vs. bacteria [terHorst]
- There is trait variation within species
  - Thickness of plant cuticula [Saloniemi, 1993]
  - Strength of gastropod shell [Saloniemi, 1993]
- Incorporating trait variation provides richer dynamics than classical Lotka-Volterra models



### Normally Distributed Trait Values

Assume Prey and Predator have normally distributed trait values.

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

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

#### **Variables**

- $n \equiv \text{Prey Trait Value}$
- $\overline{n} \equiv$  Average Prey Trait Value
- $m \equiv \text{Predator Trait Value}$
- $\overline{m} \equiv$  Average Predator Trait Value

- $\beta^2 \equiv \text{Prey Trait Variance}$
- $\sigma^2 \equiv \text{Predator Trait Variance}$

### Attack Rate as a function of Normally Distributed Trait Values

## Attack Rate is a Function of the Prey's Trait Value and the Predator's Trait Value

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

#### **Variables**

- $n \equiv \text{Prey Trait Value}$
- $\bullet$   $\overline{n} \equiv$  **Average** Prey Trait Value
- $m \equiv \text{Predator Trait Value}$
- $\overline{m} \equiv$  **Average** Predator Trait Value

- $\alpha \equiv \text{Maximum attack rate}$
- ullet  $\theta \equiv \text{Optimal trait difference}$
- $\tau^2 \equiv \text{Specialization Constant}$

### Attack Rate as a function of Normally Distributed Trait Values

## Attack Rate is a Function of the Prey's Trait Value and the Predator's Trait Value

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

### Average Attack Rate

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

#### **Variables**

- $n \equiv \text{Prey Trait Value}$
- ullet  $\overline{n} \equiv$  **Average** Prey Trait Value
- $m \equiv \text{Predator Trait Value}$
- $\overline{m} \equiv$  Average Predator Trait Value

- $\alpha \equiv \text{Maximum attack rate}$
- ullet  $\theta \equiv \text{Optimal trait difference}$
- $\tau^2 \equiv \text{Specialization Constant}$
- $oldsymbol{\circ}$   $\beta^2 \equiv \operatorname{Prey}$  Trait Variance
- $\sigma^2 \equiv \text{Predator Trait Variance}$

### Intrinsic Growth Rate as a function of the Normally Distributed Trait Value

### Prey Growth Rate is is a Function of the Prey's Trait Value

$$r(n) = \rho \exp \left[ -\frac{(n-\phi)^2}{2\gamma^2} \right]$$

#### **Variables**

- $n \equiv \text{Prey Trait Value}$
- $\overline{n} \equiv$  Average Prey Trait Value

- $\rho \equiv Maximum Growth Rate$
- ullet  $\phi \equiv \operatorname{Prey}$  Optimum Trait Value
- $\gamma^2 \equiv$  Stabilizing Selection Constant

### Intrinsic Growth Rate as a function of the Normally Distributed Trait Value

### Prey Growth Rate is is a Function of the Prey's Trait Value

$$r(n) = \rho \exp \left[ -\frac{(n-\phi)^2}{2\gamma^2} \right]$$

### Averge Growth Rate

$$\overline{r}(\overline{n}) = \int_{-\infty}^{\infty} r(n) \cdot p(n, \overline{n}) dn$$

$$= \frac{\rho \gamma}{\sqrt{\beta^2 + \gamma^2}} \exp\left[-\frac{(n - \phi)^2}{2\gamma^2}\right]$$

### **Variables**

- $n \equiv \text{Prey Trait Value}$
- $\overline{n} \equiv$  Average Prey Trait Value

- $\rho \equiv Maximum Growth Rate$
- $\phi \equiv \text{Prey Optimum Trait Value}$
- $\gamma^2 \equiv$  Stabilizing Selection Constant
- $\beta^2 \equiv \text{Prey Trait Variance}$

### Fitness Assumptions

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

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

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

#### **Variables**

- N ≡ Prey Density
- $n \equiv \text{Prey Trait Value}$
- $M \equiv \text{Predator Density}$
- m ≡ Predator Trait Value

- $r \equiv$  Intrinsic Prey Growth Rate Function
- $K \equiv \text{Prey Carrying Capacity}$
- $d \equiv \text{Predator Death Rate}$
- $e \equiv \text{Efficiency}$

### Average Fitness

$$\overline{Y}(N, \overline{n}, M, \overline{m}) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} Y(N, n, M, m) \cdot p(m, \overline{m}) \cdot p(n, \overline{n}) \, dm dn$$

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

$$\overline{W}(N, \overline{n}, M, \overline{m}) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} W(N, n, M, m) \cdot p(m, \overline{m}) \cdot p(n, \overline{n}) \, dm dn$$

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

#### **Variables**

- N ≡ Prey Density
- $\overline{n} \equiv$  Average Prey Trait Value
- $M \equiv \text{Predator Density}$
- m 

  Average Predator Trait
   Value

- $\bar{r} \equiv$  Average Intrinsic Prey Growth Rate Function
- $K \equiv \text{Prey Carrying Capacity}$
- $d \equiv \text{Predator Death Rate}$
- e ≡ Efficiency → ← □ → ← □ → □ ▼ → へへ

### **Ecological Components**

$$\begin{split} \frac{dN}{dt} &= N \cdot \overline{Y}(N, \overline{n}, M, \overline{m}) = N \bigg[ \overline{r}(\overline{n}) \left( 1 - \frac{N}{K} \right) - M \overline{a}(\overline{n}, \overline{m}) \bigg] \\ \frac{dM}{dt} &= M \cdot \overline{W}(N, \overline{n}, M, \overline{m}) = M [eN \overline{a}(\overline{n}, \overline{m}) - d] \end{split}$$

$$\xrightarrow{\overline{r(\overline{n})N(1-\frac{N}{K})}} N \xrightarrow{\overline{a}(\overline{n},\overline{m})MN} - \xrightarrow{e\overline{a}(\overline{n},\overline{m})MN} M$$

#### **Variables**

- N ≡ Prey Density
- $\bar{n} \equiv$  **Average** Prey Trait Value
- $M \equiv \text{Predator Density}$
- $\overline{m} \equiv \text{Average}$  Predator Trait Value

- $\bar{r} \equiv$  Average Intrinsic Prey Growth Rate Function
- $\bullet$   $K \equiv$  Prey Carrying Capacity
- $d \equiv \text{Predator Death Rate}$

### **Evolutionary Components**

 The evolution of the average trait value is always in the direction which increases the mean fitness in the population. [Lande, 1976]

$$\begin{split} \frac{d\overline{n}}{dt} &= \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}} = \beta_G^2 \left[ \overline{r}(\overline{n}) \left( 1 - \frac{N}{K} \right) \frac{(\phi - \overline{n})}{\beta^2 + \gamma^2} + \frac{M(\theta - (\overline{m} - \overline{n}))}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n}) \right] \\ \frac{d\overline{m}}{dt} &= \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}} = \sigma_G^2 \frac{eN(\theta + \overline{n} - \overline{m})}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n}) \end{split}$$

#### **Variables**

- N ≡ Prey Density
- $\overline{n} \equiv$  **Average** Prey Trait Value
- M ≡ Predator Density
- $\overline{m} \equiv$  Average Predator Trait Value

- $\phi \equiv \text{Prey Optimum Trait Value}$
- $\gamma^2 \equiv$  Stabilizing Selection Constant
- $\overline{r} \equiv$  Average Intrinsic Prey Growth Rate Function
- $\beta_G^2 \equiv \text{Prey genetic variance}$
- $\sigma_G^2 \equiv$  Predator genetic variance



# The Complete $1 \times 1$ Model (One Predator Species, One Prey Species)

### **Ecological Components**

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

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

### **Evolutionary Components**

$$\begin{split} \frac{d\overline{n}}{dt} &= \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}} = \beta_G^2 \left[ \overline{r}(\overline{n}) \left( 1 - \frac{N}{K} \right) \frac{(\phi - \overline{n})}{\beta^2 + \gamma^2} + \frac{M(\theta - (\overline{m} - \overline{n}))}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n}) \right] \\ \frac{d\overline{m}}{dt} &= \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}} = \sigma_G^2 \frac{eN(\theta + \overline{n} - \overline{m})}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n}) \end{split}$$

$$\frac{dN}{dt} = N \cdot \overline{Y}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}}$$

$$\frac{dM}{dt} = M \cdot \overline{W}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

#### **Extinction**

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (0, 0, \underline{\hspace{1em}}, \underline{\hspace{1em}})$$

#### **Exclusion**

$$(N^*,M^*,\overline{n}^*,\overline{m}^*)=(K,0,\_,\_)$$

$$\frac{dN}{dt} = N \cdot \overline{Y}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}}$$

$$\frac{dM}{dt} = M \cdot \overline{W}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

### **Extinction** *Unstable*

$$(N^*,M^*,\overline{n}^*,\overline{m}^*)=(0,0,\_,\_)$$

#### **Exclusion**

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (K, 0, \_, \_)$$

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**Extinction** *Unstable* 

$$(N^*,M^*,\overline{n}^*,\overline{m}^*)=(0,0,\_,\_)$$

**Exclusion** Locally stable under certain conditions

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (K, 0, \_, \_)$$

$$\frac{dN}{dt} = N \cdot \overline{Y}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}}$$

$$\frac{dM}{dt} = M \cdot \overline{W}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

**Extinction** *Unstable* 

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (0, 0, \underline{\hspace{1em}}, \underline{\hspace{1em}})$$

**Exclusion** Locally stable under certain conditions

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (K, 0, \underline{\hspace{1em}}, \underline{\hspace{1em}})$$

**Necessary Conditions for Locally Stable Exclusion:** 

- $d > e\overline{a}(\overline{m}^*, \overline{n}^*)K$
- $(\overline{m}^* \overline{n}^* \theta)^2 < \sigma^2 + \beta^2 + \tau^2$

$$\frac{dN}{dt} = N \cdot \overline{Y}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial Y}{\partial \overline{n}}$$

$$\frac{dM}{dt} = M \cdot \overline{W}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

#### Coexistence

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = \left(\frac{d\sqrt{A}}{e\alpha\tau} , \frac{\rho\gamma\sqrt{A}}{\alpha\tau\sqrt{B}} \left(1 - \frac{d\sqrt{A}}{Ke\alpha\tau}\right) , \theta , \theta + \phi\right)$$
 where  $A = \sigma^2 + \beta^2 + \tau^2$  and  $B = \beta^2 + \gamma^2$ 

$$\frac{dN}{dt} = N \cdot \overline{Y}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial Y}{\partial \overline{n}}$$

$$\frac{dM}{dt} = M \cdot \overline{W}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

**Coexistence** | Locally stable under certain conditions

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (\frac{d\sqrt{A}}{e\alpha\tau} , \frac{\rho\gamma\sqrt{A}}{\alpha\tau\sqrt{B}} \left(1 - \frac{d\sqrt{A}}{Ke\alpha\tau}\right) , \theta , \theta + \phi)$$
 where  $A = \sigma^2 + \beta^2 + \tau^2$  and  $B = \beta^2 + \gamma^2$ 

$$\frac{dN}{dt} = N \cdot \overline{Y}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}} 
\frac{dM}{dt} = M \cdot \overline{W}(N, \overline{n}, M, \overline{m}) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

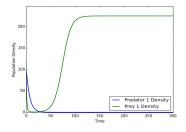
**Coexistence** | Locally stable under certain conditions

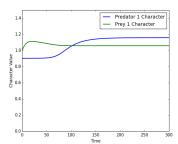
$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = \left(\frac{d\sqrt{A}}{e\alpha\tau}, \frac{\rho\gamma\sqrt{A}}{\alpha\tau\sqrt{B}}\left(1 - \frac{d\sqrt{A}}{Ke\alpha\tau}\right), \theta, \theta + \phi\right)$$
where  $A = \sigma^2 + \beta^2 + \tau^2$  and  $B = \beta^2 + \gamma^2$ 

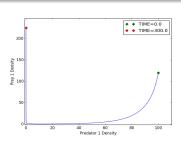
**Necessary Condition for Locally Stable Coexistence:** 

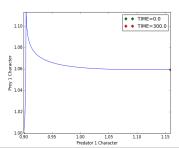
$$\bullet \ \frac{\sigma_{\mathsf{G}}^2}{\beta_{\mathsf{G}}^2} > \frac{\rho \gamma}{d\sqrt{B}} \left( 1 - \frac{d\sqrt{A}}{\mathsf{K} e \alpha \tau} \right) \left( 1 - \frac{A}{B} \right)$$

### Figures - $1 \times 1$ - Stable Exclusion



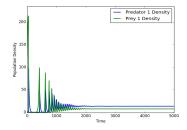


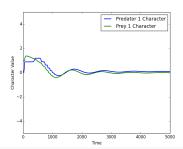


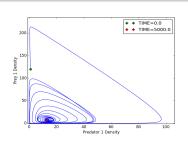


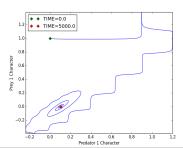


### Figures - $1 \times 1$ - Stable Coexistence



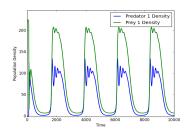


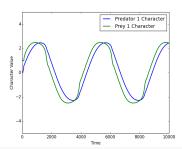


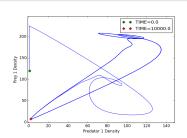


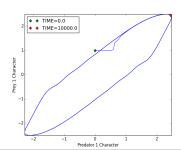


### Figures - 1 × 1 - Stable Cycles (Red Queen Dynamics) [Kindrik, Kondrashov, 1994]











### **Prey Fitness**

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

#### **Predator Fitness**

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

### **Prey Fitness**

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

$$\downarrow$$

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

#### **Predator Fitness**

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

#### Notation

$$[x_i]_{i=1}^u = x_1, \dots, x_u$$

### **Prey Fitness**

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

$$\downarrow$$

$$Y_{j}(N_{j}, n_{j}, [M_{i}]_{i=1}^{u}, [m_{i}]_{i=1}^{u}) = r_{j}(n_{j}) \left( 1 - \frac{N_{j}}{K_{i}} \right) - \sum_{i=1}^{u} M_{i} a_{ij}(n_{j}, m_{i})$$

#### **Predator Fitness**

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

$$\downarrow$$

$$W_i([N_j]_{j=1}^{\nu}, [n_j]_{j=1}^{\nu}, M_i, m_i) = \sum_{j=1}^{\nu} \left[ e_{ij} N_j a_{ij}(n_j, m_i) \right] - d_i$$

#### **Notation**

$$[x_i]_{i=1}^u = x_1, \ldots, x_u$$

### Average Fitness

$$\begin{split} \overline{Y}_{j}(N_{j}, \overline{n_{j}}, [M_{i}]_{i=1}^{u}, [\overline{m_{i}}]_{i=1}^{u}) \\ &= \int_{\mathbb{R}^{u+1}} Y_{j} \cdot \prod_{i=1}^{u} \left[ p_{i}(m_{i}, \overline{m_{i}}) \right] \cdot p(n, \overline{n}) \prod_{i=1}^{u} \left[ dm_{i} \right] dn_{j} \\ &= \overline{r_{j}}(\overline{n_{j}}) \left( 1 - \frac{N_{j}}{K_{j}} \right) - \sum_{i=1}^{u} M_{i} \overline{a}_{ij}(\overline{n}_{j}, \overline{m}_{i}) \end{split}$$

$$\begin{split} \overline{W}_{i}(N_{j}, \overline{n_{j}}, [M_{i}]_{i=1}^{u}, [\overline{m_{i}}]_{i=1}^{u}) \\ &= \int_{\mathbb{R}^{u+1}} W_{i} \cdot p_{i}(m_{i}, \overline{m_{i}}) \cdot \prod_{j=1}^{v} \left[ p(n_{j}, \overline{n}_{j}) \right] dm_{i} \prod_{j=1}^{v} \left[ dn_{j} \right] \\ &= \sum_{i=1}^{v} \left[ e_{ij} N_{j} \overline{a}_{ij} (\overline{n}_{j}, \overline{m}_{i}) \right] - d_{i} \end{split}$$

# The Complete $u \times v$ Model (u Predator Species, v Prey Species)

### **Ecological Components**

$$\frac{dN_{j}}{dt} = N_{j} \left[ \overline{r_{j}}(\overline{n_{j}}) \left( 1 - \frac{N_{j}}{K_{j}} \right) - \sum_{i=1}^{u} M_{i} \overline{a}_{ij}(\overline{n}_{j}, \overline{m}_{i}) \right]$$

$$\frac{dM_{i}}{dt} = M_{i} \left[ \sum_{j=1}^{v} \left[ e_{ij} N_{j} \overline{a}_{ij}(\overline{m}_{i}, \overline{n}_{j}) \right] - d_{i} \right]$$

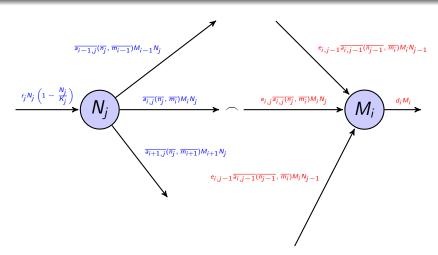
### **Evolutionary Components**

$$\frac{d\overline{n}_{j}}{dt} = \beta_{Gj}^{2} \left[ \overline{r_{j}} (\overline{n_{j}}) \left( 1 - \frac{N_{j}}{K_{j}} \right) \frac{(\phi_{j} - \overline{n_{j}})}{\beta_{j}^{2} + \gamma_{j}^{2}} + \sum_{i=1}^{u} \left[ \frac{M_{i} (\theta_{ij} + \overline{n_{j}} - \overline{m_{i}})}{\sigma_{i}^{2} + \beta_{j}^{2} + \tau_{ij}^{2}} \overline{a}_{ij} (\overline{m_{i}}, \overline{n_{j}}) \right] \right] d\overline{m}_{i}$$

$$2 \sum_{i=1}^{v} \left[ e_{ij} N_{j} (\theta_{ij} + \overline{n_{j}} - \overline{m_{i}}) - (\overline{m_{i}}) \right] d\overline{m}_{i}$$

$$\frac{d\overline{m}_i}{dt} = \sigma_{Gi}^2 \sum_{i=1}^{\nu} \left[ \frac{e_{ij} N_j (\theta_{ij} + \overline{n_j} - \overline{m_i})}{\sigma_i^2 + \beta_j^2 + \tau_{ij}^2} \overline{a}_{ij} (\overline{m_i}, \overline{n_j}) \right]$$

# The Complete $u \times v$ Model (u Predator Species, v Prey Species)



### Future Work

- Two Predators competing for One Prey
- One Specialist Predator Competing with One Generalist Predator for Two Prey Species
- Two Specialist Predators Competing with One Generalist Predator for Two Prey Species
- Further Analysis of the General  $u \times v$  Model
- Intra-Guild Predation
- Adding Evolutionary Cost to Prey
- Adding Evolutionary Cost to Predator

### Thank You!

- Pacific Coast Undergraduate Math Conference
- Dr. Alissa Crans, Dr. Karrolyne Fogel, Dr. Kendra Killpatrick, Dr. John Rock, and all other PCUMC Organizers
- National Science Foundation
- Mathematical Association of America and all other PCUMC sponsors
- Cal Lutheran University and all other PCUMC university supporters
- Dr. Helena Noronha
- Pacific Math Alliance PUMP Undergraduate Research Groups
- California State University, Northridge
- Dr. Jing Li and Dr. Casey terHorst

## Questions?



$$\frac{dN_1}{dt} = N_1 \cdot \overline{Y}_1(\overline{m}, \overline{n}_1, M, N_1) \qquad \qquad \frac{d\overline{n}_1}{dt} = \beta_{G1}^2 \frac{\partial Y_1}{\partial \overline{n}_1} \\
\frac{dN_2}{dt} = N_2 \cdot \overline{Y}_2(\overline{m}, \overline{n}_2, M, N_2) \qquad \qquad \frac{d\overline{n}_2}{dt} = \beta_{G2}^2 \frac{\partial \overline{Y}_2}{\partial \overline{n}_2} \\
\frac{dM}{dt} = M \cdot \overline{W}(\overline{m}, \overline{n}_1, \overline{n}_2, N_1, N_2) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

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#### Extinction

$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*) = (0, 0, 0, \_, \_, \_)$$

$$\frac{dN_1}{dt} = N_1 \cdot \overline{Y}_1(\overline{m}, \overline{n}_1, M, N_1) \qquad \qquad \frac{d\overline{n}_1}{dt} = \beta_{G1}^2 \frac{\partial \overline{Y}_1}{\partial \overline{n}_1} \\
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\frac{dM}{dt} = M \cdot \overline{W}(\overline{m}, \overline{n}_1, \overline{n}_2, N_1, N_2) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

**Extinction** *Unstable* 

$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*) = (0, 0, 0, \_, \_, \_)$$

$$\begin{split} \frac{dN_1}{dt} &= N_1 \cdot \overline{Y}_1(\overline{m}, \overline{n}_1, M, N_1) & \frac{d\overline{n}_1}{dt} &= \beta_{G1}^2 \frac{\partial \overline{Y}_1}{\partial \overline{n}_1} \\ \frac{dN_2}{dt} &= N_2 \cdot \overline{Y}_2(\overline{m}, \overline{n}_2, M, N_2) & \frac{d\overline{n}_2}{dt} &= \beta_{G2}^2 \frac{\partial \overline{Y}_2}{\partial \overline{n}_2} \\ \frac{dM}{dt} &= M \cdot \overline{W}(\overline{m}, \overline{n}_1, \overline{n}_2, N_1, N_2) & \frac{d\overline{m}}{dt} &= \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}} \end{split}$$

### **Extinction** *Unstable*

$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*) = (0, 0, 0, \_, \_, \_)$$

#### **Exclusion**

$$(N_1^*,N_2^*,M^*,\overline{n}_1^*,\overline{n}_2^*,\overline{m}^*)=(K_1,K_2,0,\_,\_,\_)$$



$$\frac{dN_1}{dt} = N_1 \cdot \overline{Y}_1(\overline{m}, \overline{n}_1, M, N_1) \qquad \qquad \frac{d\overline{n}_1}{dt} = \beta_{G1}^2 \frac{\partial Y_1}{\partial \overline{n}_1} \\
\frac{dN_2}{dt} = N_2 \cdot \overline{Y}_2(\overline{m}, \overline{n}_2, M, N_2) \qquad \qquad \frac{d\overline{n}_2}{dt} = \beta_{G2}^2 \frac{\partial \overline{Y}_2}{\partial \overline{n}_2} \\
\frac{dM}{dt} = M \cdot \overline{W}(\overline{m}, \overline{n}_1, \overline{n}_2, N_1, N_2) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

**Extinction** *Unstable* 

$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*) = (0, 0, 0, \_, \_, \_)$$

**Exclusion** Stable under certain conditions

$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*) = (K_1, K_2, 0, \_, \_, \_)$$



$$\frac{dN_1}{dt} = N_1 \cdot \overline{Y}_1(\overline{m}, \overline{n}_1, M, N_1) \qquad \qquad \frac{d\overline{n}_1}{dt} = \beta_{G1}^2 \frac{\partial Y_1}{\partial \overline{n}_1} \\
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### Generalist Becomes Specialist

$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*)$$

$$= \left(\frac{d\sqrt{A_1}}{e_1\alpha_1\tau_1}, K_2, \frac{r_1\sqrt{A_1}}{\alpha_1\tau_1}\left(1 - \frac{d\sqrt{A_1}}{K_1e_1\alpha_1\tau_1}\right), \mu_1^*, \mu_2^*, \mu_1^* - \theta_1\right)$$

where  $A_1 = \sigma^2 + \beta_1^2 + \tau_1^2$ ,  $\mu_1^*$  is an arbitrary value, and  $\mu_2^*$  is sufficiently far from  $\mu_1^* - \theta_1$ .

$$\frac{dN_1}{dt} = N_1 \cdot \overline{Y}_1(\overline{m}, \overline{n}_1, M, N_1) \qquad \qquad \frac{d\overline{n}_1}{dt} = \beta_{G1}^2 \frac{\partial Y_1}{\partial \overline{n}_1} \\
\frac{dN_2}{dt} = N_2 \cdot \overline{Y}_2(\overline{m}, \overline{n}_2, M, N_2) \qquad \qquad \frac{d\overline{n}_2}{dt} = \beta_{G2}^2 \frac{\partial \overline{Y}_2}{\partial \overline{n}_2} \\
\frac{dM}{dt} = M \cdot \overline{W}(\overline{m}, \overline{n}_1, \overline{n}_2, N_1, N_2) \qquad \qquad \frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}}$$

Generalist Becomes Specialist | Stable under certain conditions???

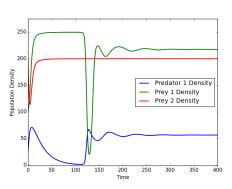
$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*)$$

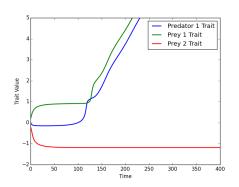
$$= \left(\frac{d\sqrt{A_1}}{e_1\alpha_1\tau_1}, K_2, \frac{r_1\sqrt{A_1}}{\alpha_1\tau_1}\left(1 - \frac{d\sqrt{A_1}}{K_1e_1\alpha_1\tau_1}\right), \mu_1^*, \mu_2^*, \mu_1^* - \theta_1\right)$$

where  $A_1 = \sigma^2 + \beta_1^2 + \tau_1^2$ ,  $\mu_1^*$  is an arbitrary value, and  $\mu_2^*$  is sufficiently far from  $\mu_1^* - \theta_1$ .

### Figures - $1 \times 2$

### **Generalist Becomes Specialist**





### Figures - $1 \times 2$

### **Unstable Coexistence**

