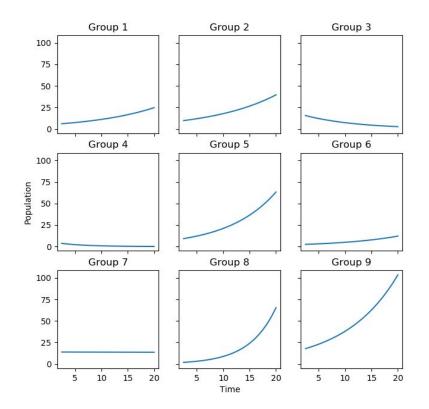
### Modeling Taylor's Law with Exponential Growth and Migration

**Dr. Ben Webb, Clark Brown, Sam Carpenter Presented by Clark Brown**BYU SRC February 29, 2020



## What is Taylor's Law?

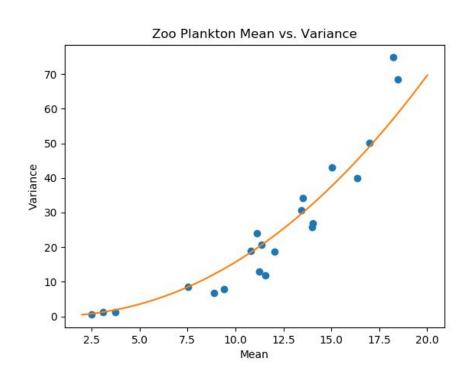
#### **Zooplankton Example**





1	2	3
4	5	6
7	8	9

#### **Zooplankton Example**



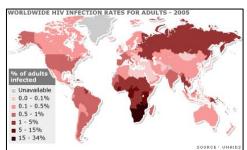
$$Var(X) = a \cdot E(X)^b$$

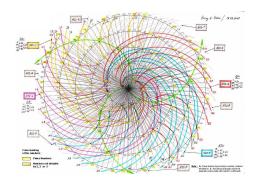
Time	Mean	Variance
0	9.4	13.7
5	11.1	19.6
10	13.5	29.9
15	12.0	23.4
20	18.5	58.8
25	13.9	32.2
30	10.8	18.5

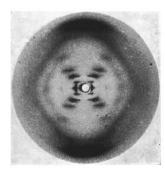
#### **Empirical Observations**

- Thousands of Species Populations
- Cancer Metastasis
- HIV Epidemiology
- Homes Built over Tonami Plain in Japan
- Distribution of Stores in Beijing
- Gene Structure
- Distribution of Prime Numbers











Images from zekkeijapan.com, proofmathisbeautiful.tumblr.com, bbc.co.uk, jstor.org

#### **Taylor's Law Definition**

Let  $N_i(t)$  be the i<sup>th</sup> subpopulation and N(t) be a vector of our subpopulations.

$$Var(\mathbf{N}(t)) = a \cdot E(\mathbf{N}(t))^{b}$$

#### **Taylor's Law Definition**

Taylor's Law holds *in the limit of large time* if there exist a > 0 and b such that

$$\lim_{t \to \infty} [\log(\text{Var}(\mathbf{N}(t))) - b \cdot \log(E(\mathbf{N}(t)))] = \log(a)$$

#### **Taylor's Law Definition**

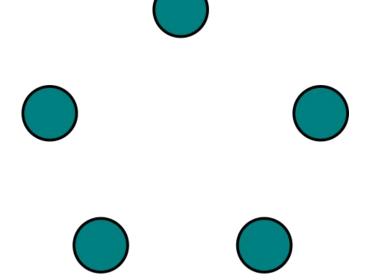
Taylor's Law holds in the limit of large time if there exist a > 0 and b such that

$$\lim_{t \to \infty} b(t) = \frac{d \log(Var(\mathbf{N}(t)))}{d \log(E(\mathbf{N}(t)))} = 2$$

# A Simple Model No Migration

#### Simple Model

$$R = \begin{bmatrix} r_{11} & 0 & \dots & 0 \\ 0 & r_{22} & \dots & 0 \\ \vdots & & & \vdots \\ 0 & 0 & \dots & r_{nn} \end{bmatrix}$$



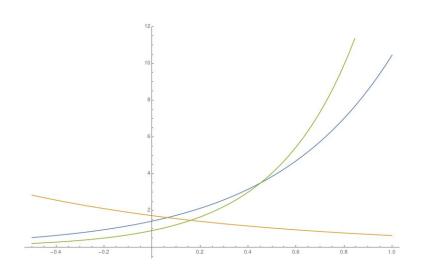
$$\frac{d\mathbf{N}(t)}{dt} = R\mathbf{N}(t)$$

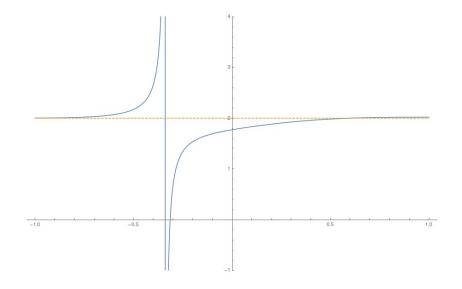
$$\mathbf{N}(t) = e^{Rt}\mathbf{N}(0)$$

From Joel E. Cohen, "Taylor's power law of fluctuation scaling and the growth-rate theorem", Theoretical Population Biology, Volume 88, 2013, Pages 94-100.

#### Simple Model Example

$$R = \left[ \begin{array}{ccc} 2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 3 \end{array} \right]$$



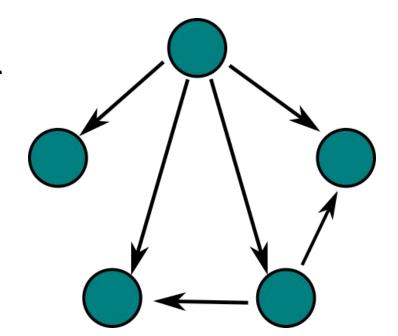


$$\mathbf{N}(t) = e^{Rt}\mathbf{N}(0)$$

# What happens if we introduce migration into the system?

#### **Our Model: Migration Model**

$$R = \begin{bmatrix} r_{11} & 0 & \dots & 0 \\ 0 & r_{22} & \dots & 0 \\ \vdots & & & \vdots \\ 0 & 0 & \dots & r_{nn} \end{bmatrix}$$



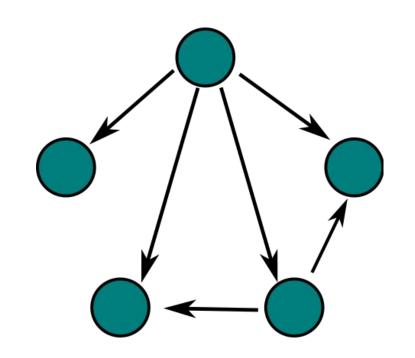
$$M = \begin{bmatrix} -\sum_{j\neq 1} m_{j1} & m_{12} & \dots & m_{1n} \\ m_{21} & -\sum_{j\neq 2} m_{j2} & \dots & m_{2n} \\ \vdots & & & \vdots \\ m_{n1} & m_{n2} & \dots & -\sum_{j\neq n} m_{jn} \end{bmatrix} \text{ where } m_{ij} \geq 0, i \neq j.$$

#### **Our Model: Migration Model**

$$A = R + M$$

$$\mathbf{N}(t) = e^{At}\mathbf{N}(0)$$

$$b(t) = \frac{d \log(Var(e^{At}\mathbf{N}(0)))}{d \log(E(e^{At}\mathbf{N}(0)))}$$

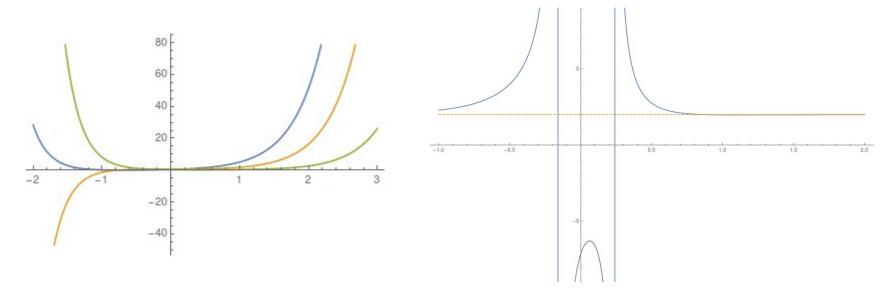


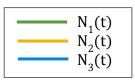
#### Migration Model Example

$$A = \begin{bmatrix} 2 & 1 & 0 \\ 1 & -1 & 1 \\ 0 & 1 & -4 \end{bmatrix} = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -3 \end{bmatrix} + \begin{bmatrix} -1 & 1 & 0 \\ 1 & -2 & 1 \\ 0 & 1 & -1 \end{bmatrix} = R + M$$

$$\sigma(A) = \{-1 - \sqrt{11}, -1 + \sqrt{11}, -1\}$$
$$\lambda_{\ell} = -1 + \sqrt{11} > 0$$
$$\mathbf{v}_{\ell} = (10 + 3\sqrt{11}, 3 + \sqrt{11}, 1) \succ \mathbf{0}$$

#### Migration Model Example





$$\mathbf{N}(t) = e^{At}\mathbf{N}(0)$$

b(t)

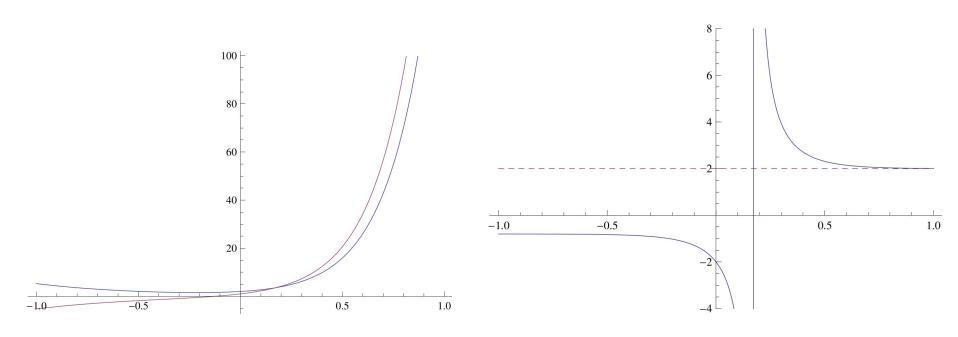
#### Properties of the Model with Migration

**Proposition 1.** In the exponential model with migration, we have:

- (1) If  $\mathbf{N}(0) \succeq 0$ , then  $\mathbf{N}(t) \succeq 0$  for all t > 0.
- (2) The matrix A has strictly non-negative entries off the diagonals.
- (3) Any matrix A that satisfies (2) can be decomposed into unique matrices M and D that represent exponential growth and migration rates.
- (4) There is a  $\lambda_{\ell} \in \mathbb{R}$  with  $\lambda_{\ell} \geq Re(\lambda)$  for all  $\lambda \in \sigma(A)$ . We call  $\lambda_{\ell}$  the "leading eigenvalue."
- (5) There exists an eigenvector associated with  $\lambda_{\ell}$  with strictly non-negative entries. We call  $\mathbf{v}_{\ell}$  the "leading eigenvector."

# Under which conditions does Taylor's Law hold?

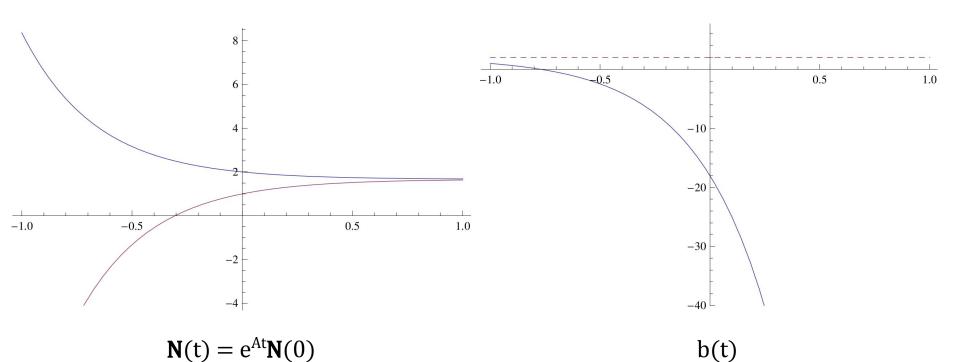
$$A = \begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix} = diag[-5, -5] + \begin{bmatrix} -4 & 3 \\ 4 & -3 \end{bmatrix}$$



$$\mathbf{N}(\mathsf{t}) = \mathsf{e}^{\mathsf{A}\mathsf{t}}\mathbf{N}(0)$$

b(t)

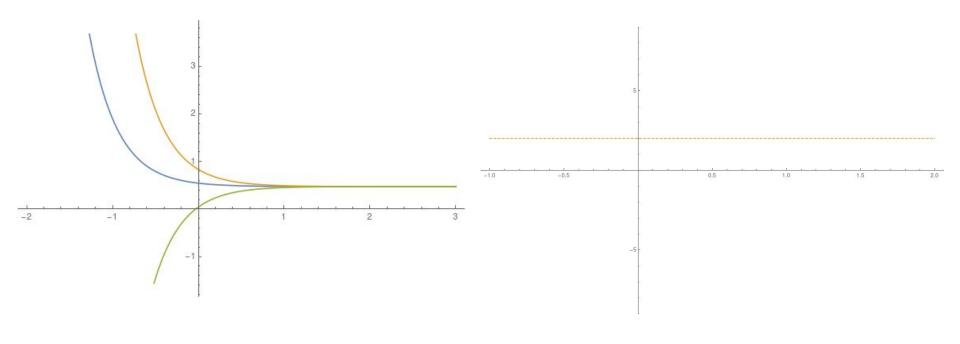
**Example 2** 
$$A = \begin{bmatrix} -1 & 1 \\ 2 & -2 \end{bmatrix} = diag[1, -1] + \begin{bmatrix} -2 & 1 \\ 2 & -1 \end{bmatrix}$$



 $\mathbf{N}(\mathbf{t}) = \mathbf{e}^{\mathbf{A}\mathbf{t}}\mathbf{N}(0)$ 

**Example 3** 
$$A = \begin{bmatrix} -2 & 2 & 2 \\ 2 & -2 & 2 \\ 2 & 2 & -2 \end{bmatrix} = diag[2, 2, 2] + \begin{bmatrix} -4 & 2 & 2 \\ 2 & -4 & 2 \\ 2 & 2 & -4 \end{bmatrix}$$

b(t)



#### **Sufficient Conditions for Taylor's Law**

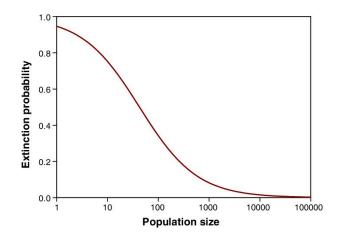
**Proposition 2** (Sufficient Conditions for Taylor's Law). Let  $A \in \mathbb{R}^{n \times n}$  where  $a_{ij} \geq 0$  for all  $i \neq j$ . Let  $\mathbf{v}_{\ell}$  be the "leading eigenvector" as defined earlier. If

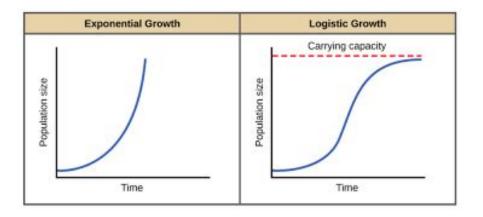
- (1)  $\mathbf{v}_{\ell}$  is non-diagonal, i.e.  $\mathbf{v}_{\ell} \neq k * \mathbf{1}$
- (2)  $c_n$  is nonzero in the solution of  $\mathbf{N}(t) = e^{At}\mathbf{N}(0)$ .

then Taylor's Law holds in the limit of large time.

#### **Future Research Possibilities**

- Modeling Logistic Growth with Migration
- Time to Extinction Prediction





# Thank you

## **Q&A**