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

Sam Fleischer, Pablo Chavarria

March 14, 2015

#### Pacific Coast Undergraduate Mathematics Conference

Advisor: Dr. Jing Li, Mathematics, CSU Northridge Consultant: Dr. Casey terHorst, Biology, Supported By: Pacific Math Alliance, PUMP, CSU Northridge

#### Overview

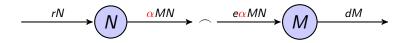
- Motivation / Observations in Nature
- Model Formulation
  - Classical Lotka-Volterra Predator-Prey Model
  - Schreiber, Bürger, and Bolnick's Extension
  - Our Extension
- Preliminary Results
- Future Work

## 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



### Classical Lotka-Volterra Model

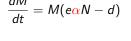


#### **Variables**

- $N \equiv \text{Prey Density}$
- $M \equiv \text{Predator Density}$

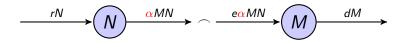
#### **Parameters**

- $\alpha \equiv$  Attack rate
- $r \equiv \text{Prey birth rate}$
- $e \equiv \text{Efficiency}$
- $d \equiv \text{Predator death rate}$



 $\frac{dN}{dt} = N(r - \frac{\alpha}{\alpha}M)$ 

### Classical Lotka-Volterra Model

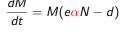


#### **Variables**

- $N \equiv \text{Prey Density}$
- $M \equiv \text{Predator Density}$

#### **Parameters**

- $\alpha \equiv$  Attack rate  $\leftarrow$  No variation!
- $r \equiv \text{Prey birth rate}$
- $e \equiv \text{Efficiency}$
- $d \equiv \text{Predator death rate}$



 $\frac{dN}{dt} = N(r - \frac{\alpha}{\alpha}M)$ 

## Schreiber, Bürger, and Bolnick's Extension

Assume the Predator Species has a normally distributed trait value.

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

#### **Parameters**

#### **Variables**

•  $m \equiv \text{Predator Trait Value}$ 

• 
$$\sigma^2 \equiv \text{Predator Trait Variance}$$

## Schreiber, Bürger, and Bolnick's Extension

Assume the Predator Species has a normally distributed trait value.

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

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

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

#### **Parameters**

#### **Variables**

•  $m \equiv \text{Predator Trait Value}$ 

- $\sigma^2 \equiv \text{Predator Trait Variance}$
- $\alpha \equiv Maximum attack rate$
- $\tau \equiv$  Specialization Constant
- $\bullet$   $\theta \equiv Optimal trait value$

## Schreiber, Bürger, and Bolnick's Extension

Assume the Predator Species has a normally distributed trait value.

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

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

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

#### **Parameters**

#### **Variables**

- $m \equiv \text{Predator Trait Value}$
- (((No Prey Trait Value)))

- $\sigma^2 \equiv \text{Predator Trait Variance}$
- $\alpha \equiv Maximum attack rate$
- $\tau \equiv$  Specialization Constant
- $\theta \equiv \text{Optimal trait value}$

↑ No variation!

### Our Extension

Assume Prey and Predator have normally distributed trait values.

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

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

#### **Variables**

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

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

#### Our Extension

Assume Prey and Predator have normally distributed trait values.

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

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

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

#### **Variables**

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

- $\beta^2 \equiv \text{Prey Trait Variance}$
- $\sigma^2 \equiv \text{Predator Trait Variance}$
- $\quad \bullet \ \, \alpha \equiv \, \mathsf{Maximum} \,\, \mathsf{attack} \,\, \mathsf{rate} \,\,$
- $\theta \equiv \text{Optimal trait difference}$
- $\tau \equiv \text{Specialization Constant}$

## 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**

- $\overline{n} \equiv$  Mean Prey Character
- $\overline{m} \equiv \text{Mean Predator}$ Character

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

## Fitness Assumptions

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

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

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

#### **Variables**

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

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



## Average Fitness

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

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

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

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

#### **Variables**

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

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

## **Ecological Components**

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

$$\begin{array}{c}
\stackrel{rN(1-\frac{N}{K})}{\longrightarrow} & & & \\
\hline
N & \overline{a}(\overline{n},\overline{m})MN & & & \\
\hline
\end{array}$$

$$\begin{array}{c}
e\overline{a}(\overline{n},\overline{m})MN \\
\hline
M & & \\
\end{array}$$

#### Variables

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

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

## **Evolutionary Components**

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

$$\frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}} = \beta_G^2 \frac{M(\theta + \overline{n} - \overline{m})}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n})$$

$$\frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}} = \sigma_G^2 \frac{eN(\theta + \overline{n} - \overline{m})}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n})$$

#### **Variables**

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

- $\beta_G^2 \equiv \text{Prey genetic variance}$
- $\sigma_G^2 \equiv$  Predator genetic variance

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

#### **Ecological Components**

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

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

#### **Evolutionary Components**

$$\frac{d\overline{n}}{dt} = \beta_G^2 \frac{\partial \overline{Y}}{\partial \overline{n}} = \beta_G^2 \frac{M(\theta + \overline{n} - \overline{m})}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n})$$

$$\frac{d\overline{m}}{dt} = \sigma_G^2 \frac{\partial \overline{W}}{\partial \overline{m}} = \sigma_G^2 \frac{eN(\theta + \overline{n} - \overline{m})}{\sigma^2 + \beta^2 + \tau^2} \overline{a}(\overline{m}, \overline{n})$$

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

#### Extinction

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

#### **Exclusion**

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

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

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

### **Extinction** *Unstable*

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

#### **Exclusion**

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

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

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

**Extinction** *Unstable* 

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

**Exclusion** Stable under certain conditions

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

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

**Extinction** *Unstable* 

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

**Exclusion** Stable under certain conditions

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

**Necessary Conditions for Stable Exclusion:** 

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

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

#### Coexistence

$$\begin{split} (\textit{N}^*,\textit{M}^*,\overline{\textit{n}}^*,\overline{\textit{m}}^*) &= (\frac{d\sqrt{\textit{A}}}{e\alpha\tau}\;,\;\frac{r\sqrt{\textit{A}}}{\alpha\tau}\left(1-\frac{d\sqrt{\textit{A}}}{\textit{Ke}\alpha\tau}\right)\;,\;\mu^*\;,\;\mu^*-\theta) \\ \text{where } \textit{A} &= \sigma^2 + \beta^2 + \tau^2 \text{ and } \mu^* \text{ is an arbitrary value}. \end{split}$$

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

**Coexistence** | Stable under certain conditions

$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (\frac{d\sqrt{A}}{e\alpha\tau}, \frac{r\sqrt{A}}{\alpha\tau} \left(1 - \frac{d\sqrt{A}}{Ke\alpha\tau}\right), \mu^*, \mu^* - \theta)$$
 where  $A = \sigma^2 + \beta^2 + \tau^2$  and  $\mu^*$  is an arbitrary value.

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

**Coexistence** | Stable under certain conditions

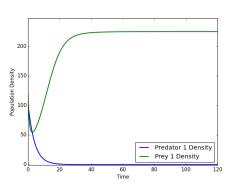
$$(N^*, M^*, \overline{n}^*, \overline{m}^*) = (\frac{d\sqrt{A}}{e\alpha\tau}, \frac{r\sqrt{A}}{\alpha\tau} \left(1 - \frac{d\sqrt{A}}{Ke\alpha\tau}\right), \mu^*, \mu^* - \theta)$$
 where  $A = \sigma^2 + \beta^2 + \tau^2$  and  $\mu^*$  is an arbitrary value.

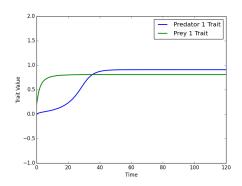
#### **Necessary Condition for Stable Coexistence:**

$$\bullet \ d\sigma_G^2 > r\beta_G^2 \left(1 - \frac{d\sqrt{A}}{Ke\alpha\tau}\right)$$

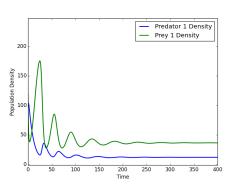


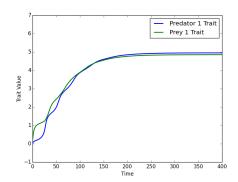
#### **Exclusion**



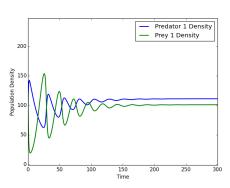


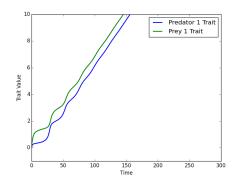
#### Stable Coexistence





#### **Unstable Coexistence**





# Ask us about our preliminary $1 \times 2$ results!

#### **Prey Fitness**

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

#### **Predator Fitness**

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

#### **Prey Fitness**

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

$$\downarrow$$

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

#### **Predator Fitness**

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

#### Notation

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



#### **Prey Fitness**

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

$$\downarrow$$

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

#### **Predator Fitness**

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

$$\downarrow$$

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

#### Notation

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

## Average Fitness

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

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

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

#### **Ecological Components**

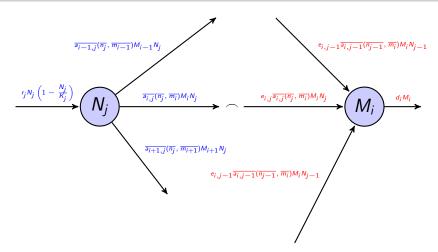
$$\frac{dN_{j}}{dt} = N_{j}\overline{Y}_{j} = N_{j}\left[r_{j}\left(1 - \frac{N_{j}}{K_{j}}\right) - \sum_{i=1}^{u} M_{i}\overline{a}_{ij}(\overline{m}_{i}, \overline{n}_{j})\right]$$

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

#### **Evolutionary Components**

$$\begin{split} \frac{d\overline{n}_{j}}{dt} &= \beta_{Gj}^{2} \frac{\partial \overline{V}_{j}}{\partial \overline{n}_{j}} = \beta_{Gj}^{2} \sum_{i=1}^{u} \left[ \frac{M_{i}(\theta_{ij} + \overline{n_{j}} - \overline{m_{i}})}{\sigma_{i}^{2} + \beta_{j}^{2} + \tau_{ij}^{2}} \overline{a}_{ij}(\overline{m_{i}}, \overline{n_{j}}) \right] \\ \frac{d\overline{m}_{i}}{dt} &= \sigma_{Gi}^{2} \frac{\partial \overline{W}_{i}}{\partial \overline{m}_{i}} = \sigma_{Gi}^{2} \sum_{j=1}^{v} \left[ \frac{e_{ij} N_{j}(\theta_{ij} + \overline{n_{j}} - \overline{m_{i}})}{\sigma_{i}^{2} + \beta_{j}^{2} + \tau_{ij}^{2}} \overline{a}_{ij}(\overline{m_{i}}, \overline{n_{j}}) \right] \end{split}$$

## 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?



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

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

#### Extinction

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

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

**Extinction** *Unstable* 

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

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

## **Extinction** *Unstable*

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

#### **Exclusion**

$$(\textit{N}_{1}^{*},\textit{N}_{2}^{*},\textit{M}^{*},\overline{\textit{n}}_{1}^{*},\overline{\textit{n}}_{2}^{*},\overline{\textit{m}}^{*})=(\textit{K}_{1},\textit{K}_{2},0,\_,\_,\_)$$



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

**Extinction** *Unstable* 

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

**Exclusion** Stable under certain conditions

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



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

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

#### **Generalist Becomes Specialist**

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

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

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

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

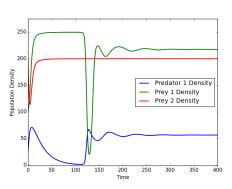
## Generalist Becomes Specialist | Stable under certain conditions???

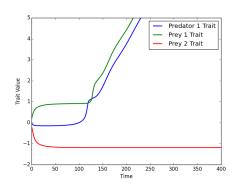
$$(N_1^*, N_2^*, M^*, \overline{n}_1^*, \overline{n}_2^*, \overline{m}^*)$$

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

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

#### **Generalist Becomes Specialist**





#### **Unstable Coexistence**

